

Longer-run economic consequences of pandemics^{*}

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Abstract

What are the medium- to long-term effects of pandemics? How do they differ from other economic disasters? We study major pandemics using the rates of return on assets stretching back to the 14th century. Significant macroeconomic after-effects of pandemics persist for decades, with real rates of return substantially depressed, in stark contrast to what happens after wars. Our findings are consistent with the neoclassical growth model: capital is destroyed in wars, but not in pandemics; pandemics instead may induce relative labor scarcity and/or a shift to greater precautionary savings.

JEL classification codes: E43, F41, I10, N10, N30, N40, O40.

Keywords: pandemics, wars, depressions, real interest rate, natural rate, local projections.

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1. INTRODUCTION

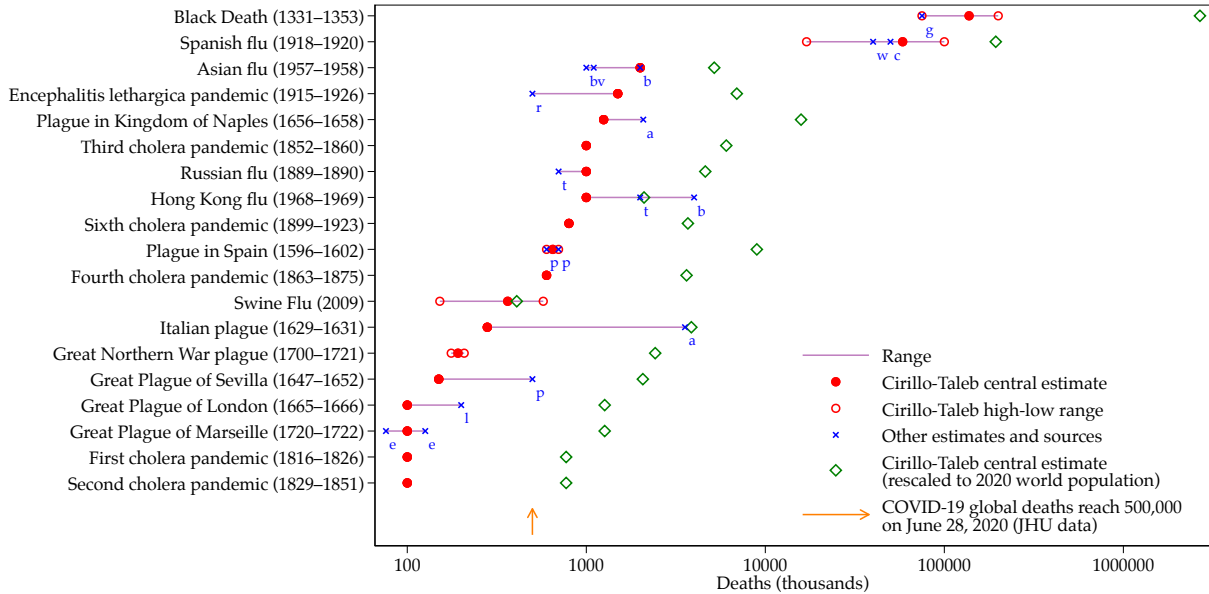
How should we think about the medium- to long-term macroeconomic effects of global pandemics? This is an important question for policymakers given the massive economic fallout from the COVID-19 outbreak. The first instinct is to draw from the experience of previous natural disasters and armed conflicts, for which the twentieth century provides ample evidence. That instinct turns out to be wrong. In this paper we use the history of major pandemics, major wars, and the rates of return since the 14th century to shed light on this problem.

In many studies of pandemics, analysis has understandably focused on short-term impacts. Even then, direct measures based on data from past episodes are not generally available (e.g., in the U.S., [Meltzer, Cox, and Fukuda, 1999](#)). An alternative would be to look at microeconomic outcomes of a given population in episodes for which high-quality administrative data are available (e.g., in Sweden, [Karlsson, Nilsson, and Pichler, 2014](#)).

Absent such data, economic historians have to use more aggregated data at the regional or national level to study the relationship between pandemic incidence and economic outcomes (e.g., the 1918 flu epidemic across U.S. states [Brainerd and Siegler, 2003](#); [Barro, Ursúa, and Weng, 2020](#)). But again, most historical studies have typically focused on one event in one country or region and have traced local outcomes for up to a decade at most.

Of course, the most devastating pandemic of the last millennium, the Black Death, has attracted a great deal of scholarly attention. Economists and historians debate its pivotal role in economic, social, and political change, particularly in Europe. Events such as the Peasants' Rebellion in England feature centrally in a narrative of rising worker power, and the data speak to an emergence of labor scarcity seen in a positive deviation in the path of real wages. This shock left England with a 25% to 40% drop in labor supply, a roughly 100% increase in real wages, and a decline in rates of return on land from about 8% to 5% ([Clark, 2007, 2010](#)). But it is an open question how representative the macroeconomic responses in the case of the Black Death are of large pandemics in general.

Figure 1: Nineteen major pandemic events from the past with at least 100,000 estimated deaths



Notes: The main data are taken from the recently published study of Cirillo and Taleb (2020). See the references therein. Additional estimates of pandemic mortality are taken from: (g) George M. Gould and Walter L. Pyle. 1896. *Anomalies & Curiosities of Medicine*. New York: Bell; (w) Wang-Shick Ryu. 2017. *Molecular Virology of Human Pathogenic Viruses*. Amsterdam: Academic Press. (c) CDC. <https://www.cdc.gov/flu/pandemic-resources/1918-pandemic-h1n1.html>; (b) *Encyclopaedia Britannica*; (v) Cécile Viboud et al. 2016. Global Mortality Impact of the 1957–1959 Influenza Pandemic; *Journal of Infectious Diseases* 213(5):738–45; (r) R.T. Ravenholt and W.H. Foege. 1982. 1918 influenza, encephalitis lethargica, parkinsonism. *Lancet* 2(8303):860–864; (a) Alfani and Murphy (2017); (t) Paul V. Targonski and Gregory A. Poland. 2017. Influenza. In *International Encyclopedia of Public Health*, 2nd ed., edited by William C. Cockerham. Amsterdam: Academic Press. (p) Stanley G. Payne. 1973. *A History of Spain and Portugal*. Madison, Wisc.: University of Wisconsin Press. (l) James Leasor. 1962. *The Plague and The Fire*. London: George Allen and Unwin. (e) Cindy Ermus. 2015. The Plague of Provence: Early Advances in the Centralization of Crisis Management. *Arcadia* (9). [Arcadia Collection: Disaster Histories.]

Evidence: Past pandemic events and lessons for COVID-19. Here we take a macroeconomic view of the consequences of pandemics, and study the average effect of pandemics across all major events since the Black Death, looking at outcomes up to 40 years out. In large scale pandemics, effects will be felt across whole economies, or across wider regions, for two reasons: either because the infection itself is widespread, or because trade and market integration—in capital and/or labor markets—eventually propagates the economic shock across borders.

To that end, our focus is on European pandemics since macroeconomic data are only available in this region before modern times. We study rates of return on assets using a dataset stretching back to the 14th century, focusing on 19 major pandemic events where more than 100,000 people died. We also look at more limited evidence on real wages. The events are listed in Figure 1 and, as is clear, the inability of researchers to precisely estimate

death tolls in past pandemics hinders our ability to consider event scaling in what follows. Even for the most famous pandemic event, the Black Death, [Herlihy \(1997\)](#) asserts: “The Black Death of 1348 and 1349, and the recurrent epidemics of the fourteenth and fifteenth centuries, were the most devastating natural disasters ever to strike Europe. We cannot cite exact losses; there are no global figures.” Our [Figure 1](#) shows that numerical uncertainty afflicts all death tolls up to, and including, the 20th century. The baseline range of estimates are taken from the newly published study of [Cirillo and Taleb \(2020\)](#), with alternatives from other sources. For some pandemics, the range of estimates varies by a factor of 5 or 10 (e.g., The Italian Plague). Many vary by a factor of 2, such as the “small” plagues, London, Sevilla, Marseille. Estimates for the Black Death itself range from 75 to 200 million. Remarkably, even for as recent an event as the Hong Kong Flu of 1968–69, estimates vary by a factor of 4, from 1 to 4 million. For the Cholera epidemics only a single, round number is given in sources, suggesting a rough guess. Perhaps this should be no surprise: even now in real time we see authorities, by accident or design, misreporting incorrect case and death counts for COVID-19. Further scaling problems are created across time by the varying size of the local or world population, which [Cirillo and Taleb \(2020\)](#) attempt to control for by scaling up death tolls in past events to a 2020 estimate using the estimated growth of world population over time. These rescaled figures are also shown in [Figure 1](#) but they do not change the ranking materially, and they remain subject to the same wide margins or error. If anything, they tell us that there have been 17 “large pandemics” in our sample and a further 2 “super pandemics”: the Black Death and the Spanish Flu. Given this discussion, we identify events purely by a binary indicator, and do not try to scale, although we do consider robustness of results to the inclusion or exclusion of the two super pandemics.

Does COVID-19 belong in this comparison group? To add context, the scenarios contemplated as early as March 16, 2020 by [Ferguson et al \(2020\)](#) placed COVID-19 as the most serious episode since the 1918 pandemic. Absent non-pharmaceutical interventions, these researchers estimated the death toll at 510,000 in Britain and 2.2 million in the U.S. Aggressive and recurrent suppression strategies would reduce the toll by a factor of 10

approximately, according to these authors.¹ Worldwide, that could still leave COVID-19 as the second most devastating event in over a century. As we write, global deaths totaled over 500,000 by end-June 2020, so COVID-19 will become the 20th large pandemic event of the kind described in [Figure 1](#), and may eventually go far beyond the 1 million mark.

Theory: The natural rate and economic mechanisms after a pandemic. Here our main interest is the response of the real natural rate of interest to a pandemic event. Introduced by Wicksell, and central to modern macroeconomic theory and empirics ([Laubach and Williams, 2003](#); [Wicksell, 1898](#); [Woodford, 2003](#)), the real natural rate of interest is the level of real returns on safe assets which equilibrates an economy's savings supply and investment demand—while keeping prices stable. This latent equilibrium variable can therefore serve as a useful barometer of medium-term fluctuations in economic dynamism. In stagnant societies, savings are plentiful relative to investment opportunities (relatedly, see, e.g. [Summers, 2014](#), and his revival of the secular stagnation hypothesis). Formally, in the canonical [Ramsey \(1928\)](#) model of neoclassical growth, it is well known that population slowdown or greater preference to save depresses the natural rate ([Rachel and Smith, 2017](#)).

In the very long run, from century to century, the natural rate may drift slowly for technological, political, or institutional reasons. But over a horizon of around 10–20 years, medium-term deviations will dominate. Economic theory presumptively indicates that pandemics could be felt in transitory downward shocks to the natural rate over such horizons: investment demand is likely to wane, as labor scarcity in the economy suppresses the need for high investment. At the same time, savers may react to the shock with increased saving, either behaviorally as precautionary motives mount in bad times (cf. [Malmendier and Nagel, 2011](#)), or simply to replace lost wealth used up during the peak of the calamity. Indeed, 2020Q2 data for the United States showed a pronounced spike in household saving.

¹However, the uncertainty around these estimates is quite large. Also, the ability and willingness of some countries to actually implement and sustain such suppression strategies properly is open to question. At the time of writing (30th June 2020), both Britain and the U.S. have put some suppression strategies to work, so the implied lower targets might yet be met, although we do not yet know how the eagerness to reopen the economy may prevent that goal from being reached as the year progresses.

2. DATA

To study the macroeconomic responses to historic pandemic events, we use data collected over many years by many economic historians collectively, and then pulled together gradually to form a continuous time series measuring economic indicators at annual frequency in cities, regions, and countries from the 14th century to the present.

Historical interest rates from 1314 to 2018 compiled by [Schmelzing \(2020\)](#) are available at the Bank of England’s data repository, where further details on data sources are given. The dataset covers France (1387–2018), Germany (1326–2018), Italy (1314–2018), the Netherlands (1400–2018), Spain (1400–1729, 1800–2018), and the U.K. (1314–2018). European real interest rates are constructed by weighting real interest rates on long-term debt by GDP shares ([Maddison, 2010](#)).² The underlying assets are debt contracts “which are not contracted short-term, which are not paid in-kind, which are not clearly of an involuntary nature, which are not intra-governmental, and which are made to executive political bodies.” More limited data for other time series are available for the UK in the same data repository. Data on real wages for Great Britain from 1311 to 2016 come from the real consumption wage series of [Clark \(2007\)](#) extended by [Thomas and Dimsdale \(2017\)](#). Data on real GDP per capita for England from 1311 to 2016 are obtained from the real GDP per capita at market prices series available in [Thomas and Dimsdale \(2017\)](#).

3. EMPIRICAL DESIGN

Pandemics, like natural disasters, offer a unique opportunity to study how economies work. They are like a randomized control trial, but at a much larger scale, with microbiology providing the natural assignment mechanism. Not surprisingly, our empirical approach shares similar features with such a trial. Specifically, we use history as a guide to compare the average path after a pandemic against its predicted counterfactual.

²The weighting follows [Schmelzing \(2020\)](#), but results are robust to alternative weightings and also hold when we employ country-specific estimation as shown below. Aggregation weights do not sway the result.

Specifically, given what we observe, and using a historical sample, one can easily construct the expected value of a future counterfactual outcome of interest. Similarly, one can compute the expectation of that same outcome following pandemics. This is the key idea behind the local projections estimator that we use (Jordà, 2005).

In particular, we characterize the response of the natural rate to a pandemic as

$$\tau(h) = E(r_{t+h}^* - r_t^* | P_t = 1; \Omega_t) - E(r_{t+h}^* - r_t^* | P_t = 0; \Omega_t), \quad (1)$$

where $r_{t+h}^* - r_t^*$ refers to the change in the natural rate from the year the pandemic ends to a future time h years later. P_t is a dummy variable that is 1 if there is a pandemic ending in year t , and is 0 otherwise, and Ω_t refers to the information set available at time t .³

We estimate $\tau(h)$ using local projections, specifically, using the set of regressions:

$$r_{t+h}^* - r_{t-1}^* = \alpha^h + \beta^h P_t + \sum_{l=1}^L \rho_l^h r_{t-1-l}^* + e_{t+h}^h; \quad h = 1, \dots, H, \quad (2)$$

where clearly $\hat{\tau}(h) = \hat{\beta}_h$. We choose 10 lags ($L = 10$) though the estimate $\hat{\beta}_h$ is unbiased whether we include these controls or not.⁴

Our estimate of the natural rate of interest, r_t^* , is based on the following simple state space model given the data that is available:

$$r_t = r_t^* + u_t, \quad (3)$$

$$r_t^* = r_{t-1}^* + v_t.$$

Here, the natural rate is a latent unobserved variable that follows a random walk. Such a model is flexible enough to capture any secular trends without the need to specify them

³Most pandemics last for more than one year. We adopt as a timing convention the year when they end.

⁴Note that because pandemics are unpredictable and completely exogenous to the economy, we could omit Ω_t from the conditioning information set. $\hat{\tau}(h)$ would still be unbiased. Including right-hand side information improves the efficiency of the estimator. In other instances, when treatment is determined by observables, this will not be true. Thus $\hat{\tau}(h)$ is basically the cumulative Average Treatment Effect of the pandemic on interest rates, h periods later.

directly. The observed rate of interest r fluctuates around the natural rate r^* . The error terms are assumed to be Normal and Equation 3 can be estimated using the Kalman filter and maximum likelihood methods.

4. RESULTS

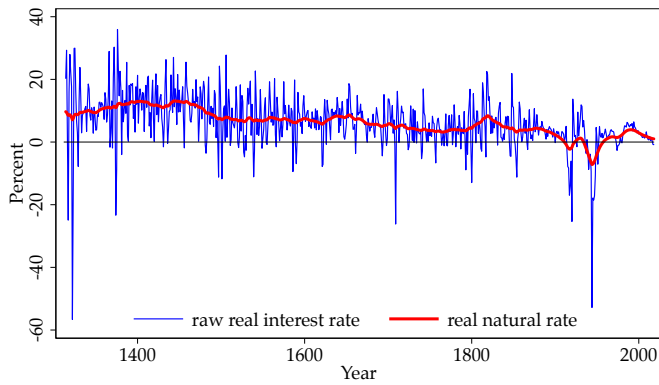
These tools deliver the estimate of the natural rate shown in Figure 2a, based on an aggregate using data from France, Germany, the Netherlands, Italy, Spain and the U.K. We simply call this aggregate “Europe.” The figure displays the raw data on interest rates, along with our estimate of the natural rate of interest. Our estimates of the natural rate show the now well-documented secular trend decline over the centuries, from about 10% in medieval times, to 5% at the start of the industrial revolution, and nowadays hovering near 0%. It is easy to see that our estimate of the natural rate goes a long way towards addressing the considerable annual volatility seen in the raw data. Beyond measurement error, that noise also reflects wild fluctuations in harvests, armed conflict, and other events to which pre-industrial societies were exposed to a much greater degree versus today. With industrialization and modern finance, those fluctuations diminished considerably.

Figure 2b contains our main result, and displays $\hat{\tau}(h)$, the response of the natural rate to a pandemic, 1 to 40 years into the future. Pandemics have effects that last for decades. Following a pandemic, the natural rate of interest declines for years thereafter, reaching its nadir about 20 years later, with the natural rate about 150 bps lower had the pandemic not taken place. At about four decades later, the natural rate returns to the level it would be expected to have had if the pandemic not taken place.

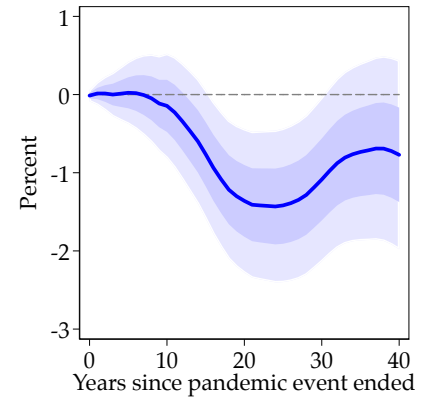
These results are staggering and speak of the disproportionate effects on the labor force relative to land (and later capital) that pandemics had throughout the centuries. It is well known that after major recessions associated with financial crises, history shows that real safe rates can be depressed for 5 to 10 years (Jordà, Schularick, and Taylor, 2013), but the responses here display even more pronounced persistence.

Figure 2: *The European real natural rate and response after pandemics*

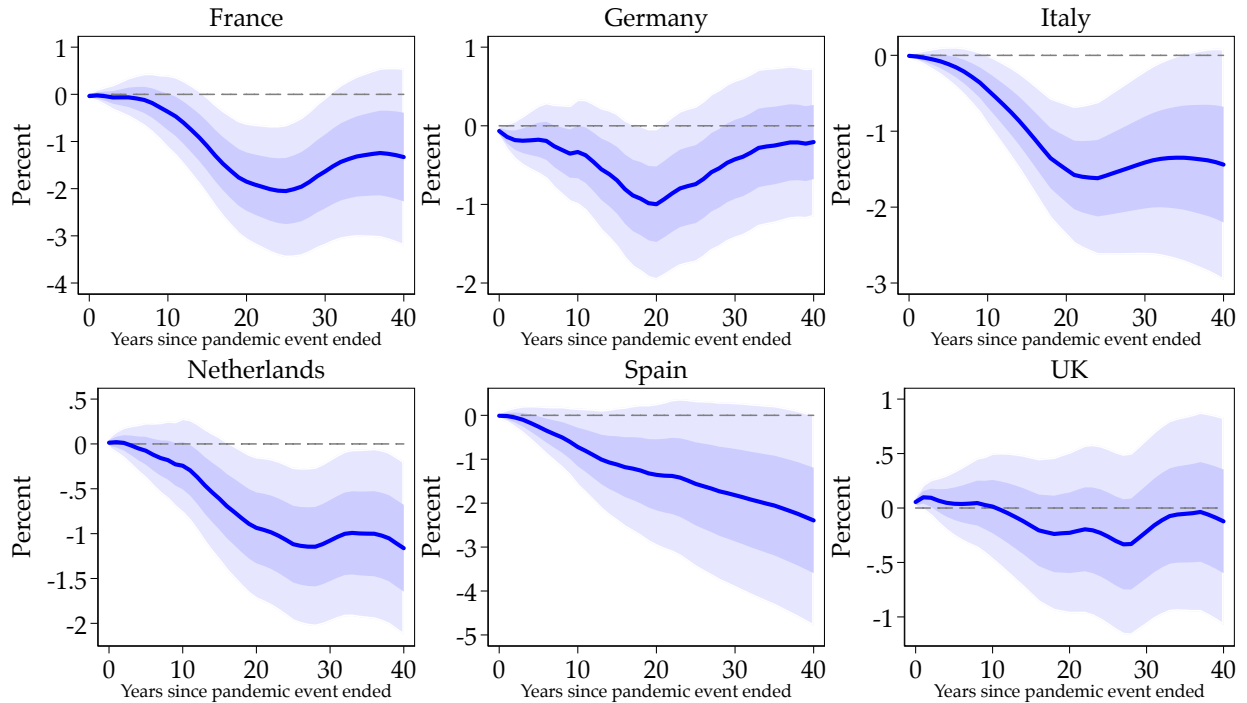
(a) *European real natural rate of interest, 1315–2018*



(b) *Response after pandemics*



(c) *Country-specific response after pandemics*



Notes: Raw interest data from [Schmelzing \(2020\)](#). The real rate is based on the model in [Equation 3](#). Response calculated using [Equation 2](#). Shaded areas are 1 and 2 s.e. bands around response estimates. See text.

However, did all countries in the Europe experience pandemics in the same manner shown? Is our choice of aggregation a problem? To answer that question, we turn to [Figure 2c](#) where we present similar responses of the natural rate for each of the component economies: France, Germany, Italy, the Netherlands, Spain, and the U.K.

The heterogeneity of the responses turns out to be informative. At one end we have countries like France, Italy, and Spain where the effects of pandemics are much larger (3%–4% for France, Italy and Spain) in contrast to the Anglo-Saxon bloc of the Germany, the Netherlands and the U.K., with far more modest effects on the natural rate.

This heterogeneity reflects, among other explanations, the timing of the pandemics across countries, the relative exposure of each country to the pandemic, the relative size of the working population, and how industrialized one economy was relative to another.

5. PANDEMICS VERSUS WARS

A natural concern might be omitted variables, specifically the historical occurrence of other macro-salient events that could persistently disturb real interest rates. One obvious example is war. A clear confounding potential stems from the privations of wartime conditions which make diseases more likely. For example, a plague, possibly the first recorded influenza outbreak, occurred among the Athenians during the Peloponnesian War ([Thucydides](#), 2.47–54). However, the correlation of war and disease outbreaks is far from being one.

On the other hand, for the economic indicator of interest in this study, the bias could easily go the other way. Sovereign bond markets were to a large degree an innovation whose origin was military necessity, and the fiscal state had perhaps its most important role as an instrument of war ([Brewer, 1990](#)). Yet here the burden of raising large sums via debt finance could just as easily imply higher real interest rates via conventional crowding out arguments, or via risk-premium (e.g., default) channels, or simply due to capital scarcity created by wartime physical capital destruction (a feature absent in pandemics).

To address this concern here we control for wars, by using an indicator variable War_t which is set equal to one in any year in which war-time deaths in Europe exceed 20,000. We obtain the war-time military personnel casualties from [Schmelzing \(2020\)](#) dataset who draws data from Table 4.1 in [Levy \(1983\)](#).^{5,6}

We then estimate an augmented local projection to include controls for war. Specifically, we estimate

$$r_{t+h}^* - r_t^* = \alpha^h + \beta^h P_t + \gamma^h War_t + \sum_{l=1}^L \rho_l^h r_{t-l}^* + \sum_{l=1}^L \phi_l^h War_{t-l} + e_{t+h}^h; \quad h = 1, \dots, H, \quad (4)$$

where we include also the lagged values of this indicator as controls, where the coefficients γ^h are the impulse response of the real interest to a war event in year t .⁷

[Figure 3a](#) shows the result of this exercise. Our main finding is robust. The dynamic response of the real natural rates to pandemics is as before, and slightly amplified: lower for 30–40 years and in a statistically significant way. But as we anticipated, the effect of war goes the other way: wars tend to leave real interest rates elevated for 30–40 years and in an economically (and statistically) significant way, as shown in panel (a) of [Table 1](#).

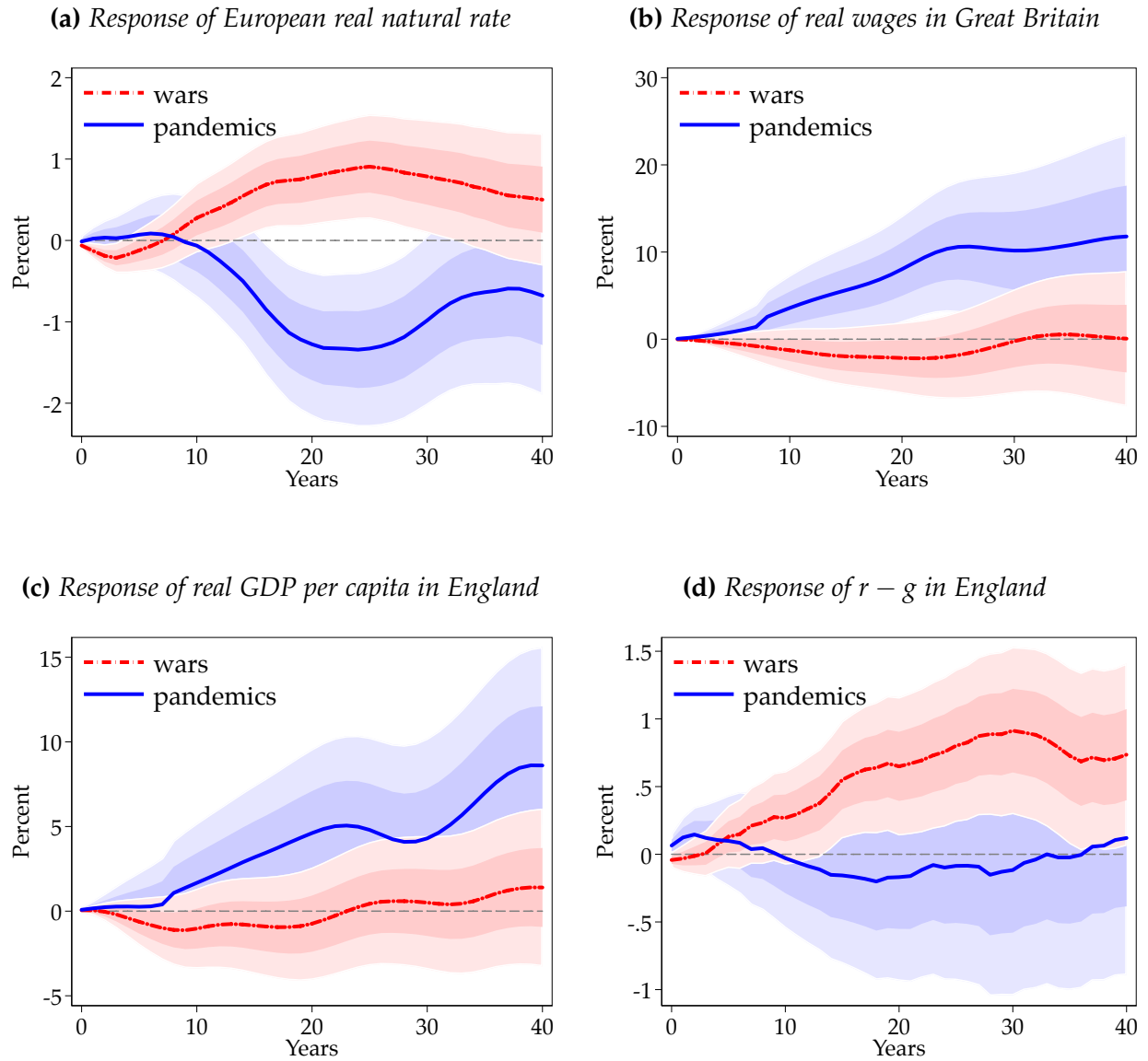
Real wages. By defining pandemics as events with 100,000 or more deaths, we have thus far identified episodes with large contractions in the labor force and, hence, in the ratio of labor to capital. We see this as one explanation for the response of the real natural rate of interest. If so, we should see a countervailing response in real wages, and this provides another consistency check. To explore whether this is indeed the case, we use a similar local projection estimator in [Figure 3b](#) where instead of the real natural rate, we use real wages in Great Britain as the response variable and in the control set.

⁵Total number of battle deaths are divided by duration of the battle (in years) to obtain an annual series of battle deaths. Since battles lasted more than a year, sometimes more than a decade, we think 20,000 deaths per year is comparable to the pandemic death toll threshold of 100,000 we employ.

⁶Readers interested in the debate on severity of pre-19th century epidemics may refer to [Alfani and Murphy \(2017\)](#), [Roosen and Curtis \(2018\)](#), and [Biraben \(1975\)](#), among others.

⁷Wars often last for several years. We focus on war-year pairs with 20,000 casualties or more. Hence, to eliminate the overlapping effect of large casualties felt over consecutive years, we add up to 10 lags of our dummy variable as a right-hand side variable.

Figure 3: Responses after pandemics and wars



Notes: Response calculated using Equation 4. Final two panels use trends estimated for real wages in Great Britain and real GDP per capita in the UK. Details on trend estimation is provided in Appendix A. Shaded areas are 1 and 2 s.e. bands around response estimates. See text.

Table 1: Responses after pandemics and wars

Horizon year	$h = 0$	$h = 10$	$h = 20$	$h = 30$	$h = 40$
<i>(a) Responses of European real natural rate at years 0 to 40 (change from year -1 baseline)</i>					
Pandemics	-0.01 (0.04)	-0.06 (0.32)	-1.28*** (0.44)	-0.98* (0.52)	-0.68 (0.61)
Wars	-0.06 (0.02)	0.28 (0.21)	0.78*** (0.29)	0.79** (0.35)	0.50 (0.41)
H_0 : Pandemics = Wars, p -value	0.26	0.38	0.00	0.01	0.11
Observations	694				
<i>(b) Responses of real wages in Great Britain at years 0 to 40 (change from year -1 baseline)</i>					
Pandemics	0.05 (0.05)	3.58* (1.87)	8.04*** (3.06)	10.17** (4.48)	11.79** (5.83)
Wars	-0.02 (0.03)	-1.27 (1.25)	-2.16 (2.04)	-0.24 (2.98)	0.06 (3.88)
H_0 : Pandemics = Wars, p -value	0.25	0.03	0.01	0.05	0.10
Observations	696				
<i>(c) Responses of real GDP per capita in England at years 0 to 40 (change from year -1 baseline)</i>					
Pandemics	0.08 (0.08)	1.67 (1.76)	4.61* (2.48)	4.30 (2.95)	8.61** (3.50)
Wars	0.05 (0.06)	-1.03 (1.18)	-0.74 (1.66)	0.50 (1.97)	1.40 (2.33)
H_0 : Pandemics = Wars, p -value	0.77	0.20	0.07	0.29	0.09
Observations	736				
<i>(d) Responses of $r - g$ in England at years 0 to 40 (change from year -1 baseline)</i>					
Pandemics	0.06 (0.04)	-0.03 (0.26)	-0.17 (0.38)	-0.12 (0.46)	0.12 (0.51)
Wars	-0.04 (0.03)	0.27 (0.17)	0.65** (0.26)	0.91*** (0.31)	0.74** (0.34)
H_0 : Pandemics = Wars, p -value	0.04	0.34	0.08	0.07	0.31
Observations	692				

Notes: Standard errors in parentheses. See text.

The response of real wages is in the opposite direction to the response of the real natural rate, with its effects being felt over decades. The figure shows that real wages grow gradually so that forty years out the cumulative deviation in the real wage is +15%. Contrast the response of wages after wars, which remain pretty much unchanged. A formal test of the differences is provided in panel (b) of [Table 1](#). These results match the predictions of the neoclassical model, and accord with historical narratives: the Black Death induced labor scarcity in the European economy, and pushed real wages up. In equilibrium, this went hand in hand with lower returns to capital.

Real GDP per capita. Finally, we display the response of real GDP per capita in [Figure 3c](#) for England (the only economy for which we have long-run data). Here again, the contrast between wars and pandemics is self-evident (with formal tests provided in panel (c) of [Table 1](#)). The destruction of capital (and loss of lives) during wars, has a noticeable negative effect on labor productivity that is absent in pandemics, where in fact, labor productivity increases.

Debt sustainability and $r - g$. While the short-run fallout from pandemics looks similar to other economic disasters—large and sudden declines in economic activity—the medium- to long-run economic consequences are staggeringly different, as we have shown. And these differences matter for policymakers. In any such abrupt downturn, the textbook response is to either borrow to smooth the shock, or to pursue aggressive stimulative fiscal policy to counter the shock. Both will likely result in a rapid buildup of public debt. However, the sustainability of such debt depends crucially on the type of economic disaster confronted.

In order to illustrate this point we turn to [Figure 3d](#). This figure shows the response of the real natural rate minus the growth rate of real GDP per capita, call it $r - g$, using natural rate data only for the U.K. and for real GDP per capita growth rates for England, as above. If this difference becomes more negative, it becomes easier to sustain higher levels of debt. Conversely, when this difference becomes more positive, debt sustainability becomes

harder. Once again, we see that pandemics and wars have very different consequences in this respect as well.

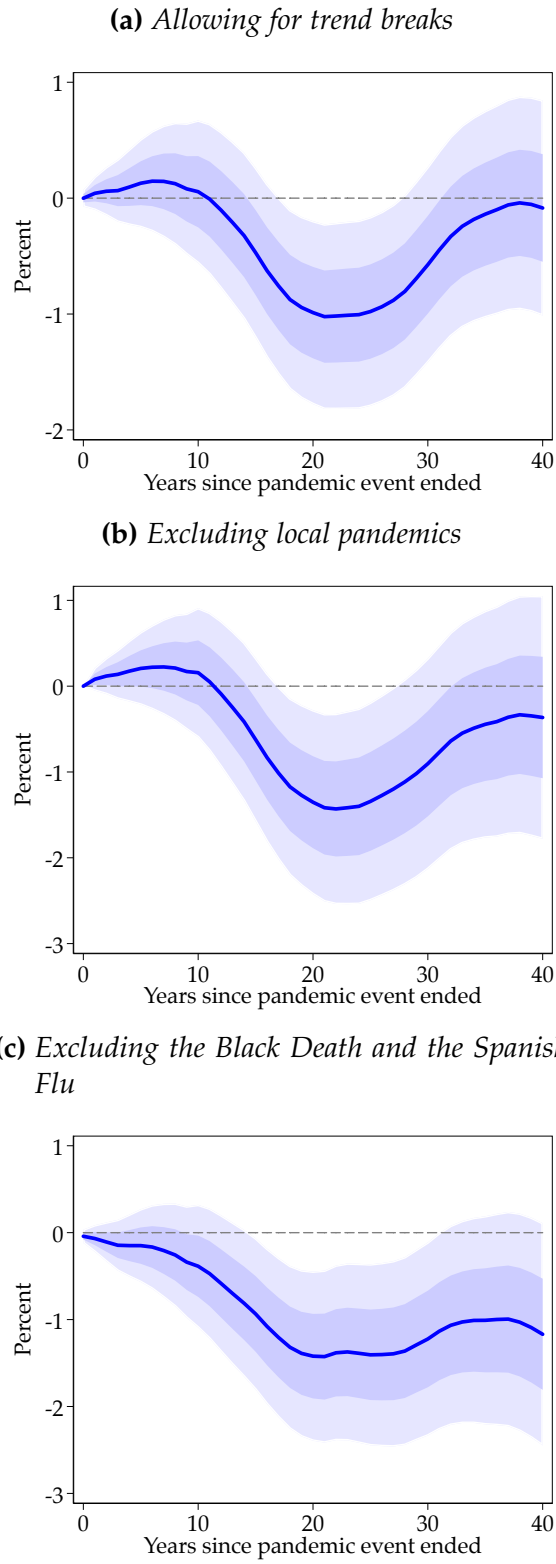
Figure 3d shows that in the aftermath of pandemics, $r - g$ becomes slightly negative, by about 50 basis points around the 20 year mark, before returning back to equilibrium. In contrast, wars result in a boost of about 100 basis points, peaking around the 30 year mark. Panel (d) in Table 1 shows evaluates the differences between these two paths every decade. Though the statistical evidence is not as strong as in previous experiments, the economic differences are patently clear.

6. ROBUSTNESS

Robustness to possible major trend breaks. Schmelzing (2020) proposes three historical dates at which the trend for the real interest rate could have changed. These are the “post-Bullion famine” period following the end of the global monetary contraction (1494), the “North-Weingast” institutional revolution that led to emergence of credible public debt mechanisms in Britain (1694), and the “post-Napoleonic” founding of the modern international state system (1820). Although the state-space model from Equation 3 is sufficiently flexible, to confirm that our results are not affected by such shifts, we add controls for time trends starting at these historical dates. Figure 4a plots the new estimated response of the European real natural rate following a pandemic event. As the figure shows, our baseline result is largely unchanged. The trough is similar and happens at about the same time.

Robustness when excluding small and localized pandemics. The plagues of London, Sevilla, and Marseille in Figure 1 were largely localized, not affecting the wider European region very much. Some other plagues had relatively small death totals. One may therefore be concerned that these relatively smaller events could attenuate the response of the aggregate European real natural rate. Figure 4b recalculates the response of the real rate

Figure 4: *Response of real natural rate: robustness*



Notes: Response calculated using Equation 2. Shaded areas are 1 and 2 s.e. bands around response estimates. See text.

when omitting the five smallest pandemics in [Figure 1](#). The peak response in the real natural rate is slightly larger when estimated with the remaining “global” pandemics, as one would expect. These smaller pandemics exerted a smaller influence on the real rate.

Robustness when excluding the Black Death and the Spanish Flu events. Finally, in [Figure 4](#) we consider removing from the sample the two largest events in history, the Black Death and the Spanish Flu. The resulting response is presented in [Figure 4c](#). The Black Death stands out for having decimated the European population, thus having an outsize weight in explaining the decline of the natural rate. Although the Spanish Flu was not quite as deadly as the Black Death, it came on the heels of WW1 and was followed ten years later by the Great Depression. Naturally one might worry that the decline in the natural interest rate was due to the Great Depression and not the pandemic. However, as [Figure 4c](#) shows, removing these two influential events from the sample does not affect our main result materially.

7. CONCLUSIONS

Summing up our findings, the great historical pandemics of the last millennium have typically been associated with subsequent low returns to assets, as far as the limited data allow us to conclude. These responses are huge. Smaller responses are found in real wages, but still statistically significant, and consistent with the baseline neoclassical model.

Measured by deviations in a benchmark economic statistic, the real natural rate of interest, these responses indicate that pandemics are followed by sustained periods—over multiple decades—with depressed investment opportunities, possibly due to excess capital per unit of surviving labor, and/or heightened desires to save, possibly due to an increase in precautionary saving or a rebuilding of depleted wealth. Either way, if the trends play out similarly in the wake of COVID-19 then the global economic trajectory will be very different than was expected only a few months ago.

Should we expect declines of 1.5%–2% in the real natural rate, however? There may be at least three factors that could possibly attenuate the decline of the natural rate predicted by our analysis, but their presence and magnitude is uncertain and unknowable until therapies to fight COVID-19 are more developed. First, the death toll of COVID-19 relative to the total population might be smaller than in the worst pandemics of the past, but we cannot know for sure at this point. Second, COVID-19 primarily affects the elderly, who are no longer in the labor force and tend to save relatively more than the young, so the demographic channels could be altered, although the recent pick up in infections is now affecting younger individuals. Third, aggressive counter-pandemic fiscal expansion will boost public debt further, reducing the national savings rate and this might put upward pressure on the natural rate, even though our analysis suggests that this expansion of public debt should be easier to sustain in the long-run.

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A. APPENDIX: ESTIMATION OF TRENDS

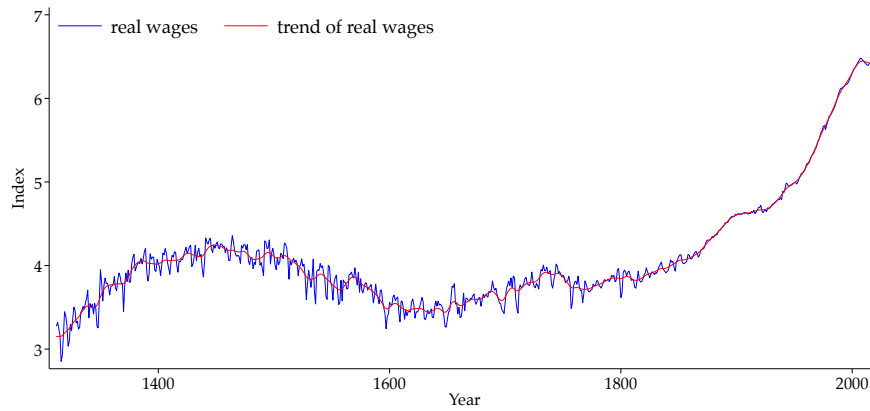
We use the following state-space system, presented in Equation 3, to estimate trends for log real wages and log real GDP per capita. Let y_t be the variable whose potential trend y_t^* is to be estimated, where g_t is the growth rate of the potential trend.

$$\begin{aligned} \hat{y}_t &= y_t - y_t^*, && \text{(gap)} \\ y_t^* &= y_{t-1}^* + g_{t-1} + u_{y,t}^*, && \text{(potential)} \\ g_t &= g_{t-1} + u_{g,t}. && \text{(potential growth)} \end{aligned} \quad (5)$$

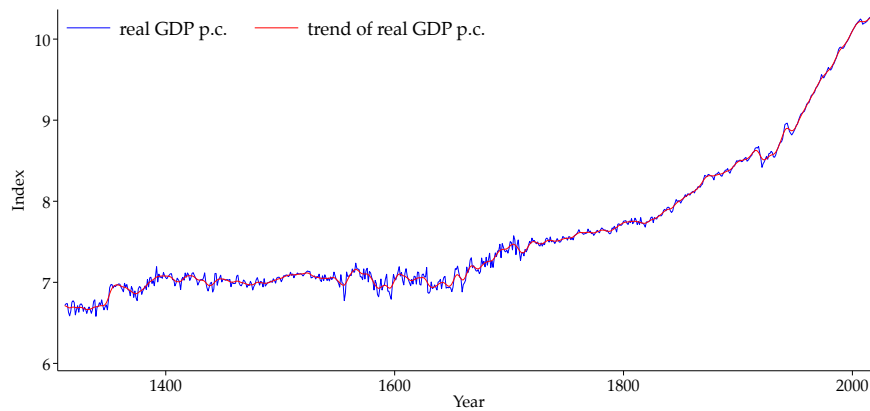
Figure A1 shows the estimated trend for real wages and real GDP per capita.

Figure A1: *Estimated trends of log real wages and log real GDP per capita in the UK*

(a) *Real wages in Great Britain*



(b) *Real GDP per capita in England*



Notes: Raw data for real wages in Great Britain and real GDP per capita of England are obtained from Thomas and Dimsdale (2017). Trends estimated with a state-space system following Laubach and Williams (2003), presented in Equation 5. Real GDP per capita series is based on market prices and is quoted in “£mn, chained volume measure, 2013 prices.” Real wages are in the form of an index where 1900 real wages are normalized to 100. In the graphs presented, these series are plotted on log-scale.