

How Much Will Global Warming Cool Global Growth?

Ishan Nath

Valerie Ramey

Pete Klenow

SF Federal Reserve

Hoover Institution & NBER

Stanford & NBER

November 9-10, 2023

Yale Climate, Environment, and Economic Growth Conference

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Motivation: Wide Divergence in Climate-GDP Projections

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- Prevailing literature: modest impacts
 - Barrage and Nordhaus (2023): 1.6% of global GDP from 3°C warming by 2100
 - IPCC (2014): 0.2 to 2% from 2°C of warming

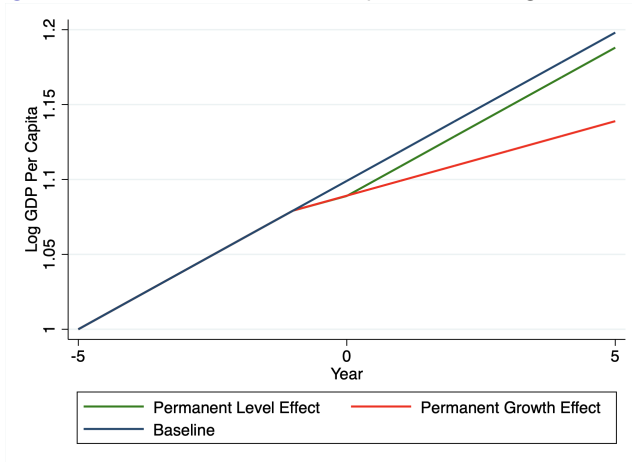
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 - IPCC (2014): 0.2 to 2% from 2°C of warming
- Prominent exception: very large effects
 - Burke, Hsiang, Miguel (2015): 23% of global GDP by 2100
 - Climate change reduces incomes by $> 80\%$ in 50% of countries

Motivation: Why impact estimates diverge

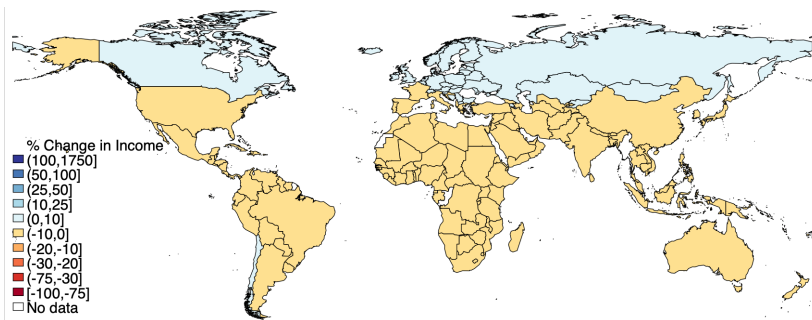
Does a permanent \uparrow in temperature affect long-run growth or levels?

Figure: Effects of Permanent Temperature Change in Year 0



Climate change impacts: permanent level effects

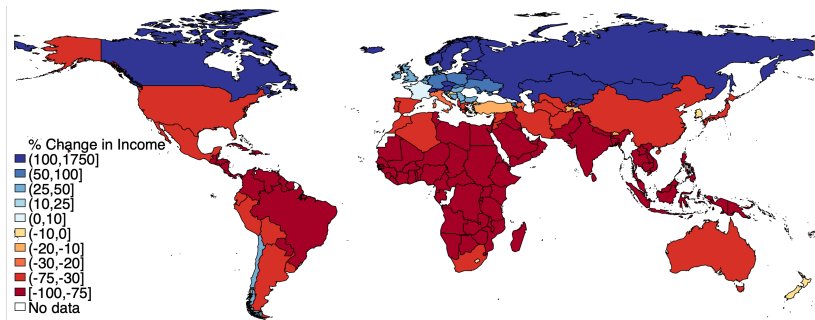
Figure: Percent Change in Annual Income in 2099



Source: Example Using Permanent Level Effect Estimates

Climate change impacts: permanent growth effects

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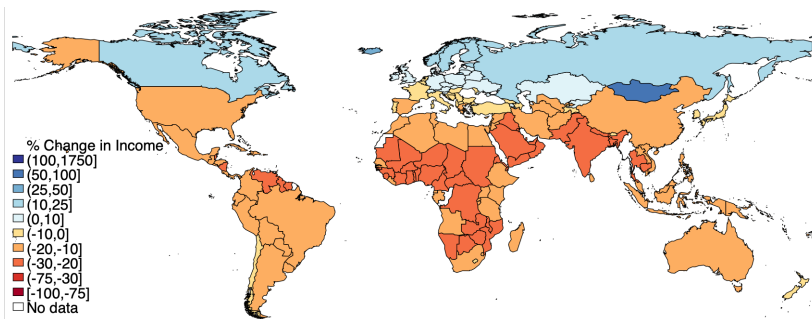
Source: Burke, Hsiang, & Miguel (2015)

This Paper

- Theory and evidence for why country growth rates should not permanently diverge
- Dynamic panel estimates of the temperature-GDP relationship
- Projections of future climate change impacts based on empirical persistence of temperature effects

Results Preview: Our Projections

Figure: Percent Change in Annual Income in 2009



Key caveat: not a comprehensive welfare estimate

- Non-market damages (e.g. mortality, civil conflict)
 - e.g. Hsiang, Burke, & Miguel (2013), Carleton et al. (2022)
- Non-temperature effects (e.g. hurricanes, coastal flooding)
 - e.g. Desmet et al. (2021), Balboni (2021), Fried (2022)
- Tipping points
 - e.g. Lemoine & Traeger (2016), Dietz et al. (2021)
- Uncertainty and risk aversion
 - e.g. Weitzman (2009), Traeger (2014), Barnett, Brock, & Hansen (2020), Lemoine (2021), Nath et al. (2022)
- Adaptation
 - e.g. Moscona & Sastry (2021), Cruz & Rossi-Hansberg (2021)

Related Literature

- Panel and time-series estimates of temperature and output
 - Country-level data: Dell, Jones, & Olken (2012); Burke, Hsiang, & Miguel (2015); Acevedo et al. (2020); Berg, Curtis, & Mark (2021); Newell, Prest, & Sexton (2021); Bastien-Olvera, Granella, & Moore (2022)
 - Subnational data: Colacito, Hoffman, & Phan (2019); Burke & Tanutama (2019)
- Empirical climate-GDP projections informed by growth models
 - Kahn et al. (2019); Kalkuhl & Wenz (2020); Casey, Fried, & Goode (2022)

Outline

- 1 Introduction
- 2 Are Country Growth Rates Connected?
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A Stylized Model of Global Growth

- Productivity in each country draws on domestic and international technologies, with varying levels of domestic efficiency μ_i :

$$Q_{it} \propto \mu_{it} \cdot (Q_{it-1})^{1-\omega} (Q_{t-1}^*)^{\omega}.$$

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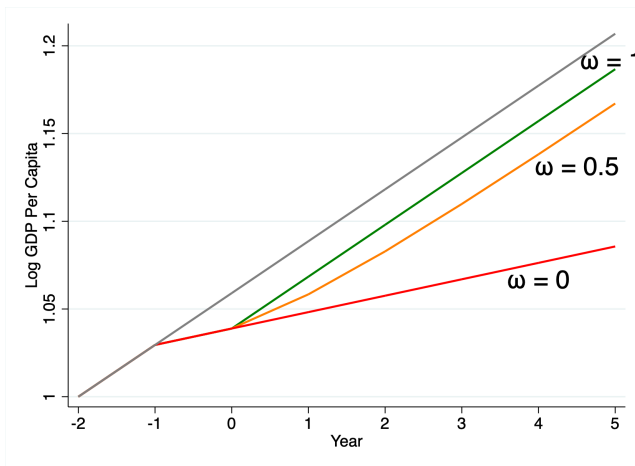
$$Q_{t+1}^* \propto \mu_t^* \cdot Q_t^*.$$

- Each country's per capita income is proportional to its productivity:

$$Y_{it}/L_{it} \propto M_{it}^{\frac{1}{\sigma-1}} \cdot Q_{it}.$$

Comparative Statics - Permanent Shock to μ_i

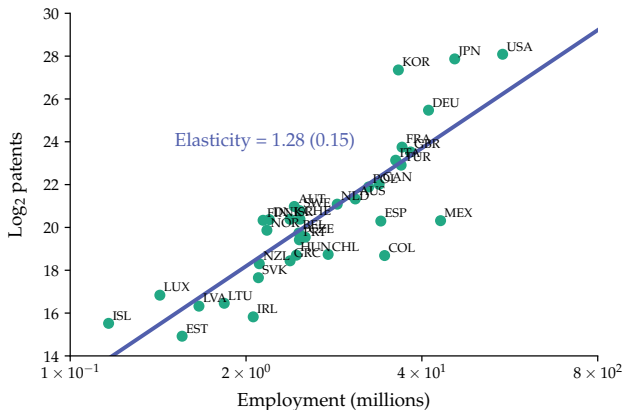
Figure: Effects of Permanent Temperature Shock Starting in Year 0



A three-part case for global growth spillovers ($0 < \omega < 1$)

1. Bigger countries innovate more ...

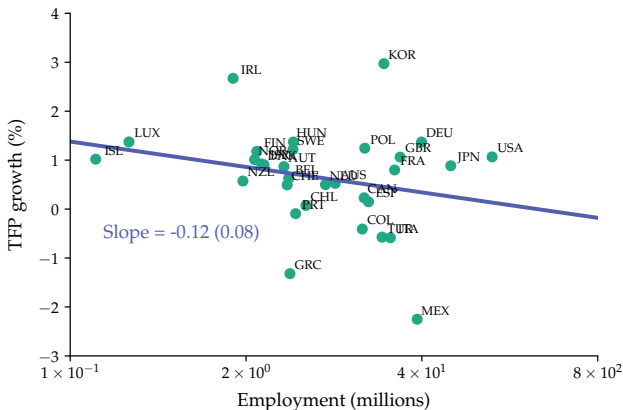
Figure: U.S. Patents and Employment in the Country of Origin in 2019



More people → more researchers → more patents

1. Bigger countries innovate more ... but don't grow faster

Figure: TFP Growth and Employment in OECD Countries, 1980-2019



More people \rightarrow more researchers \rightarrow more patents \nrightarrow more growth

2. Country differences persist in levels, but not growth

- We regress country TFP levels and growth on country and year FE:

$$y_{it} = \delta_i + \gamma_t + \epsilon_{it}$$

- We test: $H_0 : \delta_i \neq 0$ for each i

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- For TFP levels, we reject equality for 50 - 70% of countries.

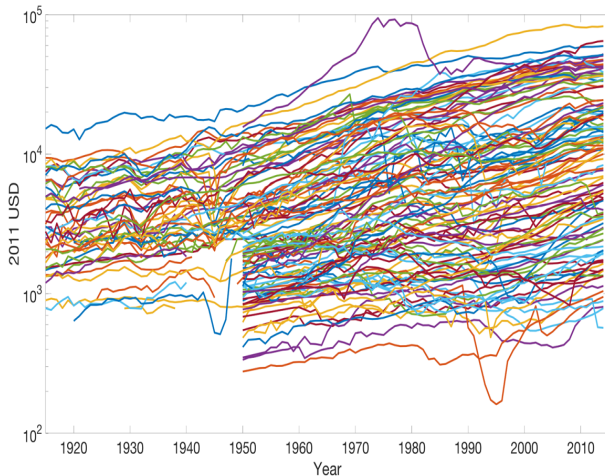
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- We test: $H_0 : \delta_i \neq 0$ for each i
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- For TFP growth rates, we reject equality for only 2 - 9% of countries

Log GDP per capita, 1915-2014, 112 countries



Sources: Penn World Tables; Müller, Stock, and Watson, (2022)

3. Frontier country technology predicts global growth

- Estimate model-implied equation of motion for technology:

$$\ln(TFP)_{it} = (1 - \omega) \ln(TFP)_{i,t-1} + \omega \ln(TFP)_{t-1}^{OECD} + \delta_i + \epsilon_{it}$$

- Estimates consistent with $\omega \approx 0.07$ - modest international spillovers

A three part case for global growth spillovers ($0 < \omega < 1$)

- 1 Rich countries grow at similar rates despite innovation differences
- 2 Country level differences persist, but growth differences do not
- 3 Frontier country technology predicts global growth

Literature on globally-interconnected growth

- Technology flows across countries (patents, equipment, hybrid seeds)
 - Eaton and Kortum (1999 IER, 2001 EER), Gollin et al. (2021 JPE)
- Growth differences are transitory
 - Klenow and Rodriguez-Clare (2005), Pritchett and Summers (2014)
- Countries can converge toward, but not surpass, frontier
 - Parente and Prescott (2002, 2005)
- Global growth models:
 - Grossman & Helpman (1991), Acemoglu (2008), Akcigit, Ates, & Impulitti (2018), Buera & Oberfield (2020 ECMA) Cai, Li, & Santacreu (2022 AEJ-Macro), Hsieh, Klenow, & Nath (2021), Hsieh, Klenow, & Shimizu (2022)

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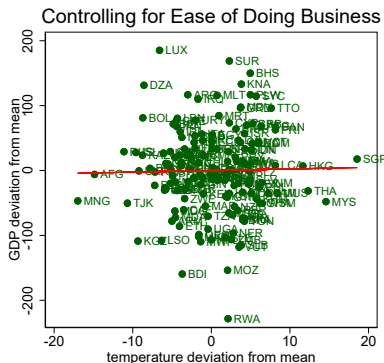
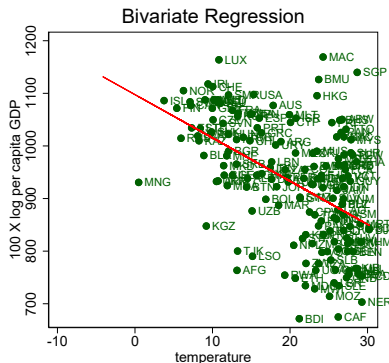
How to Estimate the Effects of Temperature on GDP

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- Cross-country regressions.
 - Advantage: Captures long-run effects, adaptation.
 - Disadvantage: historical influences of temperature on institutions
- Time-series regressions
 - Advantage: Directly measures correlation between temperature and GDP over time
 - Disadvantage: other trends in GDP, most temperature variation is temporary

Cross-Country Regression GDP on Temperature for 2015



Variable	Bivariate regression	Controlling for Business Ease
Temperature	-8.22 (1.07)	0.24 (0.97)
Business ease		7.65 (0.57)
R^2	0.26	0.67

Our Empirical Strategy: Use Panel Data Variation

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 - Effect of the shocks may depend on **average country temperature**
- Our Approach: State-dependent Local Projections (Jorda, 2005)
 - Estimate longer-horizon impulse responses

Data

- Global Meteorological Forcing Temperature dataset
 - Global grid at 0.25° by 0.25° resolution
 - Population-weighted to the country level
- World Development Indicators for GDP Per Capita

Constructing Temperature Shocks

- Estimating a temperature shock τ_{it} :

$$T_{it} = \sum_{j=1}^p (\gamma_j T_{i,t-j} + \theta_j T_{i,t-j} \cdot \overline{T_i}) + \mu_i + \mu_t + \tau_{it} \quad (1)$$

- Shock is the residual of an autoregressive model of temperature T .
- Lag coefficients vary by country mean temperature, $\overline{T_i}$.
- μ_i is country fixed effects.
- μ_t is year fixed effects (included in some specifications).
- τ_{it} is the estimated temperature shock.

Impulse Response Estimation

- Temperature response local projections:

$$T_{i,t+h} = \alpha_0^h \tau_{it} + \alpha_1^h \tau_{it} \cdot \overline{T_i} + X_{it} + \zeta_{it}, \quad h = 1, \dots, H.$$

$$\text{where } X_{it} = \{T_{i,t-j}, T_{i,t-j} \cdot \overline{T_i}\}_{j=1}^p, \quad \mu_i, \mu_t.$$

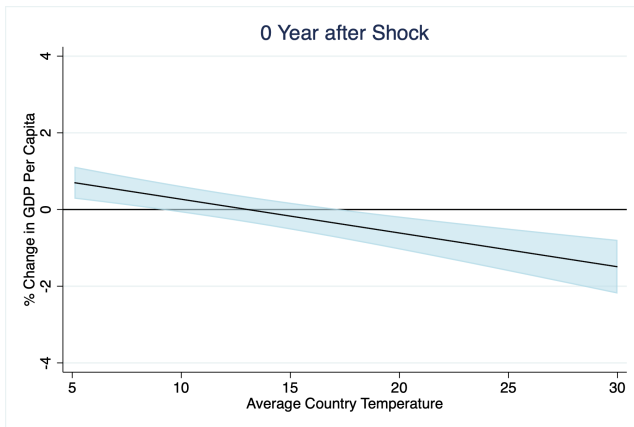
- GDP response local projections:

$$y_{i,t+h} - y_{i,t-1} = \beta_0^h \tau_{it} + \beta_1^h \tau_{it} \cdot \overline{T_i} + Z_{it} + \epsilon_{it}, \quad h = 0, \dots, H.$$

$$\text{where } Z_{it} = \{T_{i,t-j}, T_{i,t-j} \cdot \overline{T_i}, \Delta y_{i,t-j}\}_{j=1}^p, \quad \mu_i, \mu_t.$$

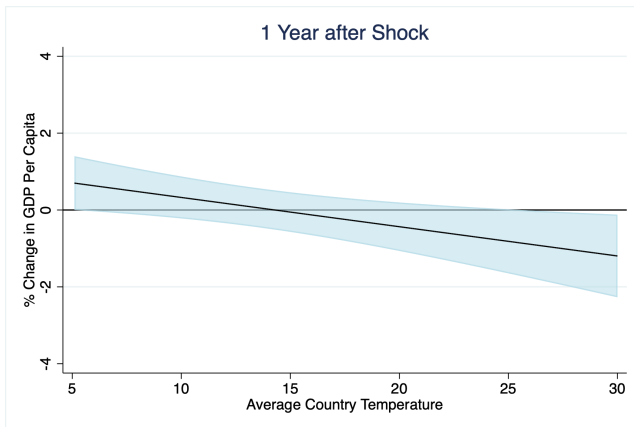
Effect of a Temperature Shock on GDP

Figure: Impact of a 1°C Temperature Shock on GDP
By Long-Run Average Temperature



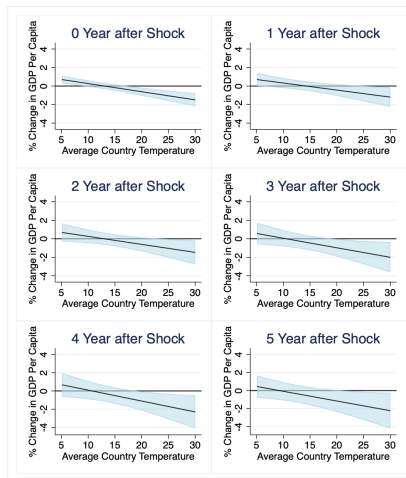
Effects on GDP Persist After Initial Shock

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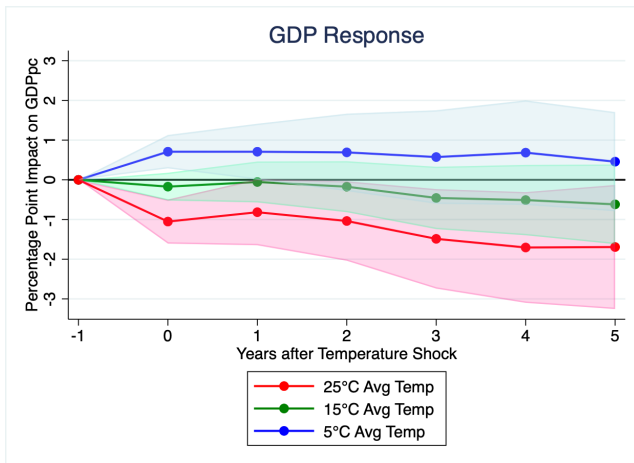
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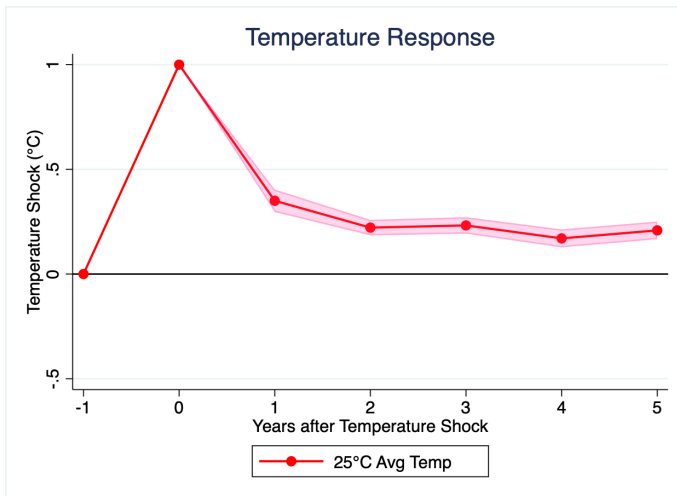
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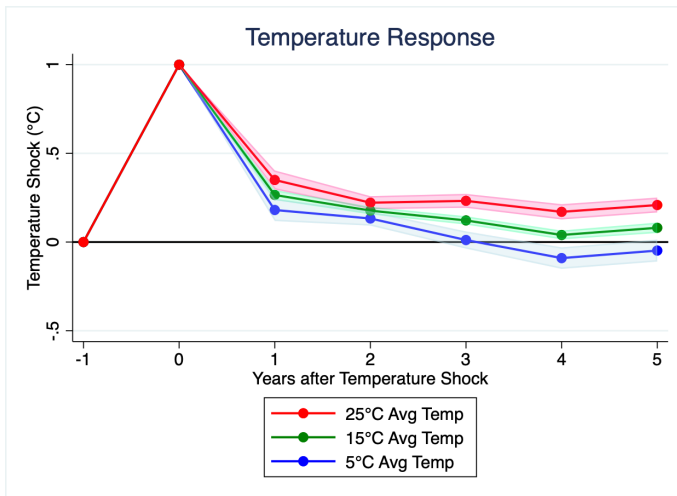
Temperature Response is also Persistent

Figure: Persistence of Temperature Response to a 1°C Shock In Hot Countries



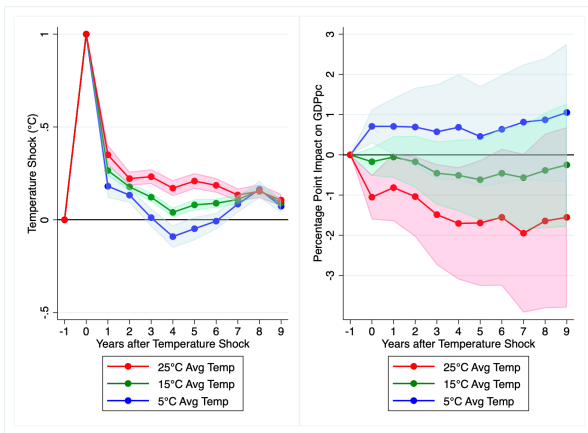
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Both Temperature and GDP Effects of a Shock Persist

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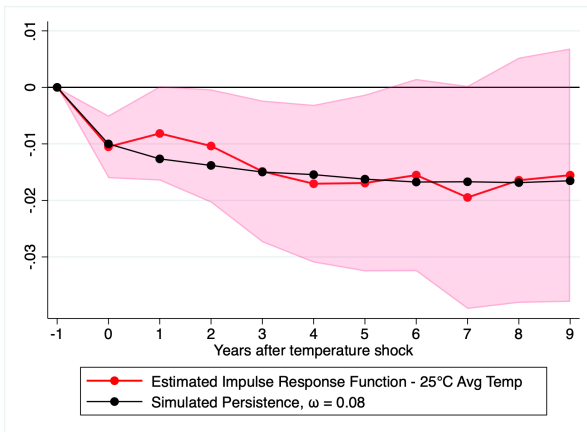


Using Empirical IRFs to Back Out ω

- We construct a simulation of a temperature shock with persistence to compare to the empirical IRF
- Magnitude of 1°C shock to μ_{it} calibrated to match year 0 effect
- Calibrate path of temperature following the shock to match empirical temperature IRF
 - Search for ω that minimizes sum of squared errors between model and empirical IRF

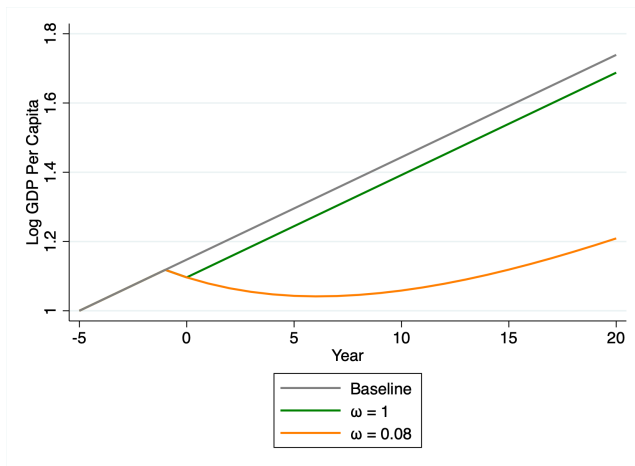
Comparing Empirical and Model IRFs

Figure: Simulated and Empirical Effects of Identical *Persistent* Temperature Shock in Year 0
 $\omega = 0.08$



Implications of $\omega = 0.08$

Figure: Simulated Effects of Permanent Temperature Shock Starting in Year 0



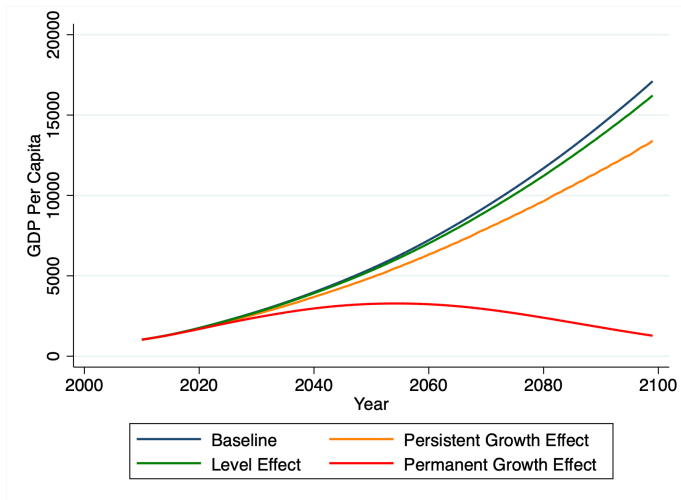
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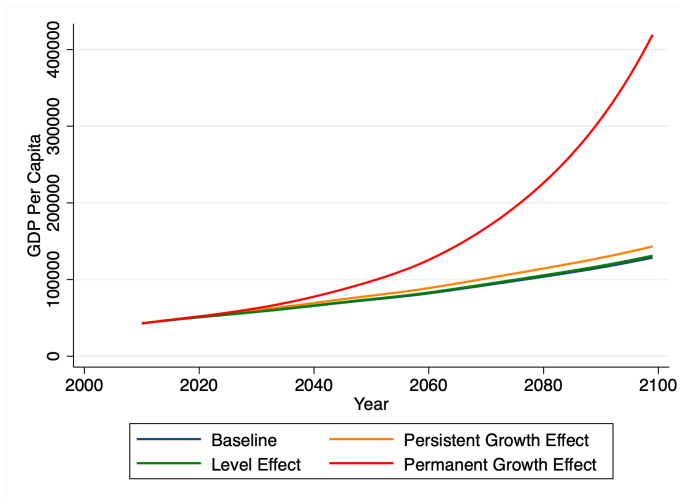
Projection Approach

- Use 10-year *cumulative response ratio* (GDP effect / temperature effect) to project long-run impact of temperature change
- Cumulative response ratio varies by initial temperature
- Temperature projections come from BHM (2015 Nature)
 - Average over many climate models in “baseline” emissions scenario
 - ΔT varies by country, slightly under 4°C for the world

Projection Results: India

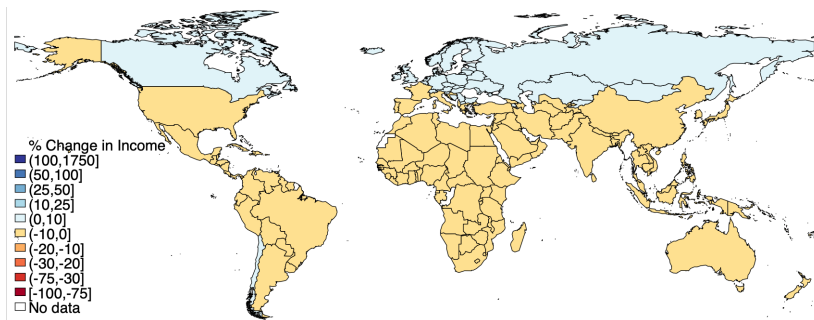


Projection Results: Sweden



Climate Change Projections - Permanent Level Effects

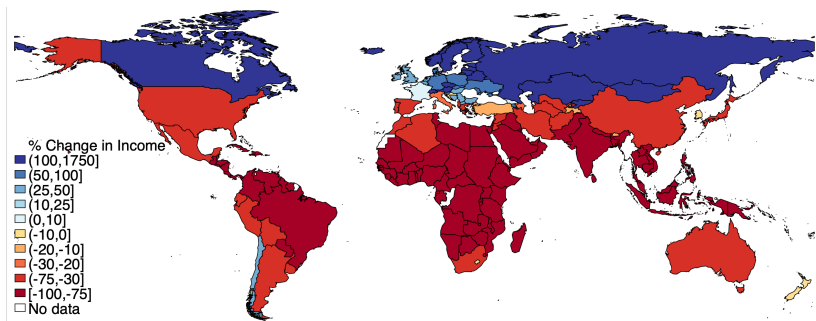
Figure: Impact of Climate Change on Annual Income in 2099



Source: Example Using Our Estimated Contemporaneous Effects Only

Climate Change Projections - Permanent Growth Effects

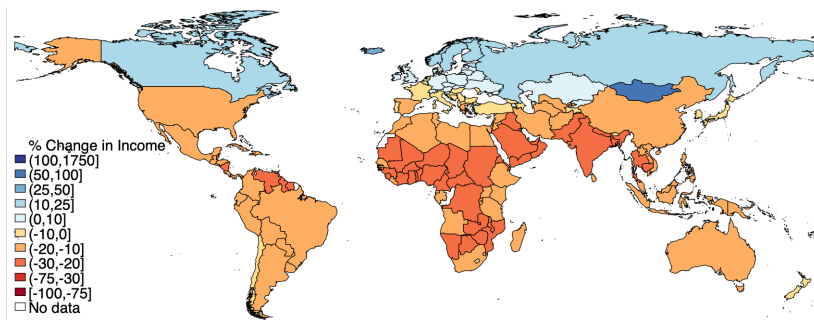
Figure: Impact of Climate Change on Annual Income in 2099



Source: Burke, Hsiang, & Miguel (2015)

Climate Change Projections - Our Estimates

Figure: Impact of Climate Change on Annual Income in 2099



Source: Our estimates using accumulated level effect from 10 lags

Projection Summary

Table: Projected Effects of Unabated Global Warming on 2099 Income
Year Fixed Effect Specification

Region	Persistent Growth Effects	Level Effects	Permanent Growth Effects
Global GDP	-11.5	-2.2	-26.6
Global Population Average	-16.4	-3.6	-58.7
Sub-Saharan Africa	-20.6	-4.8	-86.1
Middle East & North Africa	-20.1	-4.3	-82.5
Asia	-18.0	-4.0	-73.3
South & Central America	-16.1	-3.3	-74.6
North America	-9.6	-1.4	-20.0
Europe	0.6	0.4	96.6

Projection Summary

Table: Projected Effects of Unabated Global Warming on 2099 Income
US TFP Control Specification

Region	Persistent Growth Effects	Level Effects	Permanent Growth Effects
Global GDP	-6.8	-1.9	-26.6
Global Population Average	-10.0	-3.1	-58.7
Sub-Saharan Africa	-13.0	-4.2	-86.1
Middle East & North Africa	-12.1	-3.7	-82.5
Asia	-11.0	-3.4	-73.3
South & Central America	-9.5	-2.8	-74.6
North America	-4.8	-1.2	-20.0
Europe	0.2	0.4	96.6

Conclusion

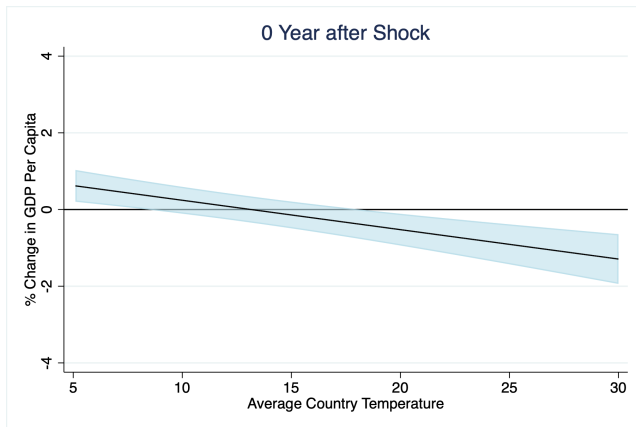
- Model & evidence suggest growth is tied together across countries
 - Temperature unlikely to have permanent country growth effects
 - Trending temperatures can still have global growth effects
- Dynamic estimates show persistent effects of temperature on GDP
 - Moderate persistence of temperature itself
- Projections suggest warming reduces global income 6-12% by 2100
 - ~ 3-5x larger than permanent level effects
 - ~ 3-4x smaller than permanent growth effects
 - Country-specific effects differ even more dramatically

Appendix

EXTRA SLIDES

Effect of a Temperature Shock on GDP

Figure: Impact of a 1°C Temperature Shock on GDP
By Long-Run Average Temperature - US TFP Control Instead of Year FE

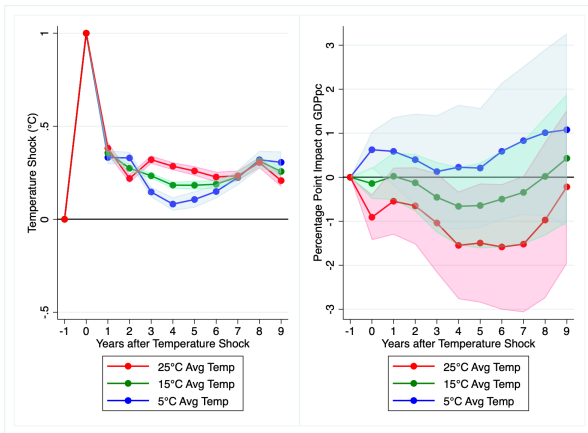


Controls for contemporaneous US TFP instead of year FE

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Both Temperature and GDP Effects of a Shock Persist

Figure: Persistent Effects of a 1°C Temperature Shock
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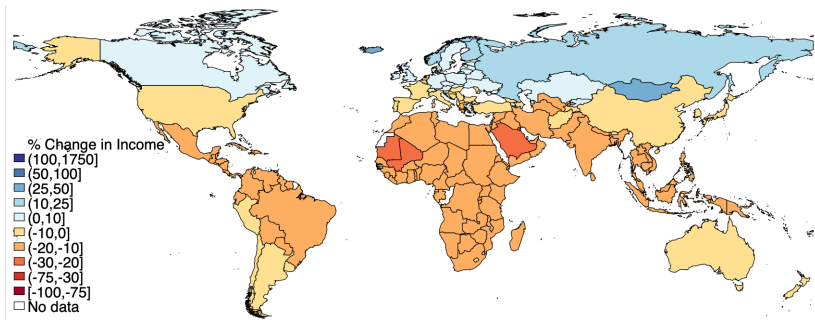


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Climate Change Impact Comparison

Figure: Difference in 2099 Climate Change KNR Estimates vs. Temporary Level Effects

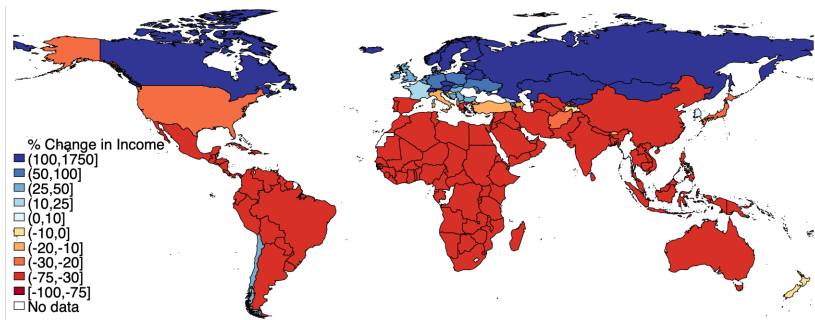


Source: Our dynamic estimates minus pure level effects only

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Climate Change Impact Comparison

Figure: Difference in 2099 Climate Change Permanent Growth Effects vs. KNR Estimates



Source: Burke-Hsiang-Miguel (2015) estimates minus our estimates

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