COVID-19 and Emerging Markets: An Epidemiological Model with International Production Networks and Capital Flows

Cem Çakmaklı, Selva Demiralp, Şebnem Kalemli Özcan, Sevcan Yeşiltaş, Muhammed A. Yıldırım

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Abstract

We quantify the macroeconomic effects of COVID-19 for a small open economy by calibrating a SIR-multi-sector-macro model. We measure sectoral supply shocks utilizing teleworking and physical job proximity, and demand shocks with credit card purchases. Both shocks are also affected from changing infection rates under different lockdown scenarios. Being an open economy amplifies the economic costs through two main channels. First, the demand shock has domestic and external components. Second, the initial shock is magnified due to domestic and international input-output linkages.

JEL Classification Numbers: E61, F00, C51

Keywords: COVID-19; Debt Monetisation, External Finance; SIR; Input-Output Tables; Emerging Markets

Authors’ E-Mail Addresses: ccakmakli@ku.edu.tr, sdemiralp@ku.edu.tr, SKalemli-Ozcan@imf.org, syesiltas@ku.edu.tr, mayildirim@ku.edu.tr.
COVID-19 and Emerging Markets:
An Epidemiological Model with International Production Networks and Capital Flows∗

Cem Çakmaklı† Selva Demiralp‡ Şebnem Kalemlı Özcan§ Sevcan Yeşiltaş¶
Muhammed A. Yıldırım∥

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Abstract

We quantify the macroeconomic effects of COVID-19 for a small open economy by calibrating a SIR-multi-sector-macro model to Turkey. Sectoral supply shocks are based on the proximity requirements in each sector and the ability to work from home. Physical proximity determines the supply shock through its effect on infection rates. Sectoral demand shocks incorporate domestic and foreign demand, both of which adjust with infection rates. We calibrate demand shocks during COVID-19 using real-time credit card purchase data. Our results show that the optimal policy, which yields the lowest economic cost and saves the maximum number of lives, can be achieved under a full lockdown of 39 days. Economic costs are much larger for an open economy as the shocks are amplified through the international production network. A decline in foreign demand leads to losses in domestic sectors through international input-output linkages, accounting for a third of the total output loss. In addition, the reduction in capital flows deprives the network from its trade financing needs, where sectors with larger external finance needs experience larger losses. The policy options are limited given sparse fiscal resources to fight the pandemic domestically, while serving the external debt. We present historical evidence from 2001 crisis of Turkey, when fiscal, monetary and exchange rate policies were employed altogether to deal with a triple crisis of balance of payments, banking, and sovereign debt.

Keywords: External Finance; Market Sentiment; International I-O Tables; FX Debt
JEL Codes: E61, F00, C51

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†Koç University.
‡Koç University.
§IMF, University of Maryland, NBER and CEPR.
¶Koç University.
∥Koç University & Harvard University.
1 Introduction

The COVID-19 shock may soon lead to the biggest emerging market (EM) crisis of modern times. EMs observe a collapse in domestic and external demand, record capital outflows, higher external borrowing costs, a commodity bust, and depreciating currencies. At the same time, EM governments increase domestic borrowing via unconventional policies to assemble fiscal resources to fight the pandemic. This puts further pressure on external finance premium, reducing capital inflows and making it harder to rollover the external debt. In this paper, we focus on this complex embrace between domestic fiscal needs and external financing needs. In order to estimate the amount of fiscal resources needed to cover the economic costs, we develop an epidemiological Susceptible-Infected-Recovered (SIR)- multi-sector-macro model for a small open economy that is linked to an international production network. Our work differs starkly from the rapidly growing COVID-19 literature that focuses on closed economies, mostly advanced countries.

The key characteristics of several large EMs are that they are part of global supply chains and they have domestic and external debt, both denominated in local currency and in foreign currency (FX). Most of them have low FX reserves in spite of high private sector FX debt. They generally have lower policy credibility relative to advanced economies and most are heavily dependent on capital inflows to meet the demand in their fast growing economies. EMs learned their lessons from the crises of 1990s and early 2000s in the sense that they have manageable fiscal deficits and better capitalized banking systems as they face the COVID-19 crisis. Nevertheless, they will still operate with low fiscal space during their response to the COVID-19 shock given the size of the shock, domestic health needs and capital flow reversals that lead to increasing external financing gaps.

Our analysis contributes to the literature by developing a SIR model at the sectoral level with international I-O linkages. The model is developed to calculate the economic costs of the pandemic.

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1 See Cakmakli et al. (2020) for the earlier working paper version of our model, April (2020).
2 See Arellano et al. (2020) who also focuses on emerging markets by analysing sovereign debt and default risk under COVID shock with a SIR+sovereign debt model.
under different lockdown scenarios. We take into account the effects of the COVID-19 shock on domestic demand as well as foreign demand and related trade and capital flows in an emerging market. Our approach has the advantage of being simple and easily mapped to real time data. We calibrate our model to Turkey by using Turkey’s links to 65 other countries through 36 sectors within the international production network.

We first classify sectors into essential and non-essential ones. The ability to work from home determines the size of the supply shock in non-essential sectors (as also done in the closed economy literature). For the essential sectors, the nature of the physical proximity of the job dictates the supply shock because infection rates are higher in those sectors. A pure economic model in the absence of a SIR model might incorrectly assume a zero supply shock in the essential sectors.

Turning to the demand side, our model contains a domestic component and a foreign component for sectoral demand shocks. Demand declines as the number of infections increase. We have a short run model where the output is demand determined with fixed prices. We exploit international input-output (I-O) linkages to transform changes in final foreign demand into changes in demand for domestic intermediate sectors. Analysis at the sectoral level is critical because economic disruptions are likely to be more severe and protracted in those sectors with greater exposure to international spillovers. We use real time data on credit card purchases to pin down the COVID-19 related demand shocks.

Once we estimate the economic cost for each sector from the model, we document the role of international spillovers in sectoral economic costs. To that end, we empirically link each sector’s economic loss to its I-O linkages with the other domestic and international sectors as well as its external financing needs. The I-O link to other sectors is measured as a weighted average of the links to countries that Turkey trades with. External finance needs are captured by using data on country-pair capital flows. We show that sectors with stronger I-O links suffer from larger COVID-19 related losses and sectors who finance these stronger production links through capital flows suffer even more.

Our model based estimate for the total cost of containing the pandemic immediately, with a

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3 Each sector is linked to other countries such that the links with some countries are stronger through intermediate inputs trade. We use country-pair specific capital flows to gauge the external finance needs of that sector. See also Kalemi-Ozcan et al. (2014).
Chinese style full lockdown is about 5.8 percent of the GDP (at an annualized rate). This implies that output declines by 17.5 percent during the quarter in which the lockdown is imposed, compared to the previous quarter. After the lockdown ends, if the economy returns to normal during the rest of the year, as demand normalizes, then the shock is smoothed out, leading to a decline of 5.8 percent of annual GDP. We show that, almost 30 percent of these economic costs stem from external demand.

Contrary to the popular belief that no lockdown policies would minimize economic costs, we show that such policies are actually costlier than an effective full lockdown by bringing forward the importance of the demand side. In fact, under no lockdown, the economic cost increases from 5.8 to 11 percent of GDP annually. Even though businesses remain open, there are still interruptions in supply, and demand declines as people get infected. Full lockdown, on the other hand, is optimal since it is able to contain the pandemic more quickly, within approximately one month. Hence it yields the minimum economic cost which saves the maximum number of lives.

In no lockdown scenario, most of the population is fully exposed to the outbreak. Nevertheless, the working population is under higher risk compared to the non-working population. In partial lockdown scenario, teleworkable occupations start working from home and hence the basal infection rate declines for this group. It is important to note that the individuals in the highest risk group, ages 65 and above, as well as the younger people are assumed to have lower infection rates either because they do not work or because they switch to distanced learning. This is consistent with the optimal setting identified by Acemoglu et al. (2020). The infection rate is still high for the on-site workers. In full lockdown, we assume that only the essential sectors require their non-teleworkable employees on-site. The infection rate declines substantially for the remainder of the population that stays home.

Several recent closed economy papers employing epidemiological models similar to us, including Acemoglu et al. (2020), Alvarez et al. (2020), Farboodi et al. (2020), and Eichenbaum et al. (2020) reach similar conclusions where imposing full lockdowns or stricter measures at the early stages of the pandemic lower economic costs. Although none of these papers include sectoral demand shocks within a SIR framework, the aggregate demand operates in a similar way. We argue that, for an open economy, sectoral demand is indispensable. The decline in external demand amplifies the domestic demand shock via sectoral I-O linkages. We show that once the number of infections reach a certain threshold, demand stalls and remains rather sluggish so long as the infection num-
bers do not exhibit a substantial decline. In fact, even if the domestic infection numbers are reduced, economic recovery will not be complete until the pandemic is contained abroad and foreign demand improves consequently. Thus, the course of the pandemic abroad affects domestic sectors differentially via international I-O linkages. In this framework, even if the supply channels remains unrestricted, demand drags the equilibrium output down and elevates the size of economic costs so long as the health issues are not resolved.

Our model based costs are estimated in the absence of any policy action. Costs will naturally decline when fiscal and monetary policy responses are taken into consideration. We prefer to provide our baseline estimates based on no policy action so that the minimum magnitude of the fiscal policy packages can be clearly identified. This approach also makes our findings relevant if there is a second wave after full opening where countries need to go back to lockdowns. If the economy opens up prematurely, the increase in the number of infections would stall demand again, even if the businesses remain open. The consequent economic costs may lead to lasting economic damage by extending the duration of the recession. Indeed, we show that the duration of a lockdown that is needed to contain the virus increases to more than one year, if the lockdown ends prematurely.

Turkish experience can be considered as an “enhanced partial lockdown” which is a mixture of full lockdown and partial lockdown periods. Thus, the annual costs for Turkey might fall within our range of 5.8 and 11 percent of GDP. It is hard to compare model based estimates to real data from emerging markets, without the release of second quarter GDPs as most EMs observed the pandemic later than advanced economies and started their lockdowns almost at the end of the first quarter. In our model, we have a 39 day full lockdown that is implemented relatively early in the pandemic. During the quarter that the lockdown is implemented, output declines by about 17 percent, relative to the previous quarter. If we focus only on the period of full lockdown of 39 days, then the output loss is 22.5 percent during the duration of the lockdown. If we do the same exercise for the 240 days of a partial lockdown, then the output loss is 14.2 percent as economy is partly open. However, when we compare the annual costs, the economic costs of partial lockdown are about twice as much of a full lockdown. In terms of comparison to real numbers, industrial production in Turkey fell 7

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4See [https://www.thelancet.com/journals/lanpub/article/PIIS2468-2667(20)30073-6/fulltext](https://www.thelancet.com/journals/lanpub/article/PIIS2468-2667(20)30073-6/fulltext) that argues that reopening too soon before the R number is below one might trigger another peak. The case of Singapore is an example with recurring lockdowns: [https://www.theguardian.com/world/2020/apr/21/singapore-coronavirus-outbreak-surges-with-3000-new-cases-in-three-days](https://www.theguardian.com/world/2020/apr/21/singapore-coronavirus-outbreak-surges-with-3000-new-cases-in-three-days)
percent in March, 2020 and 32 percent in April, 2020. Our estimates are not far from these numbers on industrial production, on average, for Turkey.

Our model and findings are also consistent with the early experiences of New Zealand, Denmark and Greece. These countries implemented full lockdown before the number of patients reached critical levels and contained the virus rather rapidly. Consequently, they gradually began to lift lockdown restrictions before the end of April. Another country is South Korea, who did not implement a full lockdown but did extensive testing and contact tracing. If we look at industrial production numbers for April, we observe that the decline in industrial production in these countries is similar to our baseline estimates. For example, in Greece, industrial production shrunk 12.3 percent in April compared to previous month, respectively, which is similar to our 17 percent loss. Greece and South Korea annualized GDP growth rates turn out to be -6.2 and -5.3 percent, respectively, which is close to our -5.8 percent annualized estimate.

After estimating the economic costs of COVID-19 under alternative scenarios, we investigate how to finance the economic costs of the lockdown and the contraction in GDP. Financing these costs need sizable resources that can be obtained though a combination of domestic and external finance. For a typical EM, a large part of its external debt is in the form of domestic banks borrowing from global banks and hence debt relief from official lenders will not help. Turkey is no exception to this typical pattern in external debt as we document in section 3. If domestic banks cannot rollover their external debt due to increased borrowing costs and insolvency risks of domestic firms, then there is a potential threat not only to domestic financial stability but also to global financial stability from elevated stress in such EMs.

In terms of domestic policies, the risk of a deep recession could be reduced under a targeted and transparent asset purchase program by EM central banks, accompanied by external funding granted by an international institution in the absence of any financing from international capital markets. The literature shows that Quantitative Easing (QE), that is asset purchases programs, require policy credibility to keep inflation expectations under control so that such programs do not turn into long term monetary financing of government debt. With rising country risk premia under COVID shock,

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5Recent work suggests that testing and tracing can mimic the benefits of a full lockdown. See Zilibotti et al. (2020).
6Avdjiev et al. (2017), Bruno and Shin (2015).
7See https://www.brookings.edu/wp-content/uploads/2019/12/Bernanke_ASSA_lecture.pdf for Chairman
the perceptions of global investors and the market sentiment will be key determinants of capital flows. In this context, international financial institutions and central banks of reserve currency countries such as the Federal Reserve, can play a key role in assuring global investors and changing the risk perceptions so that EMs will have access to international financial markets.

The remainder of this paper is organized as follows: In Section 2, we provide an overview of the literature on COVID-19 pandemic, explaining our contribution. In Section 3, we briefly go over the environment in which Turkey entered the COVID-19 crisis, the policies adopted by Turkey to deal with the pandemic so far, and compare to other countries. Section 4 describes the model that allows us to estimate the sectoral and aggregate COVID costs for the Turkish economy. Our quantitative findings are summarized in Section 5, where we show that costs are larger for an open economy, such as Turkey, with strong links to an international production network and capital flows that finances the network. Section 6 considers the policy alternatives to finance the economic costs of the pandemic related crisis with their pros and cons. Section 7 describes the historical experience. Section 8 concludes.

2 COVID-19 Literature and Our Contribution

There is a rapidly growing literature that aims to capture the economic impact of COVID-19 crisis. Many papers utilize SIR models or its extensions to incorporate the infection dynamics into their analysis. Papers such as Stock (2020) and Alvarez et al. (2020) consider a standard SIR model and focus on the trade off between unemployment that arises from lockdowns versus the number of deaths due to the pandemic. They reach the conclusion that the optimal policy is a full lockdown that covers the majority of the population where the restrictions are removed gradually afterwards.

Acemoglu et al. (2020) considers a multi-risk SIR model by focusing on the structural differences in the severity of infections for distinct age groups that affect lockdown policies and economic costs. They show that targeted measures such as full lockdown for the elderly group could be more effective. Alon et al. (2020) also considers a closed economy model but approaches the problem from the developing country perspective, considering market distortions and the presence of an informal

Bernanke’s ASSA lecture and the references therein.
sector and hand to mouth consumers. They realize that such economies cannot fully lockdown and argue that lockdowns on the elderly population might be better.

Combining supply and demand in a SIR framework but with no sectoral heterogeneity, Farboodi et al. (2020) internalizes the individual choices for social distancing and study both laissez-faire and social optimum scenarios. They find that even in the laissez-faire case individuals choose to sharply reduce their activity but the socially optimal response imposes severe restrictions at the onset of the outbreak. Eichenbaum et al. (2020) incorporate supply and demand in a SIR model as well, where the government is assumed to alter the individuals’ activities through a consumption tax and again find that relatively severe containment at the beginning of the pandemic is the most socially optimum response. Krueger et al. (2020) extends the model by Eichenbaum et al. (2020) and introduces differential transmission rates based on the consumption or employment choice. They aim to capture the interplay between infection dynamics and the demand side or the supply side—but not both of them simultaneously—highlighting the importance of sectoral heterogeneity.

A parallel trend in the literature considers the effect of I-O linkages. Barrot et al. (2020) consider the effects of sectoral decline in production based on the lockdown measures in France. Bonadio et al. (2020) studies the propagation of supply shocks caused by the protection measures against the COVID-19 pandemic through inter-country input-output linkages. Similarly, incorporating I-O linkages, Baqee et al. (2020) analyzes the consequences of sectoral supply shocks caused by the pandemic using an SIR model. They show that a strong economic reopening is possible with restrictions on non-work social contacts to avoid a second wave. Baqee and Farhi (2020b) considers a Keynesian model and considers supply shocks at the sectoral level while demand shocks are in aggregate form. They show a large amplification of both aggregate demand and sectoral supply shocks in a network economy with input-output linkages, absent an infection dynamics model. The work by Guerrieri et al. (2020) do not include an infection dynamics model either but underlines the importance of a multi-sector economy, where supply shocks can turn into larger aggregate demand shocks.

In a model with no infection dynamics (SIR), Baqee and Farhi (2020a) exploits nonlinear production networks in a general equilibrium framework and show that non-linearities amplify the impact of COVID-19 between 20 to 100 percent. They illustrate that reductions in labor supply are less costly to aggregate output if they are spread out across many sectors.
To the best of our knowledge, our paper is the first to bridge these trends in the literature by incorporating both supply and demand shocks at the sector level for an open economy operating within the international production network\textsuperscript{10}. We estimate our COVID losses by considering the effects of domestic and foreign sectoral shocks and their amplification through I-O linkages. Linking the economic losses to I-O trade links and capital flows empirically, provides further evidence on the fiscal needs of EMs.

Figure 1: The effects of COVID-19 in a multi-sector open economy: A Schematic of the Model

\textbf{NOTES:} We implement two main lockdown scenarios: \textit{partial} and \textit{full}. Under partial lockdown, all industries remain open while the teleworkable portion of the employees work from home. The restrictive measures result in a low infection rate for the teleworkables and the general public, but the infection rate remains high for the on-site workers. Under full lockdown, only the essential industries remain open and the workers in the non-essential sectors stay at home. With these extreme measures, the infection rates are lowered for almost everyone. The lockdowns affect the supply channel directly via workers and the demand channel by mitigating the number of infected individuals, which in turn change the consumption profiles.

Figure 1 summarizes our theoretical framework. We ponder the figure for a given industry. After we separate sectors into essential and non-essential sectors, we capture supply shocks by quantifying how susceptible each industry is to the transmission of the virus among its employees. The transmission dynamics of the virus would differ depending on whether the workers are on-site or

\textsuperscript{10}The only other paper that we are aware of considering both demand and supply shocks at the sectoral level is by del Rio-Chanona et al. (2020) with an extension in Pichler et al. (2020). Their analysis is also for an open economy, focusing on the UK. They incorporate domestic demand shocks for the UK based on hypothetical predictions of an influenza pandemic from a CBO report in 2006.
at a remote location like home. We use Dingel and Neiman (2020)’s list of teleworkable occupations to capture the proportion of employment that can be fulfilled at remote locations in each industry. Among the professions that need to be carried out on the work site, we assume that the viral transmission depends on the physical proximity between the workers or between the workers and the customers. An on-site worker could be exposed to infection either at work or outside work. Using the teleworkable share of an industry and the physical proximity measure as part of the SIR model, we estimate the proportion of the work force in each industry that would be impaired during the time of the pandemic. Of course, the viral transmission dynamics will be affected by the implementation of different lockdown policies. Thus, we calculate our sectoral supply shock under different scenarios. Under partial lockdown scenario, we assume that all businesses are open, but the teleworkable share of the employees remain home. The viral transmission is lower among the teleworkable employees and the general public, but the transmission rate is still high among the on-site workers. Under full lockdown scenario, we assume that all businesses except the essential ones are closed and all employees working in the closed sectors remain home. The viral transmission rates drop to a lower level for all the workers in the non-essential sectors.

The pandemic affects the demand side as well. The economics profession unanimously agrees that the prerequisite for economic recovery is the elimination of the virus so that demand normalizes.\textsuperscript{11} Former Federal Reserve Chairman Bernanke noted in late March that “Nothing will work if health issues aren’t resolved,” sending a clear message to governments.\textsuperscript{12} Unless we contain the virus, economic confidence will not return, businesses will not open and people will not return to their normal lives or maintain their usual patterns of consumption. As a result, demand shocks will play a key role in this crisis, where standard short-run policies to stimulate demand will not work so long as people stay fearful.

We have a disproportionate role of demand in our model as we focus on both domestic and external demand shocks, where shocks are propagated through domestic and international input-output linkages. In the upper half of Figure 1, we illustrate the changes in demand due to the pandemic that ultimately affect the equilibrium output. We consider two scenarios for demand:

\textsuperscript{11}See IMF World Economic Outlook, April 2020. Also contributions in Baldwin and di Mauro (2020).
\textsuperscript{12}The transcript of Bernanke’s interview on March 25 is available at this link: https://www.cnbc.com/2020/03/25/cnbc-transcript-former-fed-chairman-ben-bernanke-speaks-with-cnbc-andrew-ross-sorkin-on-squawk-box-today.html
one for the normal times and one during the brunt of the pandemic. To proxy for demand shocks during the peak of the pandemic, we use data on credit card purchases provided by the Central Bank of the Republic of Turkey (CBRT). For the few sectors where the credit card data is missing, we use other proxies for real-time demand reduction. For pre-COVID demand, we use consumption spending from Turkish national accounts. During the course of the pandemic, we expect demand to adjust from its pre-COVID levels to the lowest possible level under COVID shock. We model this adjustment with a reduced form function where demand deviates from its normal patterns as a function of the number of infected people. Hence, the demand profile changes depending on the infection levels in the population, which, in turn, is mitigated by the lockdown decisions. The sooner the infection numbers decline, the sooner demand normalizes.

Our open economy framework makes the role of global coordination clear. If the lockdown can be implemented with global synchronization, the pandemic will be controlled faster. As the number of infections declines globally, demand returns to pre-pandemic levels faster as both domestic and foreign demand normalize sooner. Thus, the economic costs of the pandemic can be kept at a minimum level. The last stage in Figure 1 combines demand and supply sides together to reach market equilibrium, where the minimum of both sides determines the equilibrium level of production for a given industry.

3 The Initial Conditions and External Financing Needs

3.1 Background

This section summarizes the economic environment in Turkey before the pandemic to provide a background on initial conditions. Initial conditions when the countries enter the COVID crisis matter because existing vulnerabilities such as high debt, low FX reserves, weak balance sheets, and limited policy credibility will exacerbate the impact of the crisis.

Since 2017, the inflation rate had been on the rise while Turkish Lira (TL) depreciated. Triggered

\[\text{Similarly, Andersen et al. (2020) use transaction-level customer data from the largest bank in Denmark to estimate consumer responses to the COVID-19 pandemic and the partial shutdown of the economy, see also Carvalho et al. (2020) who uses similar data from BBVA for Spain.}\]
by the political tension between Turkey and US, August 2018 marked the beginning of an exchange rate crisis, where rapidly depreciating TL brought many companies with FX debt to the edge of bankruptcy. The significant decline in economic growth led to an improvement in the current account deficit because Turkey’s production heavily relies on imports of intermediary goods. The growth rate in the first quarter of 2020 reached 4.5 percent and the unemployment rate declined to 12.7 percent.

Capital outflows by non-residents during COVID-19 led to a wave of depreciation in TL, which required FX interventions and brought FX reserves to low levels. As of the first week of April 2020, net reserves of Central Bank of the Republic of Turkey (CBRT) stood at merely $26 billion, of which $25 billion was borrowed from domestic banks. IMF-defined budget deficit that excludes one-time transfers stands close to 5 percent of GDP while the current account deficit is around 2.5 percent of GDP, as an average over the last 5 years.

Figure 2: External Debt and Currency Decomposition

(a) Decomposition in terms of Ownership
(b) Currency Decomposition

NOTES: (a) This panel plots external debt (right x-axis) alongside with its public-private composition (left x-axis) for Turkey. Debt values are expressed as percentage of GDP. (b) This panel shows the currency composition of total external debt as of December 2019. Source: Turkey Data Monitor

Turkey relies heavily on capital flows to finance its external debt, which stood at 60 percent of
GDP at the end of 2019. Figure 2a shows the changes in the composition of external debt over time. In 2001, total external debt was 57 percent of GDP. Of this, public sector debt was 24 percent, while the private sector debt was 22 percent.14 Macroprudential measures that were implemented in the aftermath of the 2001 crisis led to a substantial reduction in total external debt in the years immediately after the crisis. Nevertheless, the abundant liquidity provided by the major advanced country central banks in the post-2008 period as part of the widescale QE programs, changed the borrowing patterns in Turkey. The external debt gradually increased with the composition tilting towards private sector borrowing. By the time we reached 2019, total external debt was once again comparable to 2001 levels with 56 percent of the GDP. Different from 2001, however, this time the lion’s share was held by the private sector debt which was 36 percent of the GDP while the public debt was 21 percent of GDP. As of December 2019, almost 60 percent of total external debt is denominated in USD (see Figure 2b). The change in the composition of debt has immediate implications for the policy prescriptions needed to address the private sector debt problem during COVID-19 crisis and bring forward the importance of well designed policies to address loan restructuring, non-performing loans, and non-viable firms.

In terms of maturity structure, out of a total external debt of $437 billion, $124 billion was short-term (17 percent of GDP), and $93 billion of this was held by the private sector. BIS data reflects that $96 billion of the total external debt belongs to the banking system. Meanwhile, the external debt that needs to be rolled over in 2020 is $169 billion, which is approximately 23 percent of GDP. The banking sector’s share in short term debt is $81 billion.15 If the rollover ratios stay at the current levels, then Turkey needs around $30 billion, however, if they go down to the level observed during Great Financial Crisis (GFC), then Turkey might need around $90 billion in 2020, a number that is much larger than any existing swap line and international arrangement available for EMs.16

In terms of market sentiment and global investors’ risk perceptions, EMs seem to be in the middle

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14The sub-components do not add up because the remainder of the external debt is held by CBRT.
15We use the annual GDP of 2019 to express the January 2020 values as a percentage of GDP in this section.
16In a recent report, Bürümçekçi (2020) notes that the current rollover ratio for the banking system is around 73 percent, which receded to 45 percent during 2007-2009 crisis, and 35 percent during 2001 crisis. For the public sector, rollover risk looks minimal for 2020. Turkish government already issued $4 billion of Eurobonds so far, and another $2 billion is due in June 2020. Indeed, when we look at the last 20 years, we note that there were only three instances where debt coming due was not covered with new issuance. Yet none of these events resulted in a missed payment or recontracting of debt. The closest case of partial default was during 2001-2002 crisis that was avoided due to IMF funding. See Figure A.2 in Appendix based on Stoppok and Trebesch (2020).
of a strong risk-off shock, as risk premium, measured by five-year CDS premium increased sharply (See Figure 3). Given the surge in the sovereign risk premium after the COVID-19 shock, rolling over existing external debt would be much costlier, which could raise the fiscal deficit, leading to a prolonged period of higher indebtedness. From the beginning of the year until the week of April 24, 2020, $2.7 billion of equity and $5.5 billion of government bonds held by foreign investors were sold-off to domestic investors in the secondary market\(^{17}\). These numbers may not be as big in the context of total external debt but notice that these are local currency government bonds that were held by foreign investors. As local currency bonds become riskier with the ongoing depreciation of TL, foreigners load-off these bonds first.

**Figure 3: The Risk Premium as Measured by CDS Spread**

![CDS Graph](image)

**Notes:** This figure plots risk premium for Turkey, Brazil, South Korea, India and Argentina which are measured by the 5-year CDS rate (World Government bonds) for these countries. Panel (a) shows the raw values and Panel (b) shows the normalized values. Source: Bloomberg.

In addition to bonds and equities held by non-residents, more than 1/3 of total external funding is obtained through bank loans in Turkey, which is almost all in FX. These loans finance the foreign currency debt in the non-tradeable sector. Half of the entire corporate sector debt is in FX and most of it is borrowed from domestic banks\(^{18}\). To dig deeper into the short-term risks, and considering the market dynamics in the aftermath of the 2007-2009 global financial crisis for EMs, we also need to

\(^{17}\)As for corporate bonds, the sell off started in the last week of February but the total volume of these transactions are rather negligible with a total of $86.5 million outflow from January 3 to April 24, 2020. This is due to the low share of corporate bonds relative to bank loans and government bonds in external debt. See di Giovanni et al. (2019).

\(^{18}\)See di Giovanni et al. (2019).
look at cross-border loans. As shown in Figure 4, Turkish banks had been net payers in the external long-term loans for a while.

Figure 4: Rollovers of External Loans by Turkish Banks

NOTES: This figure plots rollovers of external loans belonging to Turkish Banks. Loan rollovers refer to monthly-net values expressed in terms of millions of USD. Source: Turkey Data Monitor.

In January 2020, Turkish banks paid to foreign financial institutions a net of $0.8 billion over what they borrowed in short-term loans and $0.7 billion in long-term loans. While short term loan rollovers have improved in the next two months, the situation worsened for long term rollovers. In April 2020, Turkish banks paid $1 billion in long-term loans. This suggests that they need to borrow large amounts each month to prevent any interruptions in their domestic lending at home.

Overall, these numbers suggest that, there is quite a bit of foreign investment still in the country given the extent of external debt. Thus, although there are still many horses in the barn, it is important not to scare the horses given the extent of FX debt in the corporate sector, which can lead to massive bankruptcies with a spiraling depreciation of TL. In fact, while most sectors are adversely affected from COVID-19, those sectors with higher levels of FX exposure are hit harder because of the increase in their debt burden during the COVID-19 shock. In Figure 5, we plot the
sectoral FX debt against the economic loss from the pandemic under the scenario in which no action is taken against the pandemic. While we elaborate on how we calculate the economic costs in the next section, we present the loss variable somewhat prematurely in this section to illustrate that the exchange rate depreciation works as an amplification mechanism during the pandemic. In fact, those sectors that rely more heavily on FX funding experience sharper declines in their output during the COVID crisis. As we show below, this relation is not about a sector being tradeable or not but rather how strong is the sector’s connection to international I-O links and how high the sector’s external financing needs which can be financed both domestically and externally, mostly in foreign currency.
3.2 Policy Response to COVID-19

In terms of monetary and financial policies, CBRT cut rates by 100 basis points immediately during their emergency meeting on March 18, 2020 and again on April 22. The announcement that came on March 31 eased collateral requirements to borrow from the CBRT and opened the door for unlimited bond purchases where it was stated that “…limits might be revised depending on market conditions.”\(^{19}\) CBRT and BRSA (Banking Regulation and Supervision Agency) introduced several financial repression measures in the following days that increase the risk exposure of the banking system, encouraging banks to lend at low rates or buy government bonds.\(^{20}\) They have also introduced certain capital flow management measures that reduced domestic banks’ reserve requirements for foreign currency deposits and put limits on the daily amounts of domestic banks’ swap transactions.\(^{21}\)

Notice that although it is important to react early to the pandemic, for an EM, market perception of such measures is just as important. This is because potential risks and hence the external borrowing costs are priced by global investors. Thus, effective and transparent communication of policy actions is as critical as the actions themselves.

In terms of fiscal policy, the stimulus package announced by government on March 18 is consistent with the general framework adopted by other countries. There is postponement of tax obligations, social security premiums and credit payments of the companies in the services sector. The limits of the Credit Guarantee Fund have been increased to make bank loans more accessible. Temporary income support is provided to those workers whose companies have ceased production due to the pandemic. Furthermore, a cash assistance program for needy families has been launched. While the original package announced on March 18 was announced to be 2 percent of GDP, the scope of the package has been expanded in line with the evolving conditions. The Minister of Finance and Treasury announced on May 29 that the pandemic related government expenditure has already reached 260 billion TL. Even with the revised numbers, however, the package still remains

\(^{19}\)https://www.tcmb.gov.tr/wps/wcm/connect/EN/TCMB+EN/Main+Menu/Announcements/Press+Releases/2020/ANO2020-22


to around 5 percent of GDP. To put this number into perspective, Figure 6 shows a comparison of the fiscal measures undertaken by the G20 countries, where the average size of the fiscal stimulus is about 10 percent with Germany leading the pack with 32 percent. It is clear that the Turkish package is small, lagging behind 16 of the G20 countries. This reflects the limited fiscal space of EMs relative to advanced economies.

Figure 6: Fiscal Measures announced by the G20 countries

Notes: This figure plots the COVID-19 relief packages adopted by the countries as a percentage of their GDPs. The fiscal policy measures that are shown in this figure are obtained from the IMF Policy Tracker [https://www.imf.org/en/Topics/imf-and-covid19/Policy-Responses-to-COVID-19] as of April 29, 2020 except for South Africa and the United Kingdom, for which the values of the stimulus packages are not explicitly stated. For these countries, we gathered the necessary information from alternative resources. A detailed comparison of the fiscal measures as well as the data sources are presented in Table A.1 of the Appendix.

In terms of the direct transfer payments, the stimulus package contains several channels of social assistance such as transfer payments for the needy families, enhanced employment protection by loosening short-term work allowance rules, a temporary ban on layoffs with a state subsidy for affected workers and unemployment insurance benefits. These transfer payments add up to 12 billion TL, which is only about 4.6 percent of the total stimulus package. Given the asymmetric nature of the crisis that hit the lower income groups much harder, transfer payments are critical to
provide the much needed income relief and revive demand. Should the transfer payments cover only those who lost their income or should they be in the form of “helicopter money” in the Turkish context, which has a sizeable informal economy? The answer depends on the extent of funding that is available to support the economy. A generous program that does not trigger longer term macroeconomic imbalances can minimize the economic damage and prevent long term risks. A less comprehensive program, on the other hand, would delay the speed of recovery. In the next section, we provide a model that estimates these costs and offers important input for the policy makers regarding the necessary size of the stimulus package.

4 Estimating the Economic Costs Under Different Lockdown Scenarios

In this section, we develop a model that illustrates how COVID-19 affects the economy. We illustrate that despite the increasing costs due to business closures, a full lockdown contains the virus in the fastest way. As we compare the recovery paths with and without the lockdown, we observe that a full lockdown lasts for approximately 40 days while partial lockdown cannot contain the virus within a year. Because the duration of the lockdown increases substantially, the economic costs of a partial lockdown are significantly higher than full lockdown. The mortality numbers present a stark contrast across alternative scenarios as well. Full lockdown, which has the lowest economic costs also stands out as the best option that minimizes the number of deaths. Only 0.002 percent of the population dies in a well implemented full lockdown whereas the numbers range between 0.32 to 0.96 percent in the case of partial lockdown. In the model we do not quantify the economic costs of lost lives (see e.g., Greenstone and Nigam (2020)) under alternative lockdown scenarios. Had we incorporated the costs of deaths, the superiority of full lockdown would be even more striking.

4.1 The SIR Model for Pandemic

We start with introducing the model of the pandemic, which is the main workhorse in many epidemiological studies, see for example Allen (2017) among others. Let’s take a population of size $N$. At any given time, we can split the population into three classes of people: Susceptible ($S_t$), Infected ($I_t$) and Recovered ($R_t$) as of time $t$. The susceptible group does not yet have immunity to disease,
and the individuals in this group have the possibility of getting infected. The recovered group, on the other hand, consists of individuals who are immune to the disease.\footnote{Immunity can be developed either because the individual goes through the infection or because she gets vaccinated.} The Susceptible-Infected-Recovered (SIR) model builds on the simple principle that a fraction of the infected individuals in the population, $\frac{I_{t-1}}{N}$, can transmit the disease to susceptible ones $S_{t-1}$ with an (structural) infection rate of $\beta$. Therefore, the number of newly infected individuals in the current period is $\beta S_{t-1} \frac{I_{t-1}}{N}$. The newly infected individuals should be deducted from the susceptible individuals in the current period. Meanwhile, in each period, a fraction $\gamma$ of the infected people recovers from the disease, which in turn reduces the number of actively infected individuals.\footnote{See also Atkeson (2020), Bendavid and Bhattacharya (2020), Dewatripont et al. (2020), Fauci et al. (2020), Li et al. (2020), Linton et al. (2020), and Vogel (2020) on different mortality estimates.} To track any changes in the number of individuals in the above-mentioned three groups, the following set of equations is used:

\[
\Delta S_t = -\beta S_{t-1} \frac{I_{t-1}}{N} \\
\Delta R_t = \gamma I_{t-1} \\
\Delta I_t = \beta S_{t-1} \frac{I_{t-1}}{N} - \gamma I_{t-1}
\]

The law of motion for the number of infected individuals shows the trajectory of the pandemic at the aggregate level. Note that, $\Delta S_t + \Delta R_t + \Delta I_t = 0$ holds at any given time, assuming that the size of the population remains constant.

We modify the conventional SIR model to allow for sectoral heterogeneity in terms of the size and working conditions that can lead to distinct infection trajectories in each sector. The transmission of the virus requires close physical proximity. Hence, employees working in the industries with higher physical proximity are infected with a higher probability.\footnote{A report by DISK labor union in Turkey claims a three-fold increase in infection rates among workers: \url{http://disk.org.tr/2020/04/rate-of-covid-19-cases-among-workers-at-least-3-times-higher-than-average/}}

We assume that the economy is composed of $K$ sectors. We denote the industries by subscript $i = 1, \ldots, K$. Each industry has $L_i$ workers and there is also the non-working population which we denote by $N_{NW}$. Each industry has two types of workers: (i) employees who can perform their jobs remotely (i.e., teleworkable) and (ii) employees who need to be on-site to fulfill their jobs. In each industry, we denote the number of employees in the first group with $TW_i$ and the second group with
\( N_i \). Hence:

\[
L_i = TW_i + N_i. \tag{4}
\]

For the disease propagation, we lump the non-working population and the employees in the teleworkable jobs together, and call them the at-home group. We denote the at-home group with index \( i = 0 \). The total number of individuals in this group is:

\[
N_0 = N_{NW} + \sum_{i=1}^{K} TW_i. \tag{5}
\]

Suppose that the infection rate in the at-home group is \( \beta_0 \). In order to account for heterogeneous physical proximities across industries, we compute the rate of infection for each industry \( i \), denoted by \( \beta_i \), as:

\[
\beta_i = \beta_0 \text{Prox}_i \quad \text{for } i = 1, \ldots, K \tag{6}
\]

where Prox\(_i\) is the proximity index for industry \( i \).\footnote{We obtain the physical proximity values at the occupation level from O*NET dataset. O*NET collects the physical proximity information through surveys with following categories: (1) I don’t work near other people (beyond 100 ft.); (2) I work with others but not closely (e.g., private office); (3) Slightly close (e.g., shared office); (4) Moderately close (at arm’s length); (5) Very close (near touching). We divide the category values by 3 to make category (3) our benchmark. Specifically, a proximity value larger than 1 indicates a closer proximity than the ‘shared office’ level and smaller than 1 corresponds a less-dense working conditions. We create a single physical proximity value for each occupation by doing a weighted average of the normalized category values. We calculate the proximity values at the industry level after removing the teleworkable portion from the employees. We use \cite{Dingel_and_Neiman_2020}’s list of teleworkable occupations to capture the proportion of employment that can be fulfilled at remote locations in each industry.} It is plausible to think that the decline in demand during COVID-19 in a particular industry would lead to a decline in proximity (see \cite{Eichenbaum_et_al._2020}). Nevertheless, we do not incorporate this in our model and take the proximity rates as exogenous.

Here, \( S_{i,t} \), \( I_{i,t} \) and \( R_{i,t} \) denote the number of susceptible, infected and recovered individuals, respectively, with \( N_i = S_{i,t} + I_{i,t} + R_{i,t} \) denoting the total number of on-site individuals in industry \( i \) and the at-home group (\( i = 0 \)). Susceptible individuals in the at-home group can get infected from the infected individuals in the entire society:

\[
\Delta S_{0,t} = -\beta_0 S_{0,t-1} \frac{I_{t-1}}{N} \tag{7}
\]

where \( I_t = \sum_{i=1}^{K} I_{i,t} + I_{0,t} \) captures the total number of infected individuals. An on-site worker in
sector $i$, however, could be exposed to infection either at work, at the rate of $\beta_i S_{i,t-1} \frac{I_{i,t-1}}{N_i}$, or outside work, that involves all the remaining activities including family life, shopping and commuting at the rate $\beta_0 S_{i,t-1} \frac{I_{i,t-1}}{N}$. Hence, the number of susceptible individuals among the on-site workers in industry $i$ changes as:

$$\Delta S_{i,t} = -\beta_i S_{i,t-1} \frac{I_{i,t-1}}{N_i} - \beta_0 S_{i,t-1} \frac{I_{i,t-1}}{N}$$  \hspace{1cm} (8)$$

The recovery rate is the same for all types of infected individuals:

$$\Delta R_{i,t} = \gamma I_{i,t-1}$$  \hspace{1cm} (9)$$

The number of infected individuals changes as the susceptible individuals get infected and some infected individuals recover from the disease:

$$\Delta I_{i,t} = - \left( \Delta R_{i,t} + \Delta S_{i,t} \right)$$  \hspace{1cm} (10)$$

According to the initial report by the World Health Organization (WHO)\(^{26}\), the median recovery time for the mild cases is reported to be approximately 2 weeks. The mean recovery time could be longer when we include the severe cases. In this paper, we err on the optimistic side and set $\gamma = 1/14 \approx 0.07$ to establish a mean recovery time of 14 days. In the same report, the $R_0 \equiv \beta / \gamma$ of the disease, which captures the average number of individuals infected by a person carrying the disease, was estimated to be 2 to 2.5. Here, we use the lower end. In the absence of industrial heterogeneity, $R_0 = 2$ and $\gamma = 0.07$ implies $\beta = 0.14$. These values are in line with those used in Stock (2020) and Pindyck (2020) who primarily focus on calibration of the SIR model for tracking the evolution of the COVID-19 pandemic under different scenarios.

With industrial heterogeneity, we match the employment size weighted average $\beta_i$’s of the infected individuals to $\beta$. For an on-site worker in industry $i$, the implied $\beta$ parameter can be approximated by $(\beta_0 + \beta_i)$\(^{27}\). For a non-working individual, this parameter is only $\beta_0$. Using Equation (6),

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27. According to the DISK report cited in Footnote 24, the infection rate is 3 times higher for workers compared to the non-working population. Here, we take a moderate stance and set the rate to be 2 times higher on average for the workers.
we impose:

\[ \beta_0 N_0 + \sum_{i=1}^{K} (\beta_0 + \beta_i) \frac{N_i}{N} = \beta_0 + \beta_0 \sum_{i=1}^{K} \text{Prox}_i \frac{N_i}{N} = \beta \] (11)

Hence, we solve for \( \beta_0 \) in terms of \( \beta \), industry size, and the proximity levels as:

\[ \beta_0 = \beta \left( 1 + \sum_{i=1}^{K} \frac{\text{Prox}_i N_i}{N} \right)^{-1} \] (12)

with \( \beta = 0.14 \) based on the WHO report.

### 4.2 Production

We specify a simplified version of the production function where output is a linear function of labor. This treatment emphasizes the impact of the pandemic on production through changes in labor supply. Here, we implicitly assume that the amount of the capital stock remains the same in the short-run, and therefore, can be omitted during normal times as well as the pandemic period. We model production as a function of the number of workers in industry \( i \) as:

\[ Y_{i,t} = Z_i L_{i,t} \] (13)

where \( Z_i \) denotes the productivity of workers in sector \( i \).

During the pandemic period, the level of production decreases because the infected individuals cannot work until they recover from the disease. We have two groups of workers, at-home and on-site. Hence, the total number of available workers at time \( t \) will be:

\[ L_{i,t} = (N_{i,t} - I_{i,t}) + TW_i \left( 1 - \frac{I_{0,t}}{N_0} \right) \] (14)

where \( N_{i,t} \) is the number of on-site workers, \( I_{i,t} \) is the number of infected workers among on-site workers, and \( TW_i \) is the number of at-home workers (i.e. those who can work remotely) in industry \( i \). The ratio \( I_{0,t}/N_0 \) captures the fraction of individuals who are infected in the at-home group, which includes the non-working population as well as all at-home workers (i.e. teleworkers) in the

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28This is typical in global supply chain models as in Bonadio et al. (2020).
economy. Therefore, the production in industry $i$ changes to:

$$Y_{i,t}^S = Z_i L_{i,t}$$

(15)

### 4.3 Demand

During the pandemic period, consumer priorities and preferences change dramatically due to many reasons. First, there is fear of infection. In order to minimize the risks of getting infected, individuals alter their behavior and change their consumption patterns, such as refraining from public events or malls. The fear of infection is related to the number of infected individuals in the society. Second, there is fear of transmitting the disease to others. Individuals may choose to minimize their social interactions with a precautionary motive, in order to avoid infecting others inadvertently. The risk of transmission is particularly high for asymptomatic cases. Third, there is uncertainty about the duration of the pandemic and the related economic outlook. Aggregate expenditure typically declines during times of elevated uncertainty. Fourth, there is a direct income effect. Individuals lose their income stream either when they get laid-off or when they experience a sharp decline in demand for their output.

In order to capture the change in demand patterns during the pandemic, we consider two demand profiles, one corresponding to normal times and the other one corresponding to the brunt of the pandemic. We determine the demand for each industry during normal times from the consumption data in national accounts. As for the COVID-19 period, we estimate changes in the expenditure levels during the pandemic using credit card spending data. As for the sectors where we do not have the credit card data, we use industry reports and expert opinions.\(^\text{29}\) The progression of the pandemic and the normalization of demand as the pandemic fades is a gradual process. In order to capture this steady adjustment, we assume that the individuals move between these two profiles smoothly, as a function of the number of infected individuals in the country. After determining demand, we use the input-output framework and map the final good consumption back to output in each industry.

\(^{29}\)Expected final demand changes and the resources we use in this estimation are presented in Table A.3 of the Appendix.
In modeling the demand side, we first express the utility function of a representative agent who maximizes her utility by optimally allocating her income on the expenditure of different final goods. Following the literature on input-output analysis (see e.g. Acemoglu et al. (2012), among others), we assume that the representative agent has a Cobb-Douglass utility function:

$$U(e_1, \ldots, e_n) = \prod_{i=1}^{n} e_i^{\alpha_i}, \quad (16)$$

with $e_i$ denoting the level of expenditure in industry $i$, and $\alpha_i$ representing the share of industry $i$ in total expenditure with $\sum_{i=1}^{n} \alpha_i = 1$ and $0 < \alpha_i < 1$ for all $i = 1, \ldots, n$. The utility function in (16) incorporates a budget restriction which implies that the total income ($w$) equals total expenditure, i.e., $w = \sum_{i=1}^{n} e_i$. With the Cobb-Douglass utility function, $\alpha_i$ determines the share of industry $i$ in the expenditure so that $e_i = \alpha_i w$ for $i = 1, \ldots, n$.

During times of the pandemic, demand patterns change. For the sake of simplicity, we assume that changes in demand come from two channels. First, the pandemic changes preferences and priorities, which implies an adjustment in sectoral weights. The utility function transforms into:

$$U(e_1, \ldots, e_n, I) = \prod_{i=1}^{n} e_i^{\tilde{\alpha}_i(I)}, \quad (17)$$

with the Cobb-Douglass exponents depending on the number of infections and $\tilde{\alpha}_i(I) = \alpha_i$ for a small number of infections, i.e., $I \leq 0.1 \bar{I}$, where $\bar{I}$ is a scaling parameter for infections. In the Turkish context, we set $\bar{I} = 50,000$ to capture a relevant range for the number of of infections (see below for our simulations). This limit implies that the utility function returns to normal times if the number of infections remain below 5,000. For large $I$, the limit level is defined as $\lim_{I \to \infty} \tilde{\alpha}_i(I) \equiv \bar{\alpha}_i$ with $\sum_{i=1}^{n} \bar{\alpha}_i = 1$ and $0 < \bar{\alpha}_i < 1$ for all $i = 1, \ldots, n$.

In addition to the changes in preferences during the pandemic, demand also changes due to the income effect. We assume that the available income for expenditure decreases by a ratio of $1 - \eta(I)$ compared to normal times. We assume that $\eta(I)$ is a decreasing function of the number of infections and satisfies $\eta(I) = 1$ for $I \leq 0.1 \bar{I}$. For large $I$, i.e., $\lim_{I \to \infty} \eta(I) = \bar{\eta}$ with $0 < \bar{\eta} \leq 1$. In this set up, the minimum level of income that is necessary for survival at the brunt of the pandemic is given by $\bar{\eta} \times w$, which can be achieved through transfer payments. While we capture the effects of the
pandemic by modelling the demand parameters $\alpha$ and $\eta$ as a function of the number of infections, the specification can be generalized to include consumer sentiment or the trustworthiness of the policies as the determinants of these key demand parameters. Hence, the impact of a decline in capital inflows, or a decline in policy credibility during the pandemic can be analyzed by adjusting the demand parameters within our framework.

To determine the level of output implied by the changes in demand during the pandemic, we first express the expenditure in each industry as a function of the number of infections. Next, we construct a ratio, $\delta_i(I)$, which shows the expenditure in industry $i$ when the infection level is $I$ relative to the expenditure during normal times. The numerator in this ratio is dependent on both the income channel and changes in priorities. By combining both channels, we can write $\delta_i(I)$ as:

$$\delta_i(I) = \frac{\tilde{\alpha}_i(I)\eta(I)}{\alpha_i}.$$  \hspace{1cm} (18)

As the demand ratio approaches 1, it signals that the number of infections decline and demand normalizes. As the demand ratio approaches 0, it reflects that the number of infections increase and demand shrinks due to the pandemic. Using this ratio, we write the limiting cases for $\delta_i(I)$. For small $I$ (i.e., $I \leq 0.1\bar{I}$), $\delta_i(I) = 1$. Thus, for a small number of infections, demand remains intact such that the ratio of demand during normal times equals demand during the pandemic.\footnote{We use the number of infected patients in Turkey as the determinant of global demand change in a particular industry. While it is possible for a given country to handle the pandemic better or worse than Turkey, we argue that the number of infections and the consequent demand shifts in Turkey would not be that far off from the global course of the pandemic. In fact, a recent report by the ECB (Chapter 3, ECB Economic Bulletin, Issue 3 / 2020) highlights the striking resemblance between the timing of demand collapse in Turkey and the Euro Area: (https://www.ecb.europa.eu/pub/pdf/ecbu/eb202003.en.pdf) on Economic Activity.}

For large $I$, which corresponds to the peak of the pandemic, $\lim_{I \to \infty} \tilde{\alpha}_i(I) = \bar{\delta}_i = \frac{\alpha_i\eta}{\alpha_i}$. If the demand for an industry $i$ completely collapses during the pandemic (e.g., the airline industry), then $\bar{\delta}_i = 0$. If there is no change in demand during the pandemic (e.g., food industry), then, $\bar{\delta}_i = 1$. We assume that $\bar{\delta}_i$ is the utmost demand change in a particular sector that is globally valid under a fully developing pandemic.

Changes in demand at any given time is a function of the number of infected individuals in the population. In this framework, we assume that the ratio of demand, $\delta_i(I)$, smoothly fluctuates between 1 when nobody is infected and $\bar{\delta}_i$ when a very large number individuals get infected using
the functional form \(^{31}\) as:

\[
\delta_i(I) = \begin{cases} 
1 & \text{if } I \leq 0.1 \bar{I} \\
\delta_i \frac{1 + (I/\bar{I} - 0.1)}{\delta_i + (I/\bar{I} - 0.1)} & \text{if } I > 0.1 \bar{I}
\end{cases}
\] (20)

It is important to note that the overwhelming uncertainty about the course of the virus may suppress economic confidence for a longer period of time. To the extent that the actual normalization is slower than what is implied by Equation (20), we err on the conservative side by assuming a faster recovery.

Given the smooth transition function, we now model the changes in the final demand levels using \(\delta\) values. Let’s illustrate the final demand of country \(c\) in industry \(i\) with \(F_{c,i}\). Accordingly, the new level of final demand in industry \(i\) in country \(c\) during the pandemic becomes:

\[
\tilde{F}_{c,i}(I) = F_{c,i} \delta_i(I)
\] (21)

where \(\tilde{F}_{c,i}(I)\) represents the revised demand during the pandemic when the number of infections is \(I\).

In order to map the changes in the final demand for each sector to the output level in each industry, we use the input-output framework. Using a closed-economy version of the input-output relations would neglect the impact of foreign trade on aggregate expenditure. Turkey is an open economy with a trade-to-GDP ratio of almost 63% as of 2019 (See Figure 7).

In order to account for the international linkages to fully capture the impact of final demand on production, we utilize OECD Inter-Country Input-Output (ICIO) Tables\(^{32}\). ICIO provides us with input usages of industry \(i\) in country \(c\) from any industry in any country. As shown in Figure A.1.

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\(^{31}\)This inverse hyperbolic functional form provides a smooth transition between the two limiting cases, for small and large \(I\), where the marginal impact of the number of infections changes at a rate that is inversely proportional to the number of infections. The specification flexibly allows for changes across sectors as \(\bar{I}\) and \(\delta_i\) are the tuning parameters that determine the limits and the speed of the convergence.

The smooth transition in Equation (20) between the limiting cases can be achieved with the following functional forms for \(\eta(I)\) and \(\bar{\alpha}_i(I)\) for \(i = 1, \ldots, n\):

\[
\eta(I) = \begin{cases} 
1 & \text{if } I \leq 0.1 \bar{I} \\
\frac{1 + (I/\bar{I} - 0.1)}{\eta + (I/\bar{I} - 0.1)} & \text{if } I > 0.1 \bar{I}
\end{cases}
\]

\[
\bar{\alpha}_i(I) = \begin{cases} 
\alpha_i & \text{if } I \leq 0.1 \bar{I} \\
\frac{\bar{\alpha}_i \eta + (I/\bar{I} - 0.1)}{\bar{\alpha}_i \delta_i + (I/\bar{I} - 0.1)} & \text{if } I > 0.1 \bar{I}
\end{cases}
\] (19)

\(^{32}\)https://www.oecd.org/sti/ind/inter-country-input-output-tables.htm
we have 36 industries and 69 entities (corresponding to 65 countries), giving us a matrix of $2484 \times 2484$ entries. The final demand vector has 2484 entries and we index the entries of this vector by $j$, corresponding to each country-industry combination. By dividing the rows of ICIO matrix with the total output of industry $j$, we obtain the direct requirements matrix $A$. This matrix summarizes the usage of each intermediate input to generate $1$ worth of output. Output of each industry is either used as an intermediate input or consumed as final demand. Using matrix notation, we decompose the total output into intermediate and final usage as:

$$Y = AY + F$$  \hspace{1cm} (22)$$

Here, $Y$ denotes the output vector and $F$ denotes the final demand vector whose entries are $Y_j$ and $F_j$ respectively, while $j$ is an index over all $(c, i)$ combinations. Therefore, we can solve for the output

\[\text{with a slight abuse of the notation, we drop the subscript to refer to vectors or matrices of the variables. For simplicity, we often use } (c, i) \text{ to refer to an entry instead of } j \text{ when we refer to vector entries.}\]
to satisfy the final demand as:

\[ Y = (I - A)^{-1} F \]  \hspace{1cm} (23)

From this equation, we write the total output of country \( c \) as:

\[ Y_c = \sum_{i=1}^{n} Y_{c,i} \]  \hspace{1cm} (24)

Using the demand change from Equation (21) during the infection, the demand channel changes the output as:

\[ Y^D_t = (I - A)^{-1} \tilde{F}(I_t). \]  \hspace{1cm} (25)

where \( Y^D_t \) represents the output and \( \tilde{F}(I_t) \) represents the vector of demand at time \( t \) as a function of the number of infections, \( I_t \). Therefore, the output also changes with the dynamics of the pandemic.

### 4.4 Equilibrium

In equilibrium, production declines by the largest magnitude that is implied by either supply or demand side. In other words, during the pandemic, we expect the output vector to be:

\[ Y^{EQ}_t = \min(Y^S_t, Y^D_t) \]  \hspace{1cm} (26)

where \( \min \) represents element by element minimum function for two vectors, namely \( Y^S_t \) and \( Y^D_t \).

The value-added of the output in industry \( i \) in country \( c \) is calculated from the shares of value added in each industry during normal times as:

\[ VA^{EQ}_{t,c,i} = Y^{EQ}_{t,c,i} \frac{VA_{c,i}}{Y_{c,i}} \]  \hspace{1cm} (27)

Therefore, GDP of the country \( c \) at time \( t \) can be obtained through:

\[ GDP^{EQ}_{t,c} = \sum_{i=1}^{n} VA^{EQ}_{t,c,i} \]  \hspace{1cm} (28)
4.5 Data

In our analysis, we use OECD ICIO Tables for 2015. As industrial classification, OECD uses an aggregation of 2-digit ISIC Rev 4 codes to 36 sectors. Throughout our analysis, we will make use of this classification labeled as OECD ISIC Codes.

To calculate the industry level teleworkable share and the physical proximity measures, we use the occupational composition of the industries. We use the list provided by Dingel and Neiman (2020) for the occupations which can fulfill their tasks remotely. For the workers that continue to do their jobs on-site, we assume that the infection rate depends on the physical proximity that is required in their workplace. To calculate the proximity requirements for the occupations, we use the self-reported Physical Proximity values available in the Work Context section of the O*NET database. We divide the O*NET categories into 3 to have values larger than 1 if the reported category for the physical proximity is 3 (Slightly close (e.g., shared office)) or higher. We create a single proximity value for each occupation by weighting the normalized score with the percentage of answers in each category. To obtain industry-level teleworkable share and proximity values, we calculate the weighted average of the values corresponding to the occupations in each industry using the Occupational Employment Statistics (OES) provided by the U.S. Bureau of Labor Statistics (BLS). OES data follows four-digit NAICS codes to classify industries. In order to convert proximity data to OECD ISIC codes, we make use of the correspondence table between 2017 NAICS and ISIC Revision 4 Industry Codes, provided by the U.S. Census Bureau. We provide the teleworkable share and the proximity index for the industries in Table A.2 of the Appendix.

We obtain employment data from the Turkish Social Security (SGK) Agency. SGK follows four-digit NACE Revision 2 codes to classify industries. In order to aggregate employment data to 36 OECD ISIC codes, we make use of the Eurostat correspondence table between NACE Revision 2 and ISIC Revision 4 Industry Codes. SGK lacks the data on the number of employees working in the “Public Administration Sector,” so we fill this information using the relevant data provided by the President’s office.

We use publicly available credit card spending data from the CBRT to calculate the estimated demand changes during the pandemic in each industry. The list of OECD ISIC industries, and the expected changes are listed in Table A.3 of the Appendix along with explanations. In Table A.5 of the Appendix, we provide the matching we used with CBRT spending data and OECD ISIC industries. The data on credit card spending is not available for the full set of sectors. In this case, we use projections based on sectoral reports, experiences of other countries and historical data on the specific sector as well as the whole manufacturing sector. While the aggregate demand shock is computed as 23% when we focus only on the sectors with credit card spending data, it is 16% when we consider the full set of sectors. Therefore, our sensitivity analysis indicate little or no change in our findings qualitatively.

Under full lockdown, only a few industries are active. We use the decree issued by Turkish Ministry of Interior on April 10, 2020 to identify these industries. Turkish full lockdowns are typically on weekends and holidays and thus the list did not include some critical sectors. We supplemented the list with the food sector as well as household and sanitary goods. The list of these sectors is given in Table A.4 of the Appendix. From these industries and using the employment data at 4 digits, we calculated the share of each OECD ISIC industry that would remain active during the lockdown. Finally, we calculated the share of public employees that are not affected by the lockdown using the publicly available information, which is listed in Table A.6 of the Appendix.

5 Quantitative Analysis

5.1 Infection Rates under Alternative Lockdown Scenarios

In this section, we illustrate the consequences of alternative lockdown scenarios within our framework. In these scenarios, we impose changes on $\beta_0$ (i.e., the infection rate of the non-working population) and possibly on $\beta_i$ for (i.e., the infection rate of the working population in industry $i$) and simulate the course of the pandemic. The decline in $\beta$ reflects the effectiveness of a particular lockdown scenario which depends on country characteristics such as demographic dynamics, whether or nor there is a more authoritarian culture with less resistant public, the influence of the scientific committees in shaping political decisions, or the ability of a trustworthy and independent media in
affecting public sentiment. The effectiveness of the lockdown also depends on the recovery rate that depends on the quality of healthcare services as well as ICU capacity.

We assume that the pandemic is successfully contained if the number of total infections declines to 5000 after observing the peak. These simulations allow us to calculate the economic costs of alternative lockdown scenarios.  

We start with the no lockdown scenario and compare it to partial lockdown where certain restrictions are imposed on daily life to incorporate social distancing rules while businesses remain open. This implies that under partial lockdown $\beta_0$ is diminished compared to the case where no action is taken, but $\beta_i$ for $i = 1, \ldots, n$ remain unchanged. We consider three cases of partial lockdown where the infection rate, $\beta_0$ is reduced by the proportion of 0.5, 0.25 and 0.10 compared to the reference setting. Figure displays the evolution of the number of infected patients under these four scenarios when a hypothetical lockdown is implemented for 240 days, starting early on the 10th day and remains active until the 250th day.

Figure 8: No lockdown versus Partial Lockdown Scenarios

As can be seen from the figure, in case no action is taken against the COVID-19 pandemic, which is shown with the blue line, the pandemic advances at a rate implied by the benchmark reproduction

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35We note that the 5000 threshold that is assigned for the containment of the pandemic differs from the notion of Critical Community Size (CCS) [Bartlett 1960]. CCS is the threshold for the number of susceptible individuals to die out by itself. Instead, the 5000 threshold that we set in the model represents the number of infectious individuals who can be feasibly tested, traced, and eventually quarantined so that the pandemic can be contained successfully. We assume that for each infected individual, we need to test ten additional people on average. Thus, if there are 5000 patients, tracing the infection requires about 50,000 tests, which is close to the current testing capacity in Turkey.
rate of $R_0 = 2$. This implies that the pandemic reaches its peak around the 150th day with a total toll of around 14 million infections. Following this state of “herd immunity”, the number of infections starts to decline. After approximately 300 days, the virus is taken under control. Under the no lockdown scenario, 1.13 percent of the population dies if we assume a 1.5 percent mortality rate. The GDP declines 11.0% in this case. We should remind the readers that the economic costs that are expressed in terms of GDP should not be misinterpreted as annual growth forecasts. We merely express the cost of the lockdown in terms of the GDP.

Under partial lockdown scenarios, the reproduction number declines below 2 due to lower infection rates but remains above 1 in all three scenarios. Specifically, we assume that the lower infection rate dampens the rate at which the pandemic evolves, nevertheless it is not sufficient to contain it altogether. This is due to the fact that businesses remain open, which feeds the virus within the industries and affects the overall course of the pandemic. If the infection rate is relatively high ($0.5 \times \beta_0$), which is shown with the red line, the GDP declines 11.6 percent. If the infection rate is moderate ($0.25 \times \beta_0$), shown with the green line, the GDP declines by 10.9 percent. If the infection rate is relatively low ($0.1 \times \beta_0$), shown with the black line, the GDP declines by 10.5 percent.

None of the 240-day partial lockdown scenarios that we considered in Figure 8 were successful in containing the pandemic. When the lockdown is removed on day 250, all three partial lockdown scenarios have approximately the same number of infections. Once the lockdown is removed, however, the virus follows a different course in each scenario. For the low infection rate scenarios (green and black lines) the number of new cases increase rapidly, leading to peak levels within 50 days after the lockdown. Meanwhile the high infection rate and no lockdown scenarios show a steady decline (the blue and red lines). This is because less people get infected during partial lockdown (and get immunity) under the low infection rate scenarios, shown by the area under the black and green lines. Hence, by the time the lockdown is removed, the number of susceptible people are significantly higher under the low infection rate scenarios, increasing the effective $R_0$ ($= \beta / \gamma$). Thus, in the absence of an efficient drug or vaccination, a partial lockdown may need to continue indefinitely, until the number of cases decline to 5000. Figure 9 shows the simulation results if partial lockdown lasts for a full year. As in Figure 8 we assume that the industries are operating as usual and thus $\beta_i$'s (for $i = 1, \ldots, K$) remain unaffected.
Figure 9: Alternative Scenarios under Partial Lockdown for Full Year

Compared to Figure 8, we observe that the main advantage of an extended partial lockdown is that it flattens the curve by spreading the number of infections over time and allowing for a larger recovery rate. In terms of the economic costs, the additional economic costs of the longer partial lockdown hover around 0.5 percent of the GDP. The added costs despite the extended duration of the lockdown are limited. This is due to the fact that the decline in demand already reaches a maximum level at the earlier stages of the lockdown and successive reductions in production only reflect the decline in supply due to increased number of infections.

Figure 10 illustrates the implications of our model under full lockdown. If the lockdown is put into practice when the number of infections is around 80,000, a fully effective procedure lowers the reproduction rate to zero \( R_0 = 0 \), which is shown by the blue line, and contains the pandemic within 39 days (the gray shaded area). The consequent decline in GDP is about 5.8 percent. If the lockdown is not very effective and the infection continues to spread with some minimal reproduction number \( R_0 = 0.02 \), then the duration of the lockdown increases by 15 days (yellow shaded area) to 54 days and the GDP declines by 7.6 percent.

The costs of delaying full lockdown are shown in Figure 11. The benchmark scenario that is illustrated in Figure 10 is shown with the blue line. If the lockdown is delayed by only one day, the number of infections increases by more than 10,000. In the model, we assume that the number of infections increases faster than the official statistics, which report only the tested patients. Under
these circumstances, a 39-day lockdown is no longer sufficient to control the pandemic. Thus, in exchange for a one-day delay, the lockdown needs to be extended by two more days (the red line), which increases the costs of the lockdown to 5.9 percent of the GDP. If there is a two-day delay (the green line), this time the duration of the lockdown increases to 43 days and the decline in GDP is 6.2 percent. If the lockdown is delayed by one week (the black line), the decline in GDP is 7.3 percent. After 100 days, the virus starts to spread again and hence prematurely ending the lockdown is rather ineffective.

As we compare the economic costs under full lockdown (Figures 10 and 11) with those of partial...
lockdown (Figures 8 and 9), we note that the costs of full lockdown are lower than any of partial lockdown scenarios.

As we compare the number of deaths under alternative scenarios, we observe that 0.001 percent of the population dies under an effective full lockdown, compared to 1 percent of the population under no lockdown and about 0.8 percent of the population under partial lockdown scenarios that last for 250 days. If partial lockdown is extended to a full year, then the number of deaths decline to about 0.5 percent of the population.

5.2 The Role of External Demand Shocks

The aggregate costs of COVID-19 shock that we calculated in the previous section embeds supply and demand channels in Turkey as well as abroad. In this section, we illustrate the role of external demand and supply in total costs. In order to better illustrate the role of international linkages for the Turkish economy, we consider two alternative scenarios.

The final demand of country $c$ in industry $i$ is met by domestic production and imports of final goods. Formally, we write the final demand as:

$$F_{c,i} = \sum_{c'} F_{c,c',i}$$

(29)

where $F_{c,c',i}$ denotes goods or services produced by industry $i$ in country $c'$ and consumed in country $c$. Following Equation 21, we write the final demand in country $c$ in industry $i$ at the peak of the pandemic as:

$$\bar{F}_{c,i} = F_{c,i}\delta_{c,i}$$

(30)

Different than Equation 21, in this section we allow for country specific demand shocks. The corresponding output to satisfy this final demand level is obtained by:

$$\bar{Y} = (I - A)^{-1}\bar{F}$$

(31)
From this equation, we write the total output of country $c$ as:

$$\bar{Y}_c = \sum_{i=1}^{n} \bar{Y}_{c,i}$$  \hspace{1cm} (32)$$

The value-added portion of the output is calculated from the shares of value added in each industry during normal times. The total value-added (GDP) in country $c$ can thus be written as:

$$\bar{GDP}_c = \sum_{i=1}^{n} \bar{Y}_{c,i} \frac{VA_{c,i}}{Y_{c,i}}.$$  \hspace{1cm} (33)$$

The matrix for intermediate goods is obtained from the direct requirements matrix and the output vector:

$$INT = A\bar{Y}.$$  \hspace{1cm} (34)$$

Each entry of the matrix $INT$ corresponds to the usage of intermediate goods by industry $i$ in country $c$ from industry $i'$ in country $c'$. Combining imports of intermediate goods and final goods, we write the total imports for country $c$ as:

$$\text{imports}_c = \sum_{c' \neq c}^{n} \sum_{i=1}^{n} \left( \bar{F}_{c,c',i} + \sum_{i'=1}^{n} \sum_{i''=1}^{n} INT_{c,i,i''} \right).$$  \hspace{1cm} (35)$$

Similarly the total exports by country $c$ is:

$$\text{exports}_c = \sum_{c' \neq c}^{n} \sum_{i=1}^{n} \left( \bar{F}_{c',c,i} + \sum_{i'=1}^{n} \sum_{i''=1}^{n} INT_{c',i',i''} \right).$$  \hspace{1cm} (36)$$

As a result, a decline in foreign demand for final goods will create sectoral output declines in many domestic sectors, which will add to aggregate output decline in Turkey. To highlight this mechanism, we present three scenarios.

Scenario 1 assumes the same proportionate demand shock in Equation 30 for the whole world. For example, if we estimate that the demand for automobiles decline by 60 percent based on Turkish data, we assume that the demand for automobiles declines by 60 percent throughout the world. Figure 12 shows how much total output, exports and imports change at the brunt of the pandemic relative to normal times for alternative scenarios. In the baseline scenario, the decline in terms of
total output is 19.8 percent (Scenario 1 in Figure 12). Interestingly, imports decline less (17.9 percent) compared to exports (23.4 percent). This is consistent with the nature of the Turkish economy which is highly dependent on imports of intermediate goods. On the exports side, a further breakdown indicates that the 27.4% decline in terms of final goods is higher than the 18.8% decline in intermediate goods (not shown). Similarly, on the imports side, the 19.7% decline in intermediate goods is higher than the 16.1% decline in final goods (not shown).

Figure 12: Demand Shocks for an Open Economy with I-O Links

NOTES: This graph illustrates the impact of three different scenarios for demand shocks. In the first scenario, all the countries are assumed to experience the same demand shifter during the pandemic. In the second scenario, only Turkey experiences a demand shock but the international demand levels are intact. In the final scenario, the international demand levels are down but the demand in Turkey is at pre-pandemic levels. The number written on each bar corresponds to the percentage change in the relevant variable in the underlying scenario relative to its pre-pandemic level.

Under scenario 2, we assume that the demand in Turkey declines but the international demand for final goods is back to its normal (see Scenario 2 in Figure 12). Using the automobile example above, this implies that the domestic demand for automobiles shrinks to 60% of normal levels but
the international demand remains at its normal levels. In this setting, the decline in terms of total output is 14.6 percent at the brunt of the pandemic. The decline in imports is 14.7% but the decline in exports is only 0.1%.

Lastly, in scenario 3, we model the setting where the demand in Turkey is intact but the demand in international markets has plummeted (see Scenario 3 in Figure 12). Under this scenario, the decline in output is 5.2% solely because of international linkages. As expected, the exports are hit the hardest with a decline of 23.3% and imports decline by 3.2%.

If we compare Scenario 1 and Scenario 3, we can see the role of demand in total economic costs. The decline in foreign demand solely account for almost 27 percent of the decline in aggregate output. Notice that we run these scenarios under no lockdown policy in the absence of any policy action.36

5.3 Sectoral Breakdown of Economic Costs

In this section, we analyze the economic costs at the sectoral level. Heterogeneity in sectoral costs may stem from several channels. Sectors that are closed down due to isolation measures (i.e. nonessential sectors), those that are hit hardest by the collapse in demand such as the services sectors, or those industries where teleworking is not very feasible will be hit harder. As for the role of international linkages, those sectors with greater exposure to international spillovers, particularly with those countries that had larger domestic outbreaks would be more affected. Similarly, those industries that rely more on external finance would experience the pinch of tightening in global financial conditions. In the next sub-section, we focus on the role of trade linkages. In the following sub-section, we calculate the sectoral economic costs in our framework under different scenarios. Using these sectoral costs, in the last sub-section, we disentangle the role of trade and external funding in sectoral costs in a regression framework.

36The work by Bonadio et al. (2020) shows that a third of the output loss is due to global supply chains. Different from our framework where we have both supply and demand shocks, Bonadio et al. (2020) only assume sectoral supply shocks. These shocks are caused by the fraction of workers not being able to work because of the measures adapted by the country. Despite these differences, however, our predicted declines in GDP are similar because their supply shock, which affects all the workers in the economy, is large. The supply shock is assumed to be caused by the workers not showing up for work and is captured by the severity index of the lockdown by the Oxford Blavatnik School of Government Coronavirus Government Response Tracker [https://www.bsg.ox.ac.uk/research/research-projects/coronavirus-government-response-tracker]. The severity index does not necessarily reflect the proportion of workers not able to work, however.
5.3.1 The Role of Trade Linkages in Sectoral Costs

In this section, we investigate the role of international linkages in determining sectoral costs. International linkages would affect economic costs through trade relationships as well as capital inflows. Those sectors that are more closely connected to international value chains as well as those sectors that are dependent on external funding would be affected more from the global developments.

Figure 13 allows us to get a glimpse of the role of trade from the international input-output tables. The figure illustrates the share of imports in total intermediate inputs (the left panel) and the share of exports in total output (the right panel). The left panel shows the potential role of supply chains in COVID-19 crisis. While we do not explicitly incorporate these supply shocks into our model, they are implicitly captured through changes in final demand. The left panel indicates that sectors such as paper products, computers and electronics, electrical equipment, rubber and plastic, coke and refined petroleum, as well as pharmaceuticals rely more on intermediate inputs. Thus, we would expect these sectors to be more severely affected from the pandemic, due to disruptions in supply chains.

Turning to the right panel, motor vehicles, transportation equipment, electrical equipment, computer and electronics and tourism relates services sectors such as accommodation and food services are the sectors that rely more on external demand. Thus, the deep recessions that are expected in Turkey’s major export markets such as the Euro Area, UK, or the US would hit these sectors the most, consistent with our analysis in Figure 12.
5.3.2 Sectoral Breakdown of Economic Costs under Different Scenarios

In the previous sub section, we highlighted the role of international linkages in sectoral costs. Armed with this intuition, we now illustrate how the economic costs related to the COVID-19 shock differ across industries under the alternative lockdown scenarios. Figure 14 shows how hard each sector is hit from the pandemic under alternative lockdown scenarios. Consistent with our earlier findings, we observe that the full lockdown has the lowest economic costs compared to the alternatives. In terms of sectoral heterogeneity, we note that teleworkable or essential sectors are less severely affected because they continue functioning for all lockdown scenarios (such as education, IT, public administration). Meanwhile, non-essential sectors or those that require on-site work are more severely affected (such as accommodation and food services, arts, entertainment, and recreation, construction).

After documenting the heterogenous economic costs of the pandemic for different sectors, we investigate whether these costs are accrued from demand or supply pressures. Figure 15 counts the days in which output implied by the demand channel or supply channel prevails to bring about the equilibrium output in a given industry.

To interpret the findings present in this figure, we consider three benchmark scenarios: Panel (a)
Figure 14: Sectoral Heterogeneity in terms of Economic Cost of COVID-19 Shock

(a) **Scenario 1:** No Lockdown, $\beta = 0.14$

(b) **Scenario 2:** Full Lockdown, $91-131, R_0 = 0$

(c) **Scenario 3:** Partial Lockdown, $10-250, 0.25 \times \beta_0$

**NOTES:** This figure shows how the economic cost of COVID-19 shock differs across sectors in a particular lockdown scenario. The panels show three alternative scenarios: (a) No action is taken against the COVID-19 pandemic; (b) A lockdown is put into practice between the 91st and 131st days of the pandemic and is fully effective with zero reproduction number; (c) A partial lockdown is put into practice between 10th-250th days of the pandemic that evolves with a moderate infection rate ($0.25 \times \beta_0$). For each scenario, we measure the sector-level economic cost as the percentage change in overall economic activity (proxied by value added) for a given sector during pandemic relative to its pre-pandemic level. Economic costs are aggregated from the 2-digit OECD ISIC codes to the 1-digit NACE code using 2-digit sector value added values that we obtain from the OECD ICIO Tables. NACE 1-digit sectors are A, B C, D&E, F, G, H, I, J, L, M&N, P, Q, R&S. In each panel, the sectors are ranked in a descending order according to the magnitude of economic cost under the corresponding scenario.

Panel (a) suggests that under the no lockdown scenario, the demand channel, shown by the red bars, drives output in almost all days until the virus is fully contained. The supply channel, presented by the blue bars, prevails only in the early days of the pandemic (not shown). Among the 15 industry groups, “Accommodation and food services,” “Arts, entertainment, recreation and other service activities,” and “Real estate activities” are those that result in the highest economic costs of 36%, 33%, and 20% of the value added generated in those sectors, respectively. This is not only because goods produced in those categories (which are all provided by the services sector) cannot be consumed from home, but also because people prefer delaying their consumption until the uncertainty regarding the containment of the pandemic resolves. Furthermore, another aspect of sectoral heterogeneity is clearly seen under no lockdown scenario such that the demand channel prevails longer in those sectors. This is because households are more likely to cut back on their expenditure on the goods produced by those non-essential sectors following the COVID-19 shock.

Under full lockdown scenario, the supply channel drives output due to the closure of all non-
(a) **Scenario 1**: No Lockdown, \( \beta = 0.14 \)

(b) **Scenario 2**: Full Lockdown, 91-131, \( R_0 = 0 \)

(c) **Scenario 3**: Partial Lockdown, 10-250, \( 0.25 \times \beta_0 \)

**Figure 15: Supply and Demand Pressures under Benchmark Lockdown Scenarios**

Notes: In this figure, each bar shows the number of days in which the supply channel (shown by the blue bars) or the demand channel (shown by the red bars) prevails to bring the economy into equilibrium in a given industry. The panels show three alternative scenarios: (a) No action is taken against the COVID-19 pandemic; (b) A lockdown is put into practice between the 91st and 131st days of the pandemic and is fully effective with zero reproduction number; (c) A partial lockdown is put into practice between 10th and 250th days of the pandemic that evolves with a moderate infection rate \( (0.25 \times \beta_0) \). For each scenario, we measure the sector-level economic cost as the percentage change in overall economic activity (proxied by value added) for a given sector during pandemic relative to its pre-pandemic level. Economic costs are aggregated from the 2-digit OECD ISIC codes to the 1-digit NACE code using 2-digit sector value added values that we obtain from the OECD ICIO Tables. NACE 1-digit sectors are A, B C, D&E, F, G, H, I, J, L, M&N, P, Q, R&S. In each panel, the sectors are ranked in a descending order according to the magnitude of economic cost under the corresponding scenario.

Essential industries, whereas the demand channel prevails approximately 30 days before the restrictions are implemented (Panel (b)). Among the 15 industry groups, “Accommodation and food services,” “Construction” and “Mining and non-quarrying of non-energy producing products” are those that result in the highest economic costs of 12%, 9.5%, and 9.1% of the valued added generated in those sectors, respectively. Different from the no lockdown scenario, sectoral heterogeneity is not highly pronounced in terms of supply and demand pressures under this scenario. To be specific, after the restrictions are implemented the supply channel dominates for all the sectors excluding “Human health & social work,” and “Public administration.”

Panel (c) shows that under partial lockdown that is put into practice between 10th-250th days of the pandemic and evolves with a moderate infection rate \( (0.25 \times \beta_0) \), the supply channel dominates in the first 100 days of pandemic. On the other hand, demand drives output for the rest of the year,
including the days in which new peak levels are reached after the partial lockdown is prematurely removed. This is because of the fact that businesses remain open, which feeds the virus within the industries and increases the uncertainty about the containment of the pandemic. Among the 15 industry groups, “Accommodation and food services,” “Arts, entertainment, recreation and other service activities,” and “Real estate activities” are those that result in highest economic costs of 36%, 34%, and 21% of the value added generated in those sectors, respectively. We note that sectoral heterogeneity in terms of supply and demand pressures is very similar to the no lockdown scenario.

5.3.3 The Role of External Finance in Sectoral Cost

In the previous sub-section, we calculated the economic costs of COVID-19 for each sector. There is heterogeneity among these sectors regarding their external finance needs. Although we have shown that the role of foreign demand is substantial in those sectoral estimates, we want to also identify the role of external finance needs in the calculated economic costs.

To do this, we consider a regression specification at the sector-level. Specifically, we regress the economic cost in each sector onto its I-O link as well as the capital needs in that sector. We proxy for the sector-level external finance needs by generating an interaction term that captures the degree to which each sector is open to capital flows. This sector-level proxy is computed as a weighted sum of net I-O trade of country-sector pairs where the weights are country specific capital flows divided by the number of countries.

A priori, we expect tradable sectors to be hit harder from the COVID-19 shock because they are exposed to adverse foreign demand shocks in addition to domestic shocks. However, via I-O links, even a non-tradeable sector can get hit hard and hence we capture both of these channels via our I-O trade variable. Furthermore, the more a particular sector relies on external borrowing, the larger should be the economic costs during COVID-19 crisis due to increased risk aversion.

\[ I-O \text{ Trade Finance}_{i} = \sum_{c=1}^{n} \left( \frac{((\text{Exports}_{c,j} - \text{Imports}_{c,j})/\text{Output}_{i}) \times \text{Capital Flows}_{c}}{n} \right) \]

The related data on capital flows is obtained from BIS and it is publicly available at https://stats.bis.org/statx/srs/table/A6.27c=TR&p=20194&m= Capital flows data of Turkey from 26 countries refers to data on Turkish banking sector external liabilities vis-a-vis those countries for 2019-Q4. We normalize flows by GDP as of 2019.
In order to test these hypotheses, we run the following regression:

\[ \Delta Y_i = \beta_0 + \beta_1 \text{I-O Trade}_i + \beta_2 \text{I-O Trade Finance}_i + \epsilon_i \]  

(37)

where \( \Delta Y_i \) stands for the economic cost of the COVID-19 shock for sector \( i \) for \( i = 1, \ldots, K \), that we estimate under no lockdown scenario in which no policy action is taken against the pandemic. We measure the sector-level economic cost as the percentage change in overall economic activity (proxied by output \( (Y_i) \) or value added \( (VA_i) \), where value added equals total production minus intermediate inputs i.e., \( VA_i = Y_i - \text{INT}_i \)) for a given sector during pandemic relative to its pre-pandemic level. I-O Trade\(_i\) is the I-O trade linkage for sector \( i \) and I-O Trade Finance\(_i\) is the sector-level proxy variable that captures the interdependence between trade linkages and external finance needs.\(^{39}\)

In equation (37), \( \beta_1 \) captures the impact of I-O trade linkages on COVID-19 related economic losses; \( \beta_2 \) captures the impact of interdependence between trade linkages and external finance needs on COVID-19 related economic losses; and \( \beta_1 + \beta_2 \) captures the total impact of being a small economy linked to production networks and borrowing externally to finance those links.

As a robustness check, we add sectoral FX debt (measured as the ratio of foreign currency debt in total debt as of 2016) to equation (37) as an additional explanatory variable. This variable also captures domestic borrowing in foreign currency in each sector as opposed to only international borrowing that the I-O Trade Finance variable captures. I-O Trade Finance variable might be skewed towards tradeable sectors, while FX debt variable also captures the domestic foreign currency borrowing of non-tradeable sectors such as construction as reflected in Figure 2.

The regression results are highly consistent with our expectations. The positive and highly significant coefficient estimates confirm the importance of international linkages. The results suggest that sectors with stronger I-O links suffer from larger COVID-19 related losses. They further suggest that sectors who finance these stronger production links through capital flows and sectors with higher FX exposure suffer even more, highlighting the additional adverse impact of COVID-19 on EMs with high external debt and domestic FX debt. These findings indicate that if the risk appetite towards EMs does not return soon, Turkey may suffer a prolonged sudden stop type recession due

\(^{39}\)See the footnotes 38 and 39 for further details on the calculation of this variable.
Table 1: **SECTOR-LEVEL REGRESSIONS**

<table>
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<tr>
<th>Dep. Variable</th>
<th>Output Loss (1)</th>
<th>Output Loss (2)</th>
<th>VA Loss (3)</th>
<th>VA Loss (4)</th>
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<td>16.49**</td>
<td>16.03**</td>
<td>16.53**</td>
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<td>35.63**</td>
<td>34.95*</td>
<td>35.80**</td>
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<td>(17.351)</td>
<td>(17.256)</td>
<td>(17.331)</td>
<td>(17.234)</td>
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<td></td>
<td>(0.076)</td>
<td>(0.076)</td>
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</tbody>
</table>

Number of Obs. | 33        | 33        | 33        | 33        |

\( R^2 \)     | 0.12      | 0.2       | 0.12      | 0.2       |

Notes: Table 1 reports the results of estimation of equation (37). Dependent variable is defined as sector-level economic cost of the COVID-19 shock that is measured as the percentage change in overall economic activity for a given sector during pandemic relative to its pre-pandemic level. In Columns (1)-(2) economic activity is proxied by output, and by value added in Columns (3)-(4), respectively. Heteroskedastic-consistent standard errors are reported in parentheses. ***, **, and * indicate significance at the 1%, 5%, and 10% levels, respectively.

5.4 Taking Stock

When we take a look at the experiences of the countries over the course of the pandemic, we note that there are several paths adopted by different countries:

(i) **Full lockdown**: Greece, New Zealand and Denmark provide good examples for an effective full lockdown. Our analysis indicates that this is the policy that minimizes economic costs by containing the pandemic in the most effective way.

(ii) **No lockdown**: Very few countries considered no lockdown since the beginning of the pandemic. No lockdown approach might yield lower economic costs but the death toll is significantly higher. The economic costs are mostly dependent on the changes in demand.

(iii) **No lockdown followed by a full lockdown**: At the beginning of the crisis, UK adopted a no lockdown approach. However, this approach was abandoned later on due to public pressure.
as the death toll rose. UK then adopted a full lockdown policy to contain the pandemic. Our analysis indicates that if the lockdown was not delayed, there would be less mortality and the economic costs would be lower because the lockdown would begin with a smaller number of infections.

(iv) **Partial lockdown followed by full lockdown:** Many countries followed this route including Italy, France, Germany, Spain, Iran, Russia among others. Several of these counties recently announced that they will gradually lift restrictions. The duration of full lockdown is longer than it could have been, had it been implemented earlier. In Italy, for example, a full lockdown went into effect on March 10, and the restrictions are announced to be removed by May 4, after approximately two months under full lockdown.

(v) **Enhanced Partial lockdown:** Turkey started with immediate partial lockdown measures which were enhanced over the course of the pandemic. Schools were closed on March 16 and the businesses were encouraged to work remotely where possible. On March 21, a curfew was imposed for people above the age of 65 and those with chronic diseases. The curfew was extended to those younger than 20 on April 5, effectively putting close to 40% of the population under full lockdown. Furthermore, a full lockdown was implemented on weekends and national holidays starting on April 9 in 31 largest cities which constitute approximately 87% of the population. After about 45 days since the beginning of enhanced partial lockdown measures, $R_0$ is reduced below 1 and the number of new patients is lower than the number of recovered patients as of the last week of April.

Where does this take us? Our analysis indicates that a full lockdown at the early stages of the crisis can bring the pandemic under control relatively quickly. There are countries who implemented this successfully but also countries such as India, who tried an early full lockdown but did not succeed. The individual performance of the country depends on several factors that affect the recovery and the infection rates. An evaluation of Turkey’s performance, two months after the introduction of lockdown measures indicates that Turkey did reasonably well. Potential reasons for the superior

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47 These cities include the 30 metropolitan municipalities and Zonguldak, which constitute close to 79% of the population. On top of these, the age-based restrictions are intact in the rest of Turkey, which increases the number close to 87%.
performance are the remarkable ICU capacity, young population, less care homes, as well as the generally compliant population where government decrees are not challenged. If an enhanced partial lockdown is already in place, which is successful in lowering $R_0$ below 1, then the need for full lockdown may not be imminent. However, our results reflect that the duration of the lockdown would have been shorter if more restrictive measures were adopted right away.

An emerging question at this stage is the removal of restrictions once the pandemic is taken under control. As the duration of lockdown increases, policy makers get anxious about opening up their economies. In this paper, we model demand as a function of the number of infections and combine this with actual spending decline during COVID-19, measured in the data with credit card purchases. Thus, our framework implies that demand would not normalize by the mere attempt of removing the restrictions, so long as the number of infections are sizable. What is worse is that the number of infections would increase again as businesses open. In the model, we do not explicitly incorporate expectations about infections and implicitly assume that the two are highly correlated. Meanwhile, one can imagine a forward looking demand curve, which could be a function of infection expectations rather than the actual number of infections. In this case, leaders might be able to affect expectations about the number of infections and revive demand by removing the restrictions. To the extent that leaders can successfully convey a more optimistic outlook, the negative demand effect that we model in this paper may weaken and the economic costs of prematurely ending a lockdown might decline.

Another imminent issue is the potential second wave once the restrictions are removed. This is particularly a problem for those countries that adopted a full lockdown at the early stages of the crisis and controlled the pandemic in their own countries. If they open their borders, there is the risk of a second wave. If they do not open their borders, then they cannot fully normalize and suffer from an extended partial lockdown given the importance of the amplification effects on economic costs for open economies. The takeaway at this stage is that if a second wave of the COVID-19 virus hits, then an immediate and potentially global lockdown would work in the most effective way.

41See https://blogs.lse.ac.uk/covid19/2020/06/04/how-has-turkey-done-in-its-fight-against-covid-19-the-jury-is-still-out/ for a detailed evaluation of Turkey’s performance based on our framework
6 What are the policy options for EMs?

The previous sections illustrate the economic costs of the pandemic due to a fall in the GDP given the large supply and demand shocks for a small open economy. A lockdown increases the short-term costs but increases the long-term gains by leading the way to a faster recovery. One of the shortcomings of the model is that it does not incorporate the damage to the productive capacity that are caused by company closures. We simply assume that the productive capacity remains intact and the companies jump back to production once the pandemic is over. This is an overly optimistic assumption and in the absence of a comprehensive support program, the liquidity issues would turn into solvency issues. This could lead to unnecessary bankruptcies, a deeper recession and a sluggish recovery.

Indeed, this is exactly why our estimates in the previous section should be interpreted as the lower bound costs of a stimulus package that is necessary to offset the damages of the COVID-19 crisis and keep the economic units alive.

A quickly implemented stimulus package that compensates the income loss due to the lockdown and enables a faster recovery would minimize the long term damage in the production capacity. If the stimulus packages are delayed, on the other hand, more companies would fail, more workers would be laid-off, and demand would decline further. This would then feed into more bankruptcies and elevate the economic costs that quickly become unmanageable. In fact, just as a drowning person needs immediate help or else her organs start to fail, the economy needs immediate help before the companies start to fail. Fiscal transfers can help to ensure that the supply chains are not destroyed, the economic units are functional and ready to go back to production once the pandemic is contained and demand returns. Fortunately, many governments around the world took decisive action. In the case of EMs, however, policy options are limited given the limited fiscal space. As put by former Colombian finance minister, Mauricio Cardenas, “We do not live in whatever it takes region, we can do whatever we can.” Thus, we next discuss the possible ways to finance the economic costs related to COVID in EMs.

42See Kalemli-Ozcan et al. (2018), Gourinchas et al. (2020), and https://voxeu.org/article/proposal-negative-sme-tax.
6.1 Quantitative Easing or Debt Monetization? What is the difference?

The buzz-word in advanced countries for the funding needed to deal with the crisis is "helicopter money." In economists’ jargon, this is called monetary financing where the central bank prints money and transfers resources to firms and households either directly, as in the Federal Reserve’s recent policy of purchasing commercial paper and corporate debt, or indirectly by purchasing government bonds and enabling the government to use the proceeds to deal with the crisis.

In the process of monetary financing of the debt, the central bank’s balance sheet will enlarge, either through direct loans to institutions or through large scale asset purchases (i.e. the so called “quantitative easing” (QE) programs). In a QE, the central bank prints money and buys sizeable amounts of government bonds. In the recent history, this was observed after the Great Recession both in the US and in Europe. The advantage of direct lending is that the liquidity is drained more easily when the loans are paid back.

How is debt monetization different? A central bank typically purchases securities through open market operations to meet the liquidity needs, consistent with its goal of price stability. The technical difference between money printing through an open market purchase and monetizing the debt is slim (Mishkin 2007). Thus, one might argue that QE policies are effectively debt monetization (Orphanides 2017). The Federal Reserve begs to differ and argues that debt monetization refers to a "permanent" source of funding for the government by the central bank and separates QE policies from debt monetization. So as long as the central bank commits to inflation targeting and normalizes its balance sheet when inflationary pressures kick in, asset purchases in the form of QE are not considered debt monetization (Andolfatto and Li 2013). Based on this nuance, one can argue that QE and debt monetization are “observationally equivalent” in the short run, and the difference becomes apparent in the long run, with the central bank’s ability to shrink its balance sheets to counteract inflationary pressures. Hence, using the Federal Reserve’s usage of the term, the criterion for bond purchases to be considered debt monetization is whether the central bank fails to drain the

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44 In the FAQs prepared by the Federal Reserve Board [https://www.federalreserve.gov/faqs/money_12853.htm], it is noted that “The term “printing money” often refers to a situation in which the central bank is effectively financing the deficit of the federal government on a permanent basis by issuing large amounts of currency. This situation does not exist in the United States.”

45 In response to a question from the lawmakers on June 3, 2009, then-chairman Bernanke had noted that the “Federal Reserve will not monetize the debt”: [https://www.c-span.org/video/?c4546512/user-clip-bernanke-debt-monetization](https://www.c-span.org/video/?c4546512/user-clip-bernanke-debt-monetization)
money effectively later on and the money remains in the system permanently such that it leads to inflationary pressures.

In advanced economies, the distinction between QE and debt monetization can be easier to ascertain where the inflation rate is well-anchored and central bank credibility is well established. In fact, the inflation rate has not exceeded the 2 percent target in the US or Europe in the aftermath of large scale quantitative easing policies after the Great Recession. The distinction between QE and debt monetization can get blurry in EMs, however, particularly for those with a history of high inflation and weak credibility.

The key to a successful QE is policy credibility. A badly managed QE would erode credibility of the monetary policy making and de-anchor inflation expectations. This would only escalate the existing crisis by pushing the inflation rate on a higher trajectory and causing sharp depreciations in domestic currency. Hence, if it is not executed properly and the money is not drained from the system at the right time, QE can turn into inflationary debt monetization. In that respect, QE is far more challenging for EM central banks with weaker track records who need to be extra careful to clearly communicate their QE policies for policies to be credible.\textsuperscript{46}

How does this apply to Turkey? There is a rapid increase in CBRT’s bond holdings (Figure 16), which reflects sizable balance sheet expansion. Although the the size of purchases is currently limited to 10% of CBRT’s balance sheet, the statement on March 31 suggests that these limits might be adjusted as needed. Thus, even though the data on purchases is publicly available, there is not yet an announcement of an explicit QE program with a clear exit strategy as is typically done in the advanced economies. We argue that transparent communication on this issue will be even more important going forward and hence rather than the absolute limit on the purchases, how foreign investors’ sentiments change with these purchases is more important. If the limits announced can be data driven and tied to economic conditions in the country in question, they will be more transparent. For example, Padhraic Garvey, global head of debt and rates strategy at ING, said: “If it were to become 5 or 10 per cent of GDP [for countries like South Africa, Indonesia or Turkey], then you are getting into the realm of the danger zone … there would be the potential for considerable

\textsuperscript{46}Foreign investors sentiments are aligned with this view, see https://www.ft.com/content/6a63d700-3a59-4e32-8092-4c818ffaa9d8.
currency depreciation.” Another example is by Sergi Lanau, chief deputy economist at the Institute of International Finance, who said: “Countries with credible, independent central banks such as Chile or those in eastern Europe will be better placed to deal with these pressures, as there is less fear among investors that the central banks themselves will enable excessive and unsustainable government spending.”

Figure 16: CBRT’s Holdings of Government Bonds

NOTES: This figure plots the holdings government bonds of the CBRT. Source: Turkey Data Monitor

If you face a 1.5 percent inflation rate as in the US, and a deep recession is on its way, inflationary consequences of QE may not be imminent. This is because the public does not expect inflation to get out of control despite these excessive measures. There is still belief that the Fed will drain the money from the system at the right time and establish price control. Furthermore, because market participants do not expect the US government to default on its debt, there will not be a sharp decline in demand for US government bonds, which will keep interest rates under control. Turkey has been missing its 5 percent inflation target for some time now and the market sentiment is such that policy...
credibility eroded over the course of years (see Cakmakli and Demiralp (2020)).

The ultimate goal is to convince the market participants that QE will not turn into inflationary debt monetization. A detailed bond purchase calendar can be communicated with spending targets and the conditions under which the money will be drained from the system. One way to increase the transparency of QE could be through a Special Purpose Vehicle (SPV). An SPV would allow central banks to buy government bonds through this entity and separate these COVID-19 related bond purchases from the daily maintenance of monetary policy. The extent of monetary expansion that is solely due to COVID-19 crisis could be easily trackable in this manner. In turn, the money that is generated through this program should be spent in targeted sectors and announced by the government.

While a transparent and well executed QE could provide immediate funding that is necessary to deal with COVID-19 crisis, it would likely be insufficient. The case of Turkey, as several other EMs, requires a joint thinking of fiscal needs and capital flows given foreign financing needs. In the next section, we discuss the magnitude of external funding needs in Turkey.

6.2 External Funding Needs, Capital Controls, and the Role of External Anchor

Considering the facts that (i) the total amount of external debt that needs to be paid or rolled-over in 2020 is 23 percent of GDP and (ii) the current open FX position of the entire corporate sector as of January 2020 (which is -$175 billion) is almost 25 percent of GDP, Turkey has serious external funding needs. Our analysis in the earlier sections highlighted that those sectors with stronger trade linkages and higher external funding needs are more vulnerable during COVID-19 crisis. In order to get a sense of the economic outlook for these sectors and highlight the potential risks, this section investigates the external funding needs for the Turkish economy in the post COVID-19 world.

The rapid increase in the risk premium (Figure 3) shows that cost of external borrowing is getting higher for most EMs, which will make it harder to rollover external debt. News reports reflect that the investor sentiment is deteriorating given the large short-term external debt of the banking

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49 See e.g. https://www.ft.com/content/a2a178f8-a109-11e9-974c-ad1c6ab5efd4 or https://www.wsj.com/articles/turkish-lira-falls-after-central-bank-defiantly-leaves-rates-unchanged-1532433561 for the unfavorable perception of credibility in alternative media outlets

50 See Unsal, Papageorgiou and Garbers, 2020 on central bank communication for unconventional policies.
Many EMs are in a similar spot where the largest capital outflow so far was observed in Brazil with 12 billion USD outflow from the stock market and 19 billion USD outflow from the bond market before May 2020. Investors cited “fear” not only about the pandemic but also about the uncertainty surrounding the economic policies. These sizable numbers for large EMs such as Turkey and Brazil suggest that several out-of-the-box policies may be needed.

In terms of policy alternatives for Turkey, a rate hike to compensate for the risk premium could be an option. Nevertheless, even in the absence of a large demand shock such as COVID, tight monetary policy may not be fully effective under a large risk-off shock, as shown by Kalemli-Özcan (2019). During risk-off shocks, raising policy rates to defend the currency and to bring back the capital flows backfires based on historical experiences, especially in countries with low policy credibility and high risk premia. In addition, given the large negative demand shock, the necessary accommodation can only be provided by loosening the monetary policy as many EMs, including Turkey, have done so far. On the other hand, EMs with high external debt cannot rely on rate cuts entirely either and need to find the balance between supporting domestic demand and limiting the volatility of their domestic currencies.

A swap agreement with the Federal Reserve or another international institution can help to address the liquidity needs arising from COVID-19 crisis, but this may not be enough on its own if the pandemic extends and weakens the businesses ability to remain in operation and service their debt, given the size of the domestic fiscal needs and external financing needs. On May 20, Turkey announced that the existing swap line with Qatar is expanded to $15 billion. However, no swap arrangements with G20 countries with whom Turkey has sizable trade relationships has been announced yet. It should also be noted that the Federal Reserve did not expand the list of countries that are eligible for a swap line since the GFC. Because Turkey was not it the original list during GFC, it is rather unlikely that a swap line can be arranged with the Fed at this time.

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51 See https://www.reuters.com/article/turkey-banks-funding-idUSL8N2DF21S.
52 See Financial Times, June 2, 2020, Brazil Record Outflows.
53 See Basu et al. (2020) who models optimal policy mix for EMs that will allow them monetary policy autonomy.
54 The average emergency swap agreement granted by the IMF is about $11bn, https://www.imf.org/en/News/A rticles/2020/04/07/sp040920-SMs2020-Curtain-Raiser, and the total outstanding amount granted by the Federal Reserve’s international swap lines is $18.9 bn as of April 14, https://apps.newyorkfed.org/markets/autorates/fxswap.
55 https://blogs.imf.org/2020/03/16/policy-action-for-a-healthy-global-economy/#Xm9rH13Oc-7A.twitter
Yet another alternative is to introduce capital controls to trap both residents’ and non-residents’ foreign currency assets in Turkey, which in turn will limit the TL depreciation. Notice that, while capital controls on inflows during a boom might reduce the future probability of a sudden stop and protect financial stability, capital controls on outflows during a large risk-off shock might have unintended consequences. Historically, capital controls on outflows have not been very effective (see Loungani and Mauro (2000)). Most likely, what EMs need is more capital inflows, not controls on outflows. If EMs breach contracts, they might face legal action by private creditors, which compromises their future access to capital markets. Especially when there is significant foreign investment in local currency bonds of EMs, a panic by foreign investors would put even more pressure on local currency and inflation. As a result, such controls might further erode the policy credibility and scare foreign capital during the recovery phase when it will be most needed, especially for a country like Turkey who is already heavily dependent on external funding. Hence, going back to our previous analogy, one might try to keep horses in the barn but needs to be also careful about not scaring the horses that can scar them for a longer time.

Since the beginning of COVID-19 outbreak, Turkey took certain steps that were perceived as mild forms of capital control by the markets, mostly on domestic residents. Such steps took the form of limiting TL supply in international swap markets, notifying the government regarding sizable FX transfers abroad, or restricting the TL transactions of large custodian banks. More recently on May 24, the tax on exchange rate and gold transactions has been increased from 0.2 percent to 1 percent. These measures deter foreign investors not only because they limit their ability to move their capital around but also because they give the impression of random changes in the legislation. Unpredictable changes in regulations that are viewed as interventions to the free market mechanism have the potential to discourage future capital inflows and damage policy credibility. If the global recovery takes longer than a few years, interest rates would remain low in advanced countries and foreign investors would likely be willing to invest in riskier EM assets driven by search for yield.

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57 Mendoza (2010), Basu et al. (2020).
58 Bartolini and Drazen (1997).
59 These arguments reflect the views of foreign investors. See e.g. https://www.reuters.com/article/turkey-banks-funding-idUSL8N2DF21S.
60 See, e.g., Financial Times article on February 9 https://www.ft.com/content/8eb5152b-4bd5-11ea-95a0-43d18cc715f5 or on April 30 https://www.ft.com/content/06c85177-39ae-42fd-90cb-1d6bc6cb11a.
motives similar to the period after 2007-2009 crisis. In that case, costs of too early capital controls might be not accessing finance in the medium term since it is well known that once capital controls on outflows put in place, it takes a long time to remove them. In terms of their benefits, it could be argued that capital controls would prevent further dollarization that might be triggered by the TL liquidity injected through the QE program. Nevertheless, dollarization can also be prevented without capital controls if there is enough policy credibility to keep inflation expectations anchored.

One final alternative is a debt moratorium on foreign lenders. However, since foreign lenders are private creditors (and not official creditors), this would involve complicated debt default and debt restructuring. Unless private creditors offer the moratorium in a synchronized way as suggested by Rogoff and Reinhart (2020), a disorderly one would again hamper the medium to long-term credibility.

While there are still significant uncertainties regarding the future course of the pandemic, the alternatives that we laid out in this section aim to provide a better sense of potential scenarios if the crisis lengthens and global financial conditions tighten. If the external financial needs cannot be met through the market mechanism, then granting FX liquidity through arrangements with international institutions seem to be the optimal solution that would minimize future risks and speed up economic recovery. Our estimates in this paper suggest that those sectors that rely more on external borrowing are hit harder during COVID-19. Thus, keeping the flow of FX credit is particularly important for these sectors to maintain their production capacities.

Although EMs did not observe a crisis similar to COVID-19, their history is full of crises, where different stabilization policies were employed. Turkey’s own historical experience involves a transparent QE program to meet the immediate liquidity needs combined with guaranteed external finance through an international institution in the aftermath of 2001 crisis. This was a combination of banking crisis, sovereign debt crisis and a balance of payments crisis. During that time, Turkey employed a sizable asset purchase program under an IMF program, keeping inflation expectations in check with a transparent inflation targeting framework. We lay out the details of this episode in the next section.

62 For example, it took Iceland 10 years to lift the controls put during the 2007-2008 crisis. See https://voxeu.org/article/capital-controls-21st-century
7 Lessons from History: Bond Purchases under an IMF program

When the financial crisis hit in February 2001, Turkey already had a standby agreement with the IMF, ongoing since December 1999.\(^{63}\) State banks and Savings Deposit Insurance Fund (SDIF) experienced significant losses during the 2001 crisis, which elevated their liquidity needs. In order to meet their liquidity needs and recapitalize these institutions, government securities were transferred to these institutions. The securities were then sold to the CBRT to receive cash to cover their liquidity needs. The size of securities purchases reached approximately 8 percent of GDP during that time. In turn, the CBRT drained the excess liquidity gradually through conventional methods (i.e., either through reverse repos or through its overnight borrowing facility) in order to prevent an unintended decline in market rates (see Statement of Intent, 2001).\(^{64}\) When the ongoing 1999 program was deemed to be insufficient, a new and more comprehensive standby agreement was signed in 2002 which particularly aimed at lowering inflation expectations by strengthening policy credibility.\(^{65}\)

The asset purchases that were undertaken in the post-2001 period took place at the same time Turkey started a new regime to take inflation under control. An amendment to the Central Bank Law (no: 1211) in 2001 granted operational independence. In the same amendment, it was stated that direct bond purchases from the government would continue until November 2001. The bond purchase program (debt monetization) was acknowledged in the 2002 agreement as well.\(^{66}\) Figure 17 displays the overall size of CBRT’s total assets in real terms. We observe that CBRT’s total assets increased about 122 percent, from January 2000 to November 2001. To provide perspective, the Federal Reserve’s balance sheet increased 100 percent after four rounds of QEs from December 2008.

\(^{63}\)The history of lending arrangements with the IMF are available at the following link: https://www.imf.org/external/np/fin/tad/extarr2.aspx?memberKey1=980&date1key=2008-03-31
\(^{64}\)https://www.imf.org/external/np/loi/2001/tur/02/index.htm
\(^{65}\)As stated in Statement of Intent, 2001: “This program is a continuation of the one initiated in late 1999, with the support of a stand-by arrangement with the International Monetary Fund. It shares the same strategy: disinflate the Turkish economy, strengthen the fiscal accounts, and reform the structure of the Turkish economy as a condition for setting economic growth on a sustainable basis and moving Turkey closer to its goal of joining the European Union. However, the program’s policies have been significantly strengthened, in response to the recent crisis that led to the float of the Turkish lira on February 22, 2001, including through increased emphasis on transparency, accountability, and good governance in both the private and public sectors. In support of our strengthened program, we request that the arrangement be augmented by the equivalent of SDR 6.3624 billion and that the purchases scheduled through end-2001 be rephased and would consequently be subject to reviews which are expected to be completed during May, June, July, September, and November 2001.”
through October 2014.

**Figure 17: Total Assets of CBRT**

![Chart of CBRT Total Assets](chart.png)

**NOTES:** This figure plots total assets of CBRT in real terms, normalized by CPI. The figure ends prior to 2007-2009 global crisis. After this date, CBRT’s balance sheet expanded again thanks to capital inflows in the post-crisis environment. Source: Turkey Data Monitor.

The 2002 standby agreement with the IMF not only met the external funding needs but it also provided the much needed credibility to boost confidence in the program and prevent excessive depreciation in the local currency. The comprehensive package of reforms that was supported by the standby agreement was instrumental in limiting domestic funding needs. In turn, this restricted the volume of asset purchases, taking the pressure off inflation expectations. Once the liquidity needs subsided, the liquidity was drained from the system promptly and transparently. As a result of these coordinate efforts, there was a successful disinflation performance as shown below in the absence of volatility in the exchange rate (Figures 18a and 18b).

An essential part of disinflationary policies in the post crisis period involved lowering inflation expectations. The program anchored inflation expectations by ensuring that large scale bond purchases would not turn into debt monetization. In order to prevent bond purchases from causing a
substantive increase in inflation expectations and restore investor confidence in the program, public finance and debt management laws were introduced to improve fiscal transparency and accountability. Furthermore, the budgetary impact of the additional funds needed to restructure the banking system was offset by increasing public savings in other areas to keep the overall budget under control. This step limited the extent of public borrowing and prevented market interest rates from rising further.

Figure 18: Inflation Rate and Exchange Rate

(a) Inflation Rate

(b) USD/TL Exchange Rate

NOTES: (a) This figure plots inflation rate for Turkey, which measured as year-on-year change of CPI. (b) This figure plots USD/TL nominal exchange rate for Turkey. Source: Turkey Data Monitor

8 Conclusion

Containing the pandemic as soon as possible is an urgent obligation to save human lives. Nevertheless, we have to act now also to deal with the economic fallout from the pandemic as the economic costs can be substantial. As put by the IMF (2020), this is “a crisis like no other” with potentially far more disastrous implications for emerging markets and developing economies relative to advanced economies.

We develop a small open economy SIR-multi-sector-macro model and calibrate it to Turkey using real time data. The annual cost of the COVID-19 crisis that we estimated ranges between 5.8 percent and 11 percent of the Turkish GDP depending on the effectiveness and the duration of the lockdown.
We estimate that the most cost effective full lockdown scenario implies a quarterly GDP contraction of 17 percent. Delays in full lockdown, prematurely ending the lockdown, or a combination of full lockdown with partial lockdown increases the toll. While the numbers are rather scary and unforgiving, we take comfort in our prologue that “Best safety lies in fear” and urge caution in removing the lockdown restrictions, not only to save more lives but also to minimize the economic toll.

Turkey has large external financing needs, with an upcoming external debt payment in 2020 as large as 23 percent of the GDP ($169 bn). Given these numbers, it seems rather challenging for Turkey to rollover its foreign currency debt and finance its domestic debt solely through a QE-type program unless monetary policy is transparent and communicated clearly similar to the QE policies of advanced countries. A well-designed and comprehensive package of macroeconomic and structural reforms will increase policy credibility and reduce external borrowing costs and spreads. Funding from international institutions would further signal support for the policy package and help to cover the financing gap in international markets. This would further lower the external finance premium and ease financial strains faced during the pandemic.

There are still substantial uncertainties ahead of us regarding the course of the pandemic. In the absence of global coordination, countries that successfully contain the virus struggle about how to enable international trade and travel with the fear of a second wave. In this paper, we highlight that the role of global coordination is essential for open economies with international I-O linkages. If the lockdown could be implemented with global synchronization, demand would return to prepandemic levels faster and the economic costs of the pandemic could be kept at a minimum level. As this is not done so far, all the policy options should be on the table for EMs given the dynamic nature of this crisis with new information arriving every day. Looking ahead, should a second wave hit, a globally coordinated full lockdown would allow for the fastest global recovery.

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References


Barrot, Jean-Noel, Basile Grassi, and Julien Sauvagnat, “Sectoral effects of social distancing,” 2020. Available at SSRN.


A APPENDIX

List of Figures and Tables:

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Figure A.1: The Structure of OECD Inter-Country Input-Output Table

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NOTES: This table illustrates the structure of OECD Inter-Country Input-Output Table (ICIO), which represents the breakdown of output corresponding to 36 industries and 69 countries, giving us a matrix of $2484 \times 2484$ entries. In any industry-country combination, the output (Y) equals intermediate use (Z) plus final demand (F) of 36 industries in 69 countries. Industry list can be found in Table A.2. Further, in any industry-country combination, final demand sums the following components of expenditures over 69 countries. fd1: Households Final Consumption Expenditure (HFCE); fd2: Non-Profit Institutions Serving Households (NPISH); fd3: General Government Final Consumption (GGFC); fd4: Gross Fixed Capital Formation (GFCF); fd5: Change in Inventories and Valuables (INVNT); fd6: Direct purchases by non-residents (NONRES); fd7: Statistical Discrepancy (DISC).
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<td>Argentina</td>
<td>3</td>
<td>Adopted measures (totaling about 3.0 percent of GDP, 1.2 percent in the budget and 1.8 percent off-budget, based on authorities' estimates)</td>
</tr>
<tr>
<td>Australia</td>
<td>10.8</td>
<td>Total expenditure and revenue measures of A$194 billion (9.9 percent of GDP). The Commonwealth government has committed to spend about an additional A$8 billion (0.3 percent of GDP). State and Territorial governments also announced fiscal stimulus packages, totaling amounting to A$11.5 billion (0.6 percent of GDP).</td>
</tr>
<tr>
<td>Brazil</td>
<td>6.5</td>
<td>The authorities announced a series of fiscal measures adding up to 6.5 percent of GDP. Public banks are expanding credit lines for businesses and households, with a focus on supporting working capital (credit lines add up to over 3 percent of GDP), and the government will back a 0.5 percent of GDP credit line to cover pay-roll costs.</td>
</tr>
<tr>
<td>Canada</td>
<td>8.4</td>
<td>Key tax and spending measures (8.4 percent of GDP, $193 billion CAD).</td>
</tr>
<tr>
<td>China</td>
<td>3.8</td>
<td>An estimated RMB 2.6 trillion (or 2.5 percent of GDP) of fiscal measures or financing plans have been announced. The overall fiscal expansion is expected to be significantly higher, reflecting the effect of already announced additional measures such as an increase in the ceiling for special local government bonds of 1.3 percent of GDP.</td>
</tr>
<tr>
<td>France</td>
<td>19</td>
<td>The authorities have announced an increase in the fiscal envelope devoted to addressing the crisis to €110 billion (nearly 5 percent of GDP, including liquidity measures), from an initial €45 billion included in an amending budget law introduced in March. A new draft amending budget law has been introduced on April 16. This adds to an existing package of bank loan guarantees and credit reinsurance schemes of €315 billion (close to 14 percent of GDP).</td>
</tr>
<tr>
<td>Germany</td>
<td>31.6</td>
<td>The federal government adopted a supplementary budget of €156 billion (4.9 percent of GDP). The government is expanding the volume and access to public loan guarantees for firms of different sizes and credit insurers increasing the total volume by at least €757 billion (23 percent of GDP). In addition to the federal government's fiscal package, many state governments (Länder) have announced own measures to support their economies, amounting to €46 billion in direct support and €27bn in state-level loan guarantees (Authors: Another 3.7% of GDP).</td>
</tr>
<tr>
<td>India</td>
<td>1.1</td>
<td>Finance Minister Sitharaman on March 26 announced a stimulus package valued at approximately 0.8 percent of GDP. These measures are in addition to a previous commitment by Prime Minister Modi that an additional 150 billion rupees (about 0.1 percent of GDP) Numerous state governments have also announced measures thus far amount to approximately 0.2 percent of India’s GDP.</td>
</tr>
<tr>
<td>Indonesia</td>
<td>2.8</td>
<td>In addition to the first two fiscal packages amounting to IDR 33.2 trillion (0.2 percent of GDP), the government announced a major stimulus package of IDR 40.5 trillion (2.6 percent of GDP) on March 31, 2020.</td>
</tr>
<tr>
<td>Italy</td>
<td>26.4</td>
<td>On March 17, the government adopted a €25 billion (1.4 percent of GDP) 'Cura Italia' emergency package. On April 6, the Liquidity Decree allowed for additional state guarantees of up to €400 billion (25 percent of GDP).</td>
</tr>
<tr>
<td>Japan</td>
<td>21.1</td>
<td>On April 7 (partly revised on April 20), the Government of Japan adopted the Emergency Economic Package Against COVID-19 of ¥17.1 trillion (21.1 percent of GDP).</td>
</tr>
<tr>
<td>Mexico</td>
<td>0.7</td>
<td>To request additional resources from Congress, that could reach up to 180 billion pesos (0.7 percent of 2019 GDP). AND The week of April 19 the President further announced an austerity program for public expenditures including wage reductions and a hiring in order to free up 2.5 percent of GDP to finance additional health expenditures and priority investment.</td>
</tr>
<tr>
<td>Republic of Korea</td>
<td>10</td>
<td>Direct measures amount to 0.8 percent of GDP (approximately KRW 16 trillion). On March 24, President Moon announced a financial stabilization plan of KRW 100 trillion (5.3 percent of GDP). This was augmented by a further KRW 35 trillion (1.8 percent of GDP) on April 22 through additional measures. On April 22, President Moon announced a key industry stabilization fund would be established for KRW 40 trillion (2.1 percent of GDP)</td>
</tr>
<tr>
<td>Russia</td>
<td>2.1</td>
<td>The total cost of the fiscal package is currently estimated at 2.1 percent of GDP.</td>
</tr>
<tr>
<td>Saudi Arabia</td>
<td>5</td>
<td>A SAR 70 billion ($18.7 billion or 2.8 percent of GDP) private sector support package was announced on March 20. they will reduce spending in non-priority areas of the 2020 budget by SAR 50 billion (2.0 percent of GDP) to accommodate some of these new initiatives within the budget envelope. on April 3, the government authorized the use of the unemployment insurance fund (SANED) to provide support for wage benefits, within certain limits, to private sector companies who retainer their Saudi staff (SAR 9 billion, 0.4 percent of GDP). on April 15, additional measures to mitigate the impact on the private sector were announced, including temporary electricity subsidies to commercial, industrial, and agricultural sectors (SAR 0.9 billion) and resource support to the health sector was increased to SAR 47 billion.</td>
</tr>
<tr>
<td>South Africa</td>
<td>0.2</td>
<td>[<a href="https://www.globalpolicywatch.com/2020/04/south-africas-economic-response-to-the-covid-19-pandemic/">https://www.globalpolicywatch.com/2020/04/south-africas-economic-response-to-the-covid-19-pandemic/</a>]</td>
</tr>
<tr>
<td>Spain</td>
<td>11.7</td>
<td>Key measures (about 1.8 percent of GDP, 11.7 billion; depending on the usage and duration of the measures the amount could be higher). In addition, the government of Spain has extended up to €100 billion government guarantees for firms and self-employed. Other measures include additional funding for the Instituto de Credito Oficial (ICO) credit (€10 billion); introduction of a special credit line for the tourism sector through the ICO (€400 million);</td>
</tr>
<tr>
<td>Turkey</td>
<td>5</td>
<td>A TL100 billion package was announced. This consists of TL7.5 billion (81.6 billion or 1.5 percent of GDP) in fiscal measures, as well as TL 25 billion ($3.8 billion or 0.5 percent of GDP) for the doubling the credit guarantee fund. Gradually, this package increased to be 5% of GDP.</td>
</tr>
<tr>
<td>United Kingdom</td>
<td>18.8</td>
<td>Policy measures adding £86 billion in 2020-21. Coronavirus business interruption loan scheme and the Covid Corporate Financing Facility: the business interruption loan scheme was announced as up to £330 billion of support for businesses. Source: [<a href="https://obr.uk/coronavirus-reference-scenario/">https://obr.uk/coronavirus-reference-scenario/</a>]</td>
</tr>
</tbody>
</table>

NOTES: This table reports the COVID-19 relief packages (as percent of GDP) by the G20 countries along with the details of the fiscal packages. Source: IMF Policy Tracker unless otherwise noted. Access Date: April 29, 2020.
Figure A.2: Sovereign Bond Issuance in Turkey

Turkey
Rollover risk, bonds issued, bonds due

Note: The graphic shows external bonds issued against bonds due on a quarterly basis between 2000q1 and 2020q2. Crisis spells, i.e. periods of severe currency and financial crises, are highlighted in grey. Red areas (2001q2, 2008q4, 2018q2, 2020q2) represent periods during which Turkey experienced a financial crisis and the rollover ratio fell below 1.
<table>
<thead>
<tr>
<th>OECD ISIC Code</th>
<th>Definition</th>
<th>Proximity Index</th>
<th>Teleworkable Share</th>
</tr>
</thead>
<tbody>
<tr>
<td>01T03</td>
<td>Agriculture, forestry and fishing</td>
<td>0.86</td>
<td>0.06</td>
</tr>
<tr>
<td>05T06</td>
<td>Mining and extraction of energy producing products</td>
<td>1.08</td>
<td>0.32</td>
</tr>
<tr>
<td>07T08</td>
<td>Mining and quarrying of non-energy producing products</td>
<td>1.06</td>
<td>0.14</td>
</tr>
<tr>
<td>09</td>
<td>Mining support service activities</td>
<td>1.21</td>
<td>0.20</td>
</tr>
<tr>
<td>10T12</td>
<td>Food products, beverages and tobacco</td>
<td>1.12</td>
<td>0.13</td>
</tr>
<tr>
<td>13T15</td>
<td>Textiles, wearing apparel, leather and related products</td>
<td>1.09</td>
<td>0.20</td>
</tr>
<tr>
<td>16</td>
<td>Wood and products of wood and cork</td>
<td>1.03</td>
<td>0.15</td>
</tr>
<tr>
<td>17T18</td>
<td>Paper products and printing</td>
<td>1.08</td>
<td>0.22</td>
</tr>
<tr>
<td>19</td>
<td>Coke and refined petroleum products</td>
<td>1.11</td>
<td>0.22</td>
</tr>
<tr>
<td>20T21</td>
<td>Chemicals and pharmaceutical products</td>
<td>1.06</td>
<td>0.25</td>
</tr>
<tr>
<td>22</td>
<td>Rubber and plastic products</td>
<td>1.10</td>
<td>0.18</td>
</tr>
<tr>
<td>23</td>
<td>Other non-metallic mineral products</td>
<td>1.08</td>
<td>0.18</td>
</tr>
<tr>
<td>24</td>
<td>Basic metals</td>
<td>1.09</td>
<td>0.14</td>
</tr>
<tr>
<td>25</td>
<td>Fabricated metal products</td>
<td>1.08</td>
<td>0.21</td>
</tr>
<tr>
<td>26</td>
<td>Computer, electronic and optical products</td>
<td>1.03</td>
<td>0.54</td>
</tr>
<tr>
<td>27</td>
<td>Electrical equipment</td>
<td>1.07</td>
<td>0.29</td>
</tr>
<tr>
<td>28</td>
<td>Machinery and equipment, nec</td>
<td>1.06</td>
<td>0.29</td>
</tr>
<tr>
<td>29</td>
<td>Motor vehicles, trailers and semi-trailers</td>
<td>1.09</td>
<td>0.19</td>
</tr>
<tr>
<td>30</td>
<td>Other transport equipment</td>
<td>1.06</td>
<td>0.31</td>
</tr>
<tr>
<td>31T33</td>
<td>Other manufacturing; repair and installation of machinery and equipment</td>
<td>1.07</td>
<td>0.32</td>
</tr>
<tr>
<td>35T39</td>
<td>Electricity, gas, water supply, sewerage, waste and remediation services</td>
<td>1.08</td>
<td>0.29</td>
</tr>
<tr>
<td>41T43</td>
<td>Construction</td>
<td>1.21</td>
<td>0.19</td>
</tr>
<tr>
<td>45T47</td>
<td>Wholesale and retail trade; repair of motor vehicles</td>
<td>1.13</td>
<td>0.37</td>
</tr>
<tr>
<td>49T53</td>
<td>Transportation and storage</td>
<td>1.18</td>
<td>0.21</td>
</tr>
<tr>
<td>55T56</td>
<td>Accomodation and food services</td>
<td>1.26</td>
<td>0.10</td>
</tr>
<tr>
<td>58T60</td>
<td>Publishing, audiovisual and broadcasting activities</td>
<td>1.11</td>
<td>0.69</td>
</tr>
<tr>
<td>61</td>
<td>Telecommunications</td>
<td>1.07</td>
<td>0.58</td>
</tr>
<tr>
<td>62T63</td>
<td>IT and other information services</td>
<td>1.01</td>
<td>0.88</td>
</tr>
<tr>
<td>64T66</td>
<td>Financial and insurance activities</td>
<td>1.02</td>
<td>0.79</td>
</tr>
<tr>
<td>68</td>
<td>Real estate activities</td>
<td>1.10</td>
<td>0.54</td>
</tr>
<tr>
<td>69T82</td>
<td>Other business sector services</td>
<td>1.09</td>
<td>0.46</td>
</tr>
<tr>
<td>84</td>
<td>Public admin. and defence; compulsory social security</td>
<td>1.16</td>
<td>0.39</td>
</tr>
<tr>
<td>85</td>
<td>Education</td>
<td>1.22</td>
<td>0.86</td>
</tr>
<tr>
<td>86T88</td>
<td>Human health and social work</td>
<td>1.28</td>
<td>0.35</td>
</tr>
<tr>
<td>90T96</td>
<td>Arts, entertainment, recreation and other service activities</td>
<td>1.18</td>
<td>0.34</td>
</tr>
</tbody>
</table>

**Notes:** This table provides the physical proximity index along with the share of those who can work remotely for the industries. To obtain these two industry-level values, we calculate the weighted average of the values corresponding to the occupations in each industry using the Occupational Employment Statistics (OES) provided by the U.S. Bureau of Labor Statistics (BLS). OES data follows four-digit NAICS codes to classify the industries. In order to convert the proximity data to OECD ISIC codes, we make use of the correspondence table between 2017 NAICS and ISIC Revision 4 Industry Codes, provided by the U.S. Census Bureau. We obtain the physical proximity values at the occupation level from the O*NET database. O*NET collects the physical proximity information through surveys with the following categories: (1) I don’t work near other people (beyond 100 ft.); (2) I work with others but not closely (e.g., private office); (3) Slightly close (e.g., shared office); (4) Moderately close (at arm’s length); (5) Very close (near touching). We divide the category values by 3 to make category (3) our benchmark. Specifically, a proximity value that is larger than 1 indicates a closer proximity than the “shared office” level, and a proximity value that is smaller than 1 corresponds to sparse working conditions. We create a single physical proximity value for each occupation by taking the weighted average of the normalized category values. We calculate the proximity values at the industry level after removing the teleworkable portion of the employees. We use Dingel and Neiman (2020)’s list of teleworkable occupations to capture the proportion of employment that can be fulfilled at remote locations in each industry.
<table>
<thead>
<tr>
<th>OECD ISIC</th>
<th>Definition</th>
<th>Change</th>
<th>Explanation</th>
</tr>
</thead>
<tbody>
<tr>
<td>01T03</td>
<td>Agriculture, forestry and fishing</td>
<td>100%</td>
<td>Based on projections on the manufacturing sector using data on credit card spending from the database of CBRT and computations using historical data.</td>
</tr>
<tr>
<td>05T06</td>
<td>Mining and extraction of energy producing products</td>
<td>100%</td>
<td>Based on projections on the manufacturing sector using data on credit card spending from the database of CBRT and computations using historical data.</td>
</tr>
<tr>
<td>07T08</td>
<td>Mining and quarrying of non-energy producing products</td>
<td>100%</td>
<td>Based on projections on the manufacturing sector using data on credit card spending from the database of CBRT and computations using historical data.</td>
</tr>
<tr>
<td>09</td>
<td>Mining support service activities</td>
<td>100%</td>
<td>Based on projections on the manufacturing sector using data on credit card spending from the database of CBRT and computations using historical data.</td>
</tr>
<tr>
<td>10T12</td>
<td>Food products, beverages and tobacco</td>
<td>100%</td>
<td>Based on projections on the manufacturing sector using data on credit card spending from the database of CBRT and computations using historical data.</td>
</tr>
<tr>
<td>13T15</td>
<td>Textiles, wearing apparel, leather and related products</td>
<td>50%</td>
<td>Based on estimates using data on credit card spending from the database of CBRT.</td>
</tr>
<tr>
<td>16</td>
<td>Wood and products of wood and cork</td>
<td>90%</td>
<td>Based on projections on the manufacturing sector using data on credit card spending from the database of CBRT and computations using historical data.</td>
</tr>
<tr>
<td>17T18</td>
<td>Paper products and printing</td>
<td>90%</td>
<td>Based on projections on the manufacturing sector using data on credit card spending from the database of CBRT and computations using historical data.</td>
</tr>
<tr>
<td>19</td>
<td>Coke and refined petroleum products</td>
<td>75%</td>
<td>Based on estimates using data on credit card spending from the database of CBRT.</td>
</tr>
<tr>
<td>20T21</td>
<td>Chemicals and pharmaceutical products</td>
<td>90%</td>
<td>Based on estimates using data on credit card spending from the database of CBRT.</td>
</tr>
<tr>
<td>22</td>
<td>Rubber and plastic products</td>
<td>90%</td>
<td>Based on projections on the manufacturing sector using data on credit card spending from the database of CBRT and computations using historical data.</td>
</tr>
<tr>
<td>23</td>
<td>Other non-metallic mineral products</td>
<td>90%</td>
<td>Based on projections on the manufacturing sector using data on credit card spending from the database of CBRT and computations using historical data.</td>
</tr>
<tr>
<td>24</td>
<td>Basic metals</td>
<td>90%</td>
<td>Based on computations using historical data and sectoral reports.</td>
</tr>
<tr>
<td>25</td>
<td>Fabricated metal products</td>
<td>90%</td>
<td>Based on computations using historical data and sectoral reports.</td>
</tr>
<tr>
<td>26</td>
<td>Computer, electronic and optical products</td>
<td>100%</td>
<td>Based on estimates using data on credit card spending from the database of CBRT.</td>
</tr>
<tr>
<td>27</td>
<td>Electrical equipment</td>
<td>90%</td>
<td>Based on projections on the manufacturing sector using data on credit card spending from the database of CBRT and computations using historical data.</td>
</tr>
<tr>
<td>28</td>
<td>Machinery and equipment, nec</td>
<td>90%</td>
<td>Based on projections on the manufacturing sector using data on credit card spending from the database of CBRT and computations using historical data.</td>
</tr>
<tr>
<td>29</td>
<td>Motor vehicles, trailers and semi-trailers</td>
<td>70%</td>
<td>Based on other countries’ experiences and sectoral reports.</td>
</tr>
<tr>
<td>30</td>
<td>Other transport equipment</td>
<td>70%</td>
<td>Same as automobiles.</td>
</tr>
<tr>
<td>31T33</td>
<td>Other manufacturing; repair and installation of machinery and equipment</td>
<td>90%</td>
<td>Based on projections on the manufacturing sector using data on credit card spending from the database of CBRT and computations using historical data.</td>
</tr>
<tr>
<td>35T39</td>
<td>Electricity, gas, water supply, sewerage, waste and remediation services</td>
<td>100%</td>
<td>No change.</td>
</tr>
<tr>
<td>41T43</td>
<td>Construction</td>
<td>75%</td>
<td>Based on computations using historical data and sectoral reports.</td>
</tr>
<tr>
<td>45T47</td>
<td>Wholesale and retail trade; repair of motor vehicles</td>
<td>110%</td>
<td>Based on estimates using data on credit card spending from the database of CBRT.</td>
</tr>
<tr>
<td>49T53</td>
<td>Transportation and storage</td>
<td>80%</td>
<td>Based on other countries’ experiences and sectoral reports.</td>
</tr>
<tr>
<td>55T56</td>
<td>Accommodation and food services</td>
<td>25%</td>
<td>Based on estimates using data on credit card spending from the database of CBRT.</td>
</tr>
<tr>
<td>58T60</td>
<td>Publishing, audiovisual and broadcasting activities</td>
<td>85%</td>
<td>Based on estimates using data on credit card spending from the database of CBRT.</td>
</tr>
<tr>
<td>61</td>
<td>Telecommunications</td>
<td>100%</td>
<td>Based on estimates using data on credit card spending from the database of CBRT.</td>
</tr>
<tr>
<td>62T63</td>
<td>IT and other information services</td>
<td>100%</td>
<td>Based on estimates using data on credit card spending from the database of CBRT.</td>
</tr>
<tr>
<td>64T66</td>
<td>Financial and insurance activities</td>
<td>100%</td>
<td>Based on estimates using data on credit card spending from the database of CBRT.</td>
</tr>
<tr>
<td>68</td>
<td>Real estate activities</td>
<td>60%</td>
<td>Based on estimates using data on credit card spending from the database of CBRT.</td>
</tr>
<tr>
<td>69T82</td>
<td>Other business service activities</td>
<td>85%</td>
<td>Based on estimates using data on credit card spending from the database of CBRT.</td>
</tr>
<tr>
<td>84</td>
<td>Public admin. and defence; compulsory social security</td>
<td>125%</td>
<td>Median Package size 5%. Public spending is close to 2% of GDP. In line with other business services.</td>
</tr>
<tr>
<td>85</td>
<td>Education</td>
<td>85%</td>
<td>Based on other countries’ experiences and sectoral reports.</td>
</tr>
<tr>
<td>86T88</td>
<td>Human health and social work</td>
<td>100%</td>
<td>Based on other countries’ experiences and sectoral reports.</td>
</tr>
<tr>
<td>90T96</td>
<td>Arts, entertainment, recreation and other service activities</td>
<td>25%</td>
<td>Based on other countries’ experiences and sectoral reports.</td>
</tr>
</tbody>
</table>

**NOTES:** This table provides the demand changes at the sectoral level along with the explanations. We use publicly available data and the credit card spending data from the Central Bank of Republic of Turkey (CBRT) to calculate the estimated demand change during the pandemic in each industry, which is categorized based on OECD ISIC Codes.
Table A.4: List of the Lockdown Sectors

Panel A: Lockdown Sectors

<table>
<thead>
<tr>
<th>NACE Rev. 2</th>
<th>Definition</th>
</tr>
</thead>
<tbody>
<tr>
<td>01</td>
<td>Crop and animal production, hunting and related service activities</td>
</tr>
<tr>
<td>1071</td>
<td>Manufacture of bread; manufacture of fresh pastry goods and cakes</td>
</tr>
<tr>
<td>1811</td>
<td>Printing of newspapers</td>
</tr>
<tr>
<td>1920</td>
<td>Manufacture of refined petroleum products</td>
</tr>
<tr>
<td>21</td>
<td>Manufacture of basic pharmaceutical products and pharmaceutical preparations</td>
</tr>
<tr>
<td>35</td>
<td>Electricity, gas, steam and air conditioning supply</td>
</tr>
<tr>
<td>36</td>
<td>Water collection, treatment and supply</td>
</tr>
<tr>
<td>4646</td>
<td>Wholesale of pharmaceutical goods</td>
</tr>
<tr>
<td>4730</td>
<td>Retail sale of automotive fuel in specialised stores</td>
</tr>
<tr>
<td>4773</td>
<td>Dispensing chemist in specialised stores</td>
</tr>
<tr>
<td>4774</td>
<td>Retail sale of medical and orthopaedic goods in specialised stores</td>
</tr>
<tr>
<td>4920</td>
<td>Freight rail transport</td>
</tr>
<tr>
<td>4941</td>
<td>Freight transport by road</td>
</tr>
<tr>
<td>5224</td>
<td>Cargo handling</td>
</tr>
<tr>
<td>53</td>
<td>Postal and courier activities</td>
</tr>
<tr>
<td>60</td>
<td>Programming and broadcasting activities</td>
</tr>
<tr>
<td>61</td>
<td>Telecommunications</td>
</tr>
<tr>
<td>639</td>
<td>Other information service activities</td>
</tr>
<tr>
<td>75</td>
<td>Veterinary activities</td>
</tr>
<tr>
<td>86</td>
<td>Human health activities</td>
</tr>
<tr>
<td>87</td>
<td>Residential care activities</td>
</tr>
</tbody>
</table>

Panel B: Additional Sectors

<table>
<thead>
<tr>
<th>NACE Rev. 2</th>
<th>Definition</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>Manufacture of food products</td>
</tr>
<tr>
<td>1722</td>
<td>Manufacture of household and sanitary goods and of toilet requisites</td>
</tr>
<tr>
<td>463</td>
<td>Wholesale of food, beverages and tobacco</td>
</tr>
<tr>
<td>4711</td>
<td>Retail sale in non-specialised stores with food, beverages or tobacco</td>
</tr>
<tr>
<td>472</td>
<td>Retail sale of food, beverages and tobacco in specialised stores</td>
</tr>
<tr>
<td>4781</td>
<td>Retail sale via stalls and markets of food, beverages and tobacco products</td>
</tr>
</tbody>
</table>

Notes: This table provides the list of the lockdown sectors. We use the decree issued by the Turkish Ministry of Interior on April 10, 2020 to identify these industries. This lockdown was effective for only two days and cover those given in Panel A. We supplement the list with those available in Panel B.
Table A.5: CBRT CREDIT CARD SPENDING TITLES CORRESPONDING TO OECD ISIC SECTORS

<table>
<thead>
<tr>
<th>CBRT</th>
<th>Definition</th>
<th>OECD ISIC Code</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>Total</td>
<td></td>
</tr>
<tr>
<td>2</td>
<td>Car Rental</td>
<td>69T82</td>
</tr>
<tr>
<td>3</td>
<td>Car Rental-Sales/Service/Parts</td>
<td>45T47</td>
</tr>
<tr>
<td>4</td>
<td>Petrol Stations</td>
<td>19</td>
</tr>
<tr>
<td>5</td>
<td>Various Food</td>
<td>10T12</td>
</tr>
<tr>
<td>6</td>
<td>Direct Marketing</td>
<td>45T47</td>
</tr>
<tr>
<td>7</td>
<td>Education/Stationary</td>
<td>45T47</td>
</tr>
<tr>
<td>8</td>
<td>Electric &amp; Electronic Goods, Computers</td>
<td>26</td>
</tr>
<tr>
<td>9</td>
<td>Clothing and Accessory</td>
<td>13T15</td>
</tr>
<tr>
<td>10</td>
<td>Airlines</td>
<td>49T53</td>
</tr>
<tr>
<td>11</td>
<td>Service</td>
<td>58T60 &amp; 68 &amp; 69T82</td>
</tr>
<tr>
<td>12</td>
<td>Accomodation</td>
<td>55T56</td>
</tr>
<tr>
<td>13</td>
<td>Club/Association/ Social Services</td>
<td>55T56</td>
</tr>
<tr>
<td>14</td>
<td>Casino</td>
<td>55T56</td>
</tr>
<tr>
<td>15</td>
<td>Jewellery</td>
<td>45T47</td>
</tr>
<tr>
<td>16</td>
<td>Marketing and Shopping Centers</td>
<td>45T47</td>
</tr>
<tr>
<td>17</td>
<td>Furnishing and Decoration</td>
<td>31T33</td>
</tr>
<tr>
<td>18</td>
<td>Contractor Services</td>
<td>41T43</td>
</tr>
<tr>
<td>19</td>
<td>Health/Health Products/Cosmetics</td>
<td>20T21</td>
</tr>
<tr>
<td>20</td>
<td>Travel Agencies/Forwarding</td>
<td>69T82</td>
</tr>
<tr>
<td>21</td>
<td>Insurance</td>
<td>64T66</td>
</tr>
<tr>
<td>22</td>
<td>Telecommunication</td>
<td>61</td>
</tr>
<tr>
<td>23</td>
<td>Building Supplies, Hardware, Hard Goods</td>
<td>25</td>
</tr>
<tr>
<td>24</td>
<td>Food</td>
<td>55T56</td>
</tr>
<tr>
<td>25</td>
<td>Government/Tax Payments</td>
<td>84</td>
</tr>
<tr>
<td>26</td>
<td>Private Pensions</td>
<td>64T66</td>
</tr>
<tr>
<td>27</td>
<td>Others</td>
<td></td>
</tr>
<tr>
<td>28</td>
<td>E-commerce Transactions</td>
<td>62T63</td>
</tr>
<tr>
<td>29</td>
<td>Mail or Phone Shopping</td>
<td></td>
</tr>
<tr>
<td>30</td>
<td>Customs Payments</td>
<td>84</td>
</tr>
</tbody>
</table>

Notes: This table provides the concordance that we use to match the titles used in the CBRT’s credit card spending data with the OECD ISIC Codes.

Table A.6: LIST OF THE ACTIVE SECTORS IN PUBLIC ADMINISTRATION DURING FULL LOCKDOWN

<table>
<thead>
<tr>
<th>Type</th>
<th>Size</th>
<th>Source</th>
</tr>
</thead>
<tbody>
<tr>
<td>Public (All)</td>
<td>2820095</td>
<td><a href="http://www.sbb.gov.tr/kamu-istihdami/">http://www.sbb.gov.tr/kamu-istihdami/</a></td>
</tr>
<tr>
<td>Security</td>
<td>273000</td>
<td><a href="https://tr.wikipedia.org/wiki/Emniyet_Genel_M%C3%BCd%C3%BCrl%C3%BC%C4%9F%C3%BC">https://tr.wikipedia.org/wiki/Emniyet_Genel_M%C3%BCd%C3%BCrl%C3%BC%C4%9F%C3%BC</a></td>
</tr>
<tr>
<td>Health</td>
<td>642184</td>
<td><a href="https://www.saglik.gov.tr/TR,11588/istatistik-yilliklar.html">https://www.saglik.gov.tr/TR,11588/istatistik-yilliklar.html</a></td>
</tr>
</tbody>
</table>

Share 37.77%

Notes: This table provides the list of occupations in Public Administration that work during full lockdown, together with the number of people within those occupations. The data sources are provided as well. The share of the active sub-sectors in the entire sector is 37%.