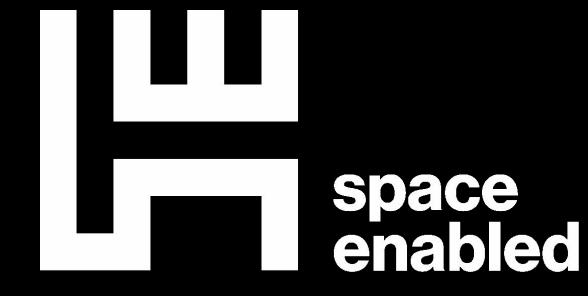
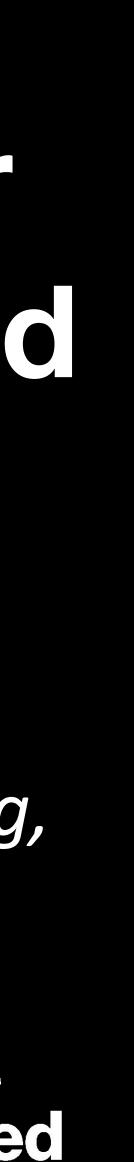
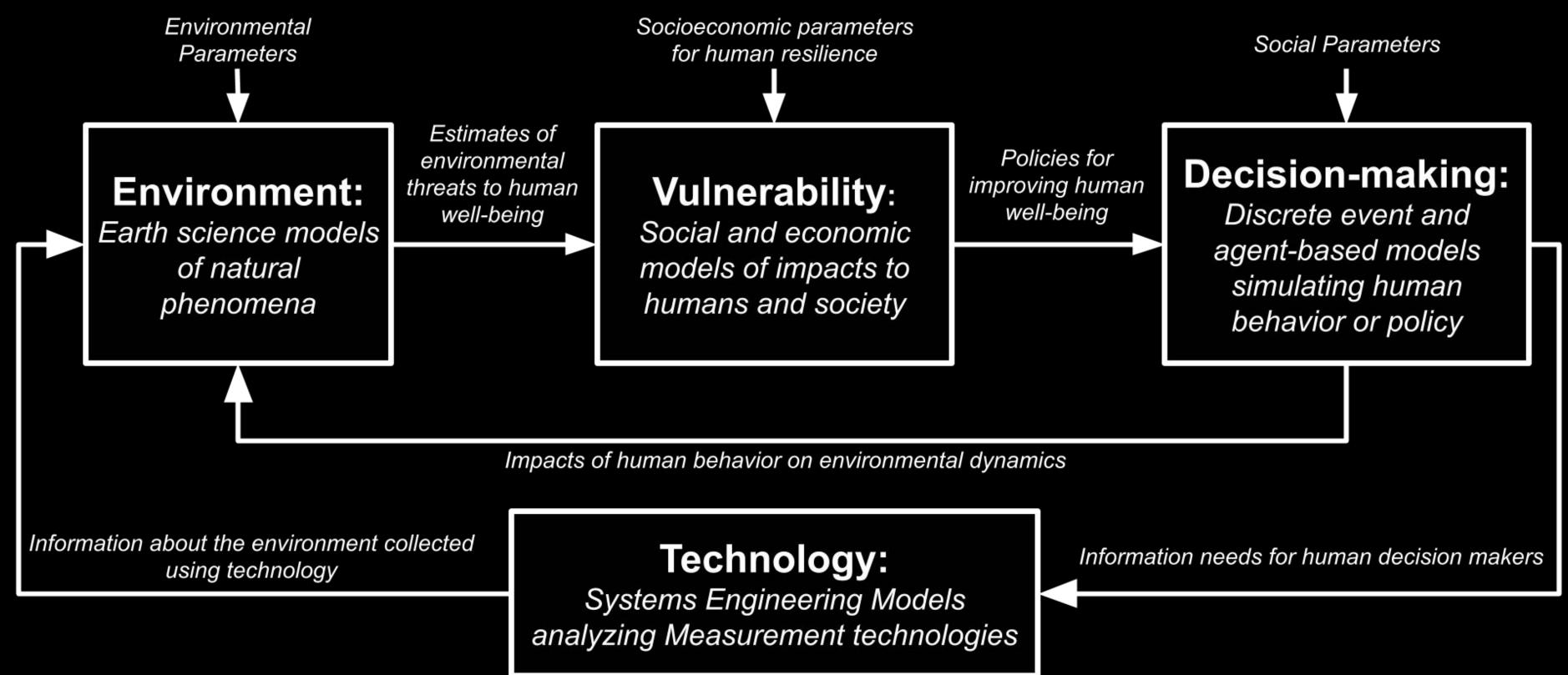
# Vida Decision Support System: An International, Collaborative Project for COVID-19 Management with Integrated Modeling

Jack Reid, Seamus Lombardo, David Lagomasino, Eric Ashcroft, Mary Bracho, Mohammad Jalali, Amanda Payton, Katlyn Turner, Maggie Zheng, Danielle Wood



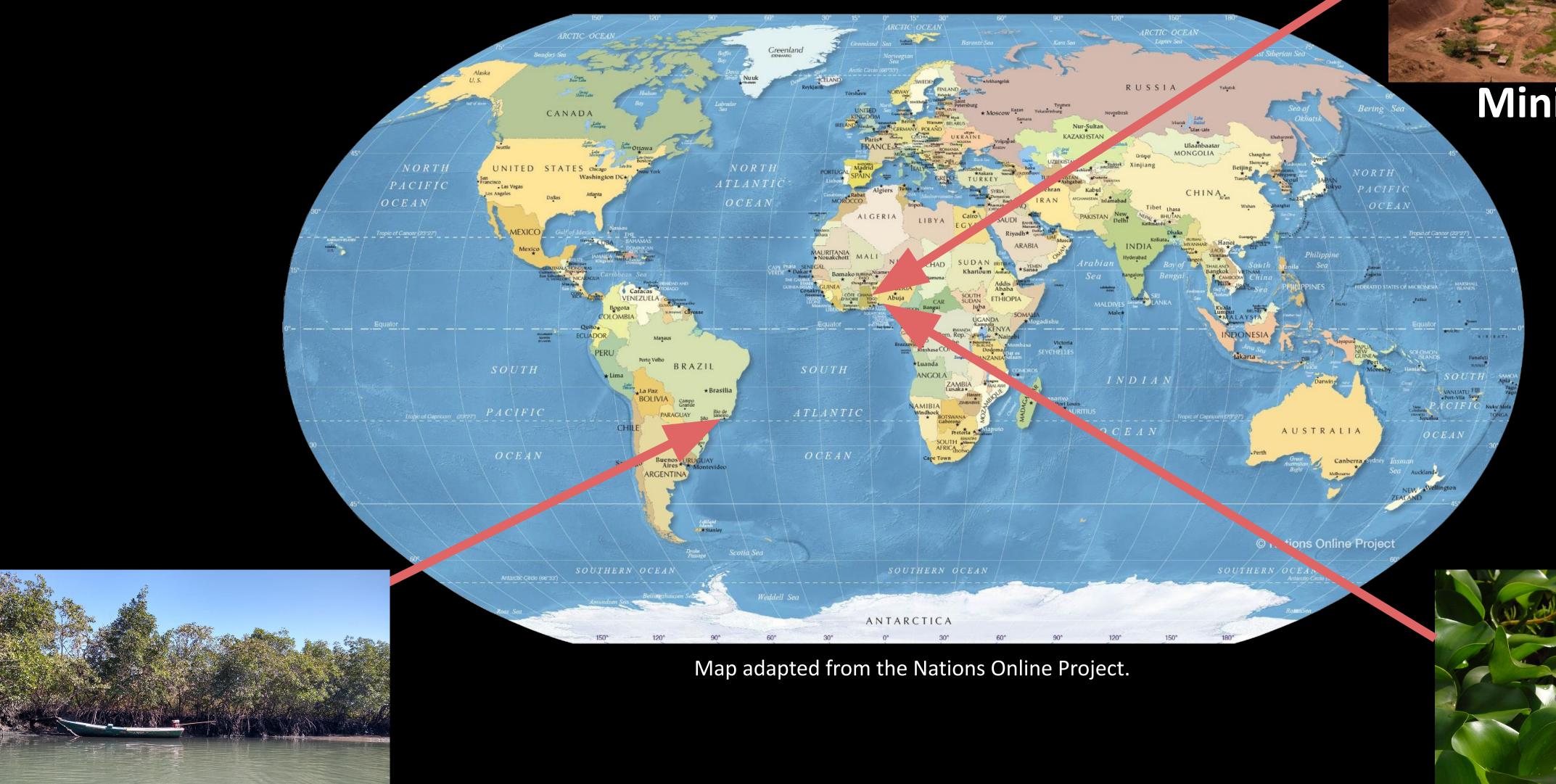


# **EVDT Framework**



- What is happening in the natural environment?
- How will humans be impacted by what is happening in the natural environment?
- What decisions are humans making in response to environmental factors and why?
- What technology system can be designed to provide high quality information that supports human decision making?

# Some Pre-Pandemic EVDT Applications



## Mangroves in Rio de Janeiro

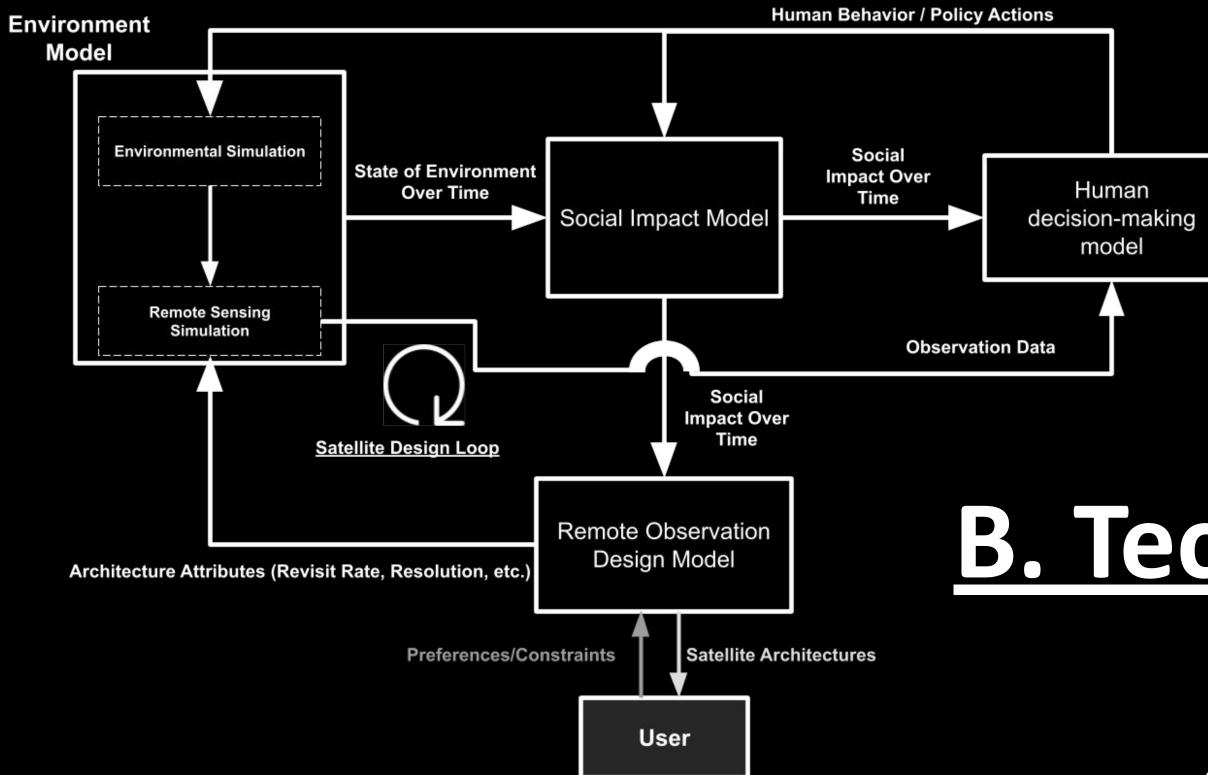


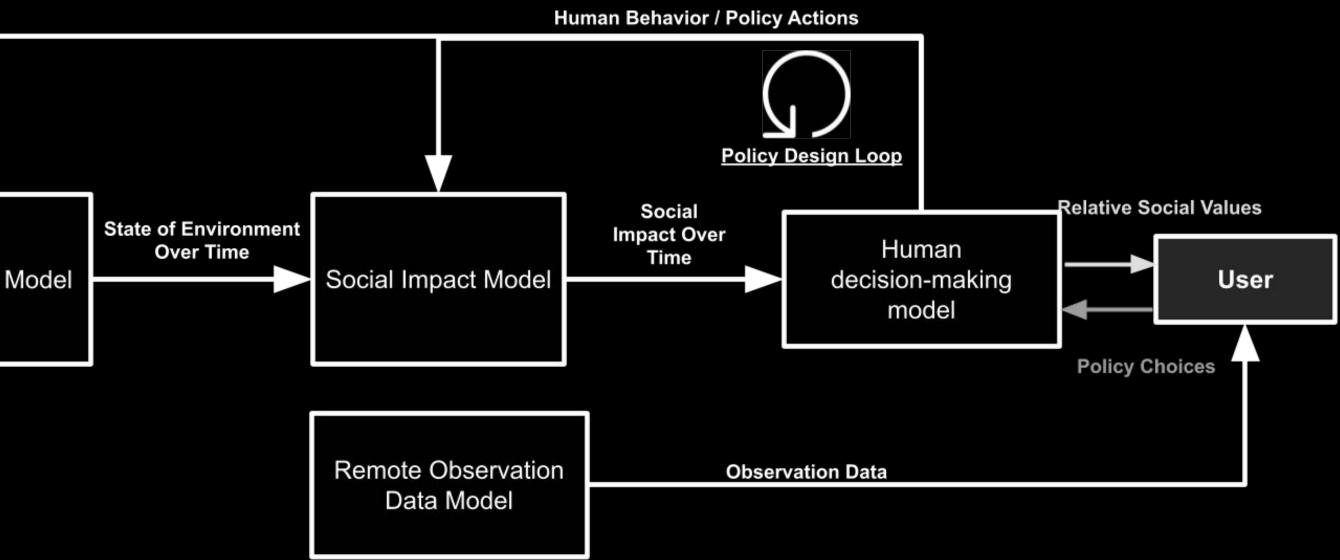




# A. Policy Design

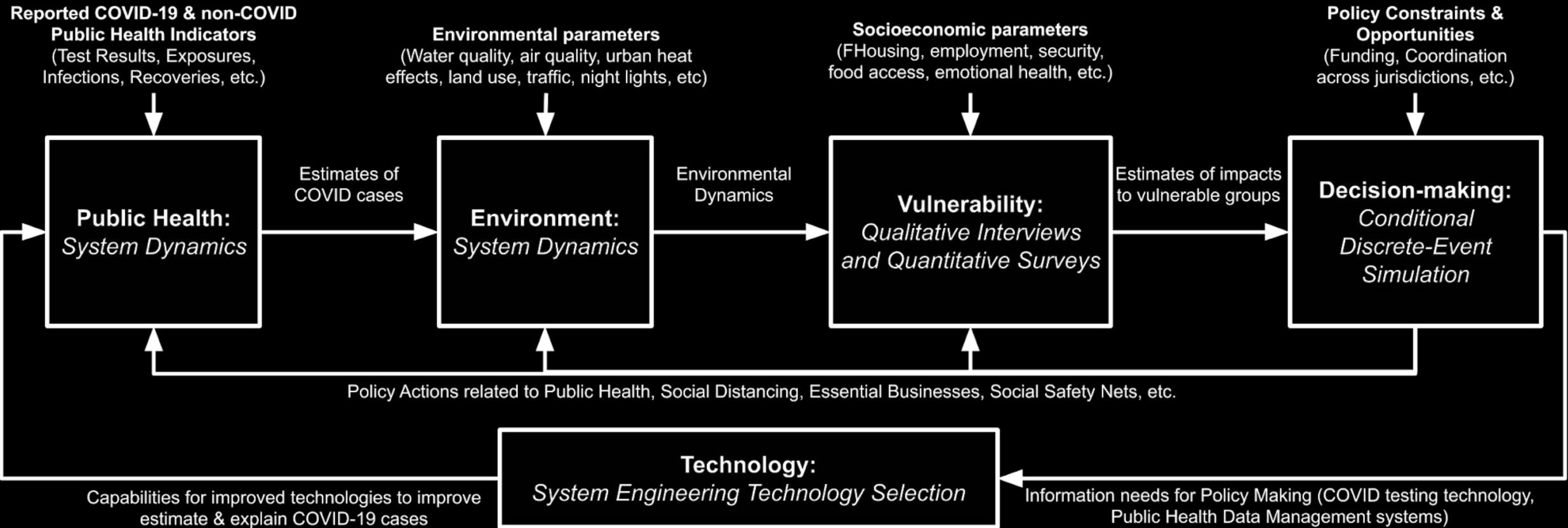
**Environment Model** 



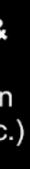


# B. Technology Design

# Vida Decision Support System



Public Health Data Management systems)



# Vida DSS International Network

SOUTh

NORTI PACIFIC

0 C E A 1

Querétaro, México

Región Metropolitana de Santiago, Chile

Boston, USA

## Java & Sulawesi, Indonesia



Río de Janeiro, Brasil

6

## Luanda, Angola

Map adapted from the Nations Online Project.



## Brasil



INSTITUTO PEREIRA PASSOS





Ministerio de Ciencia, Tecnología, Conocimiento e Innovación

Gobierno de Chile

Chile

## México

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## Indonesia





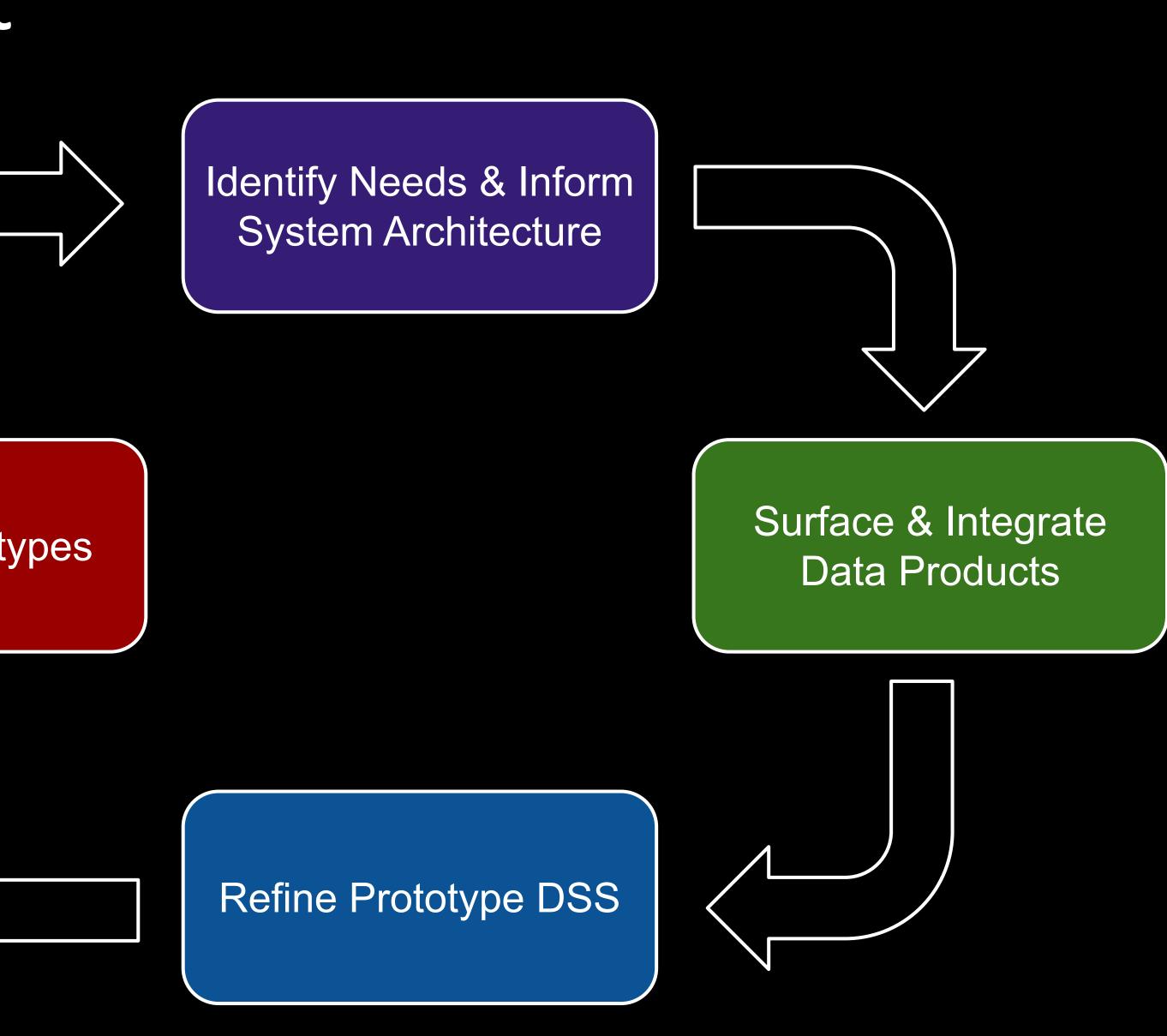


## Angola

# Stakeholder Involvement

- Weekly/Biweekly 1-on-1 meetings
- Monthly full network meetings
- Online collaboration
  - Data Repositories
  - Github  $\bigcirc$
  - Browser-based DSS

Evaluate Prototypes



### **Constraints or Opportunities:**

Inputs

- Limited resources of local leaders to address the pandemic Limited
- technical expertise of local leaders in modeling and data analysis

## System Stakeholders

- Primary stakeholders: US team and government, academic, and private collaborators directly working on Vida in each location.
- Secondary stakeholders: Other ٠ agencies and private entities who are taking actions related to the pandemic in each location
- Tertiary Stakeholders: Residents of each location . who are impacted by the virus and related policies

Allocate

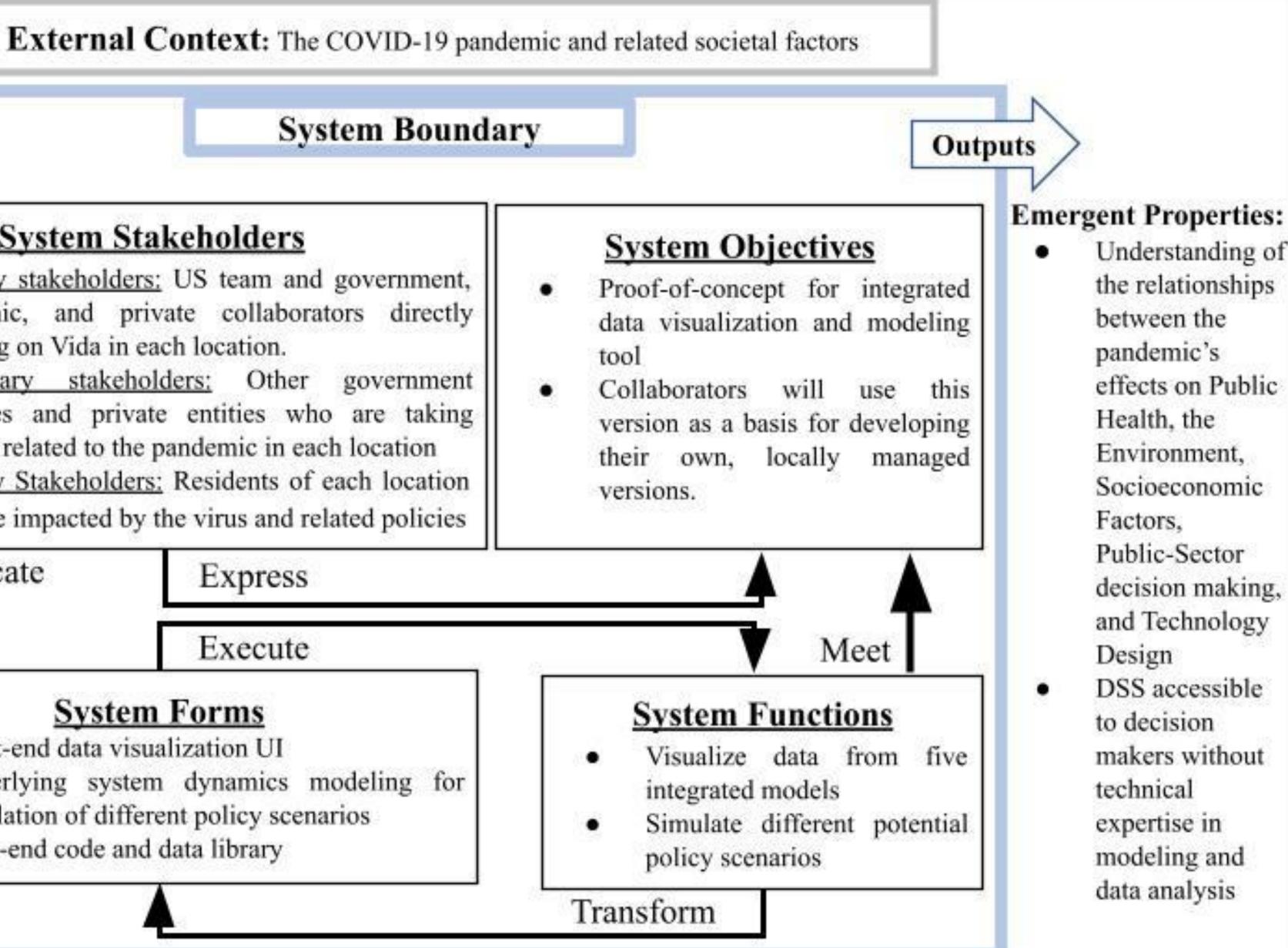
Express

Execute

### System Forms

- Front-end data visualization UI
- Underlying system dynamics modeling for ٠ simulation of different policy scenarios
- Back-end code and data library .



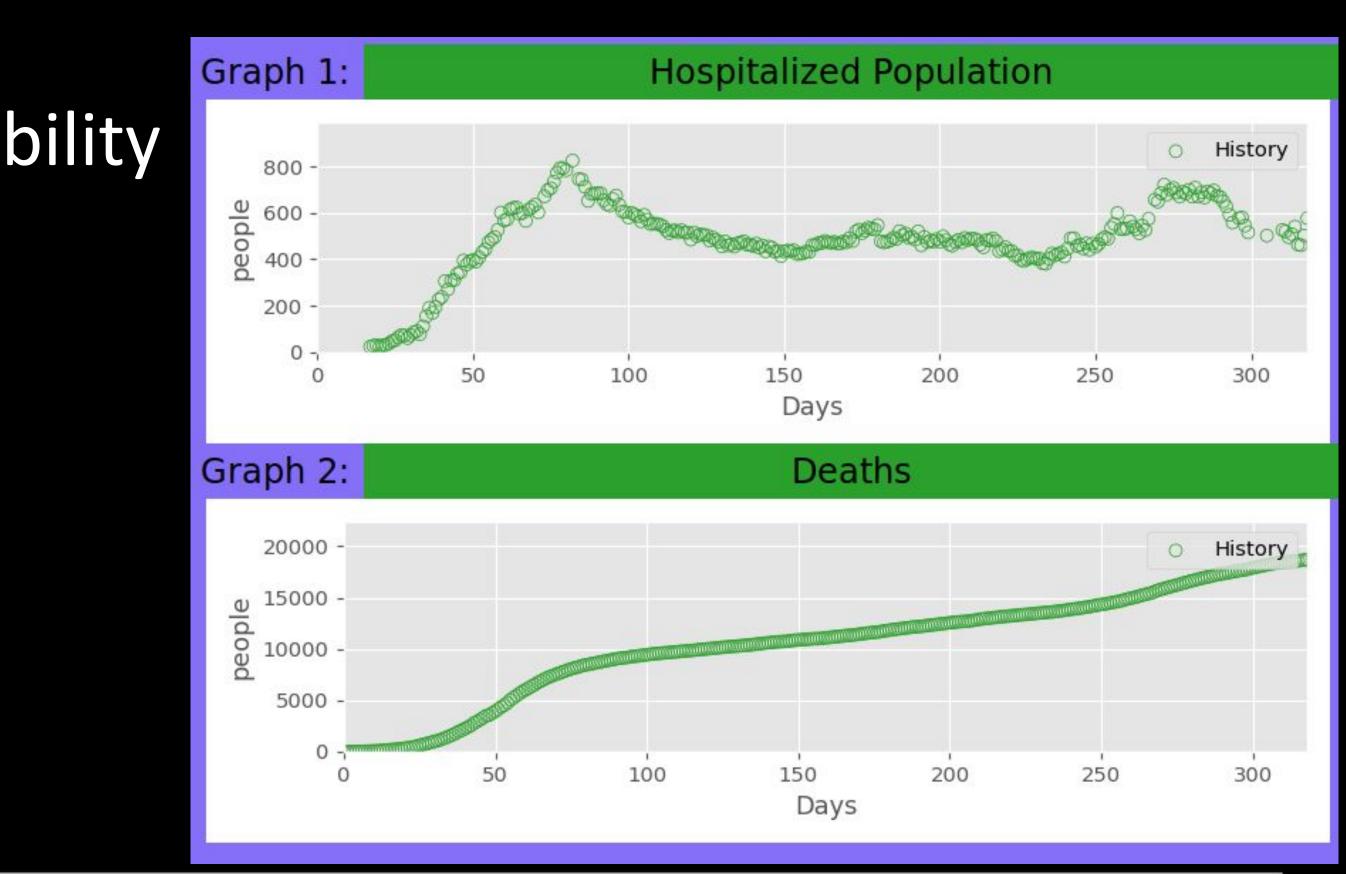




## Data & Methods: Public Health

- COVID-19 health data collected by local authorities
  - Daily infections, hospitalizations, deaths, and recoveries
  - Daily PCR tests
  - Hospital bed capacity and availability
  - Ventilator use and availability
  - Vaccination rates

## y local authorities , deaths, and recoveries



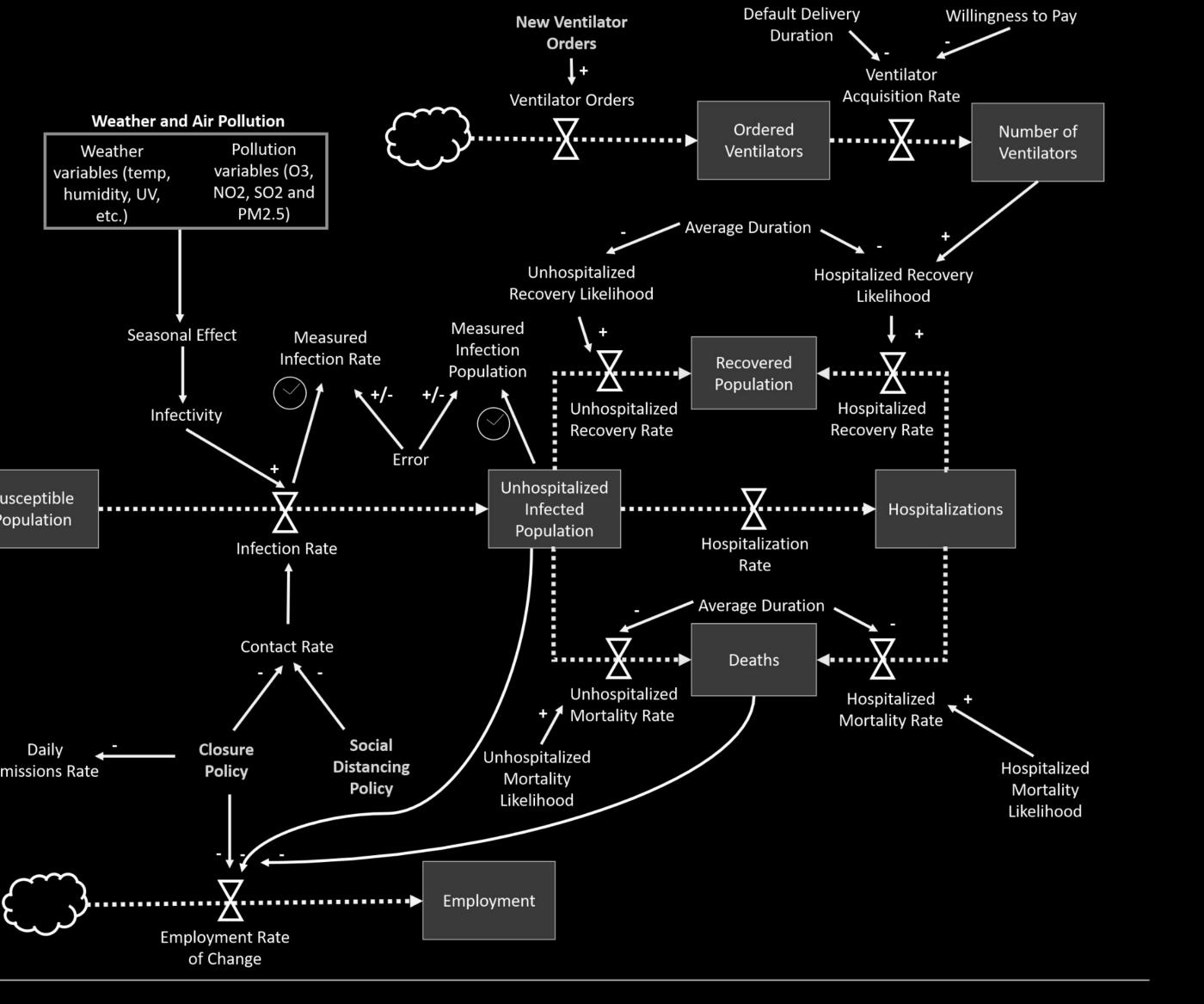
## **Data & Methods: Public Health**

- Epidemiological Model: SEIR
- Modeling Approach: System Dynamics
- Integrates aspects of other Vida components
- Current version is non-spatial

 Adjusting assumptions and policy decisions can generate alternative scenarios

Susceptible Population

Emissions Rate



**Jack Reid** 

Graduate Student, MIT Media Lab Space Enabled research group

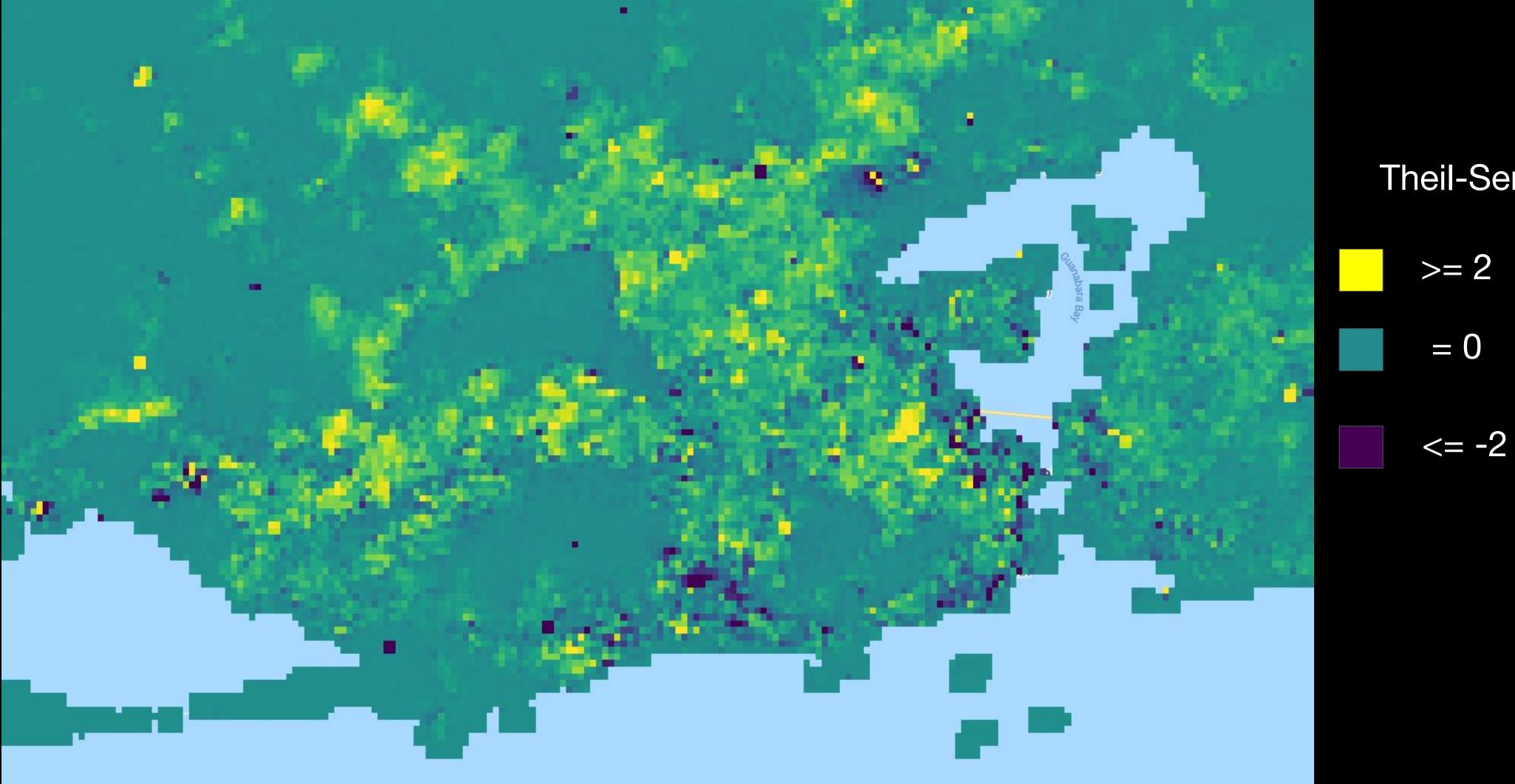
## **Data & Methods: Environment**

- Air Quality (03, NO2, SO2, PM2.5, PM10)
  - Remote: Sentinel 5P
  - In-Situ: Monitoring Stations (Brazil & Chile)
- Nightlights - VIIRS: VNP46A2 & VNP46A3
- Water Quality (NDTI, NDWI, other indices) Landsat 7 ETM+, Landsat 8 OLI, and PlanetScope





## BRAZIL Ex) Rio de Janeiro Nightlight Changes (March - July, 2020)





### Theil-Sen Slope

**Jack Reid** 

Graduate Student, MIT Media Lab Space Enabled research group



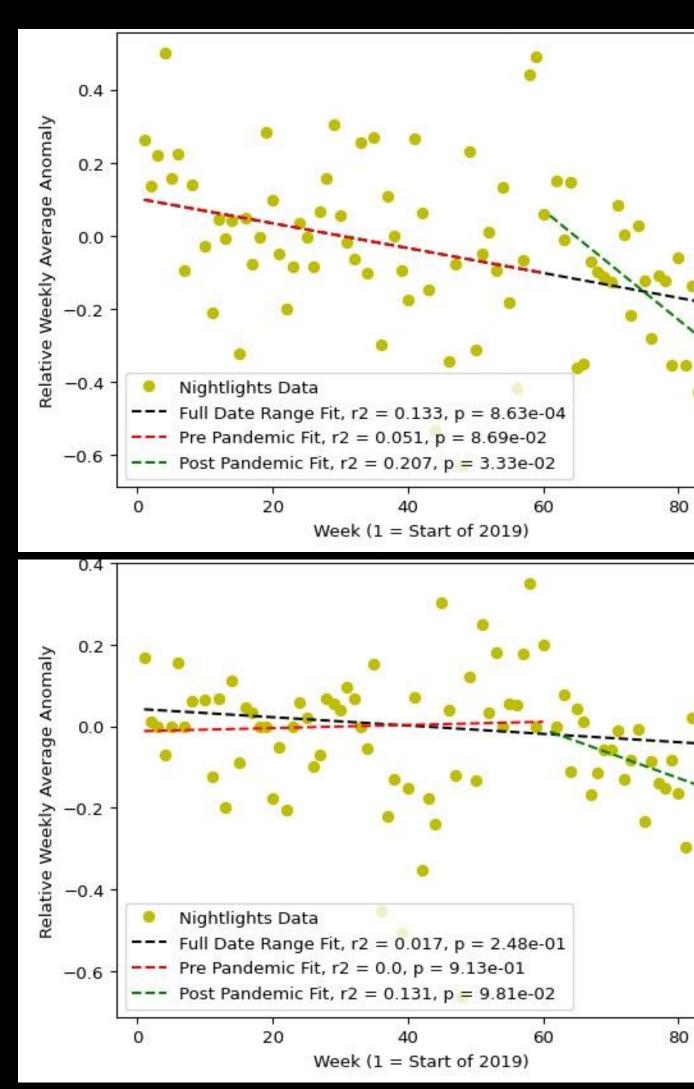




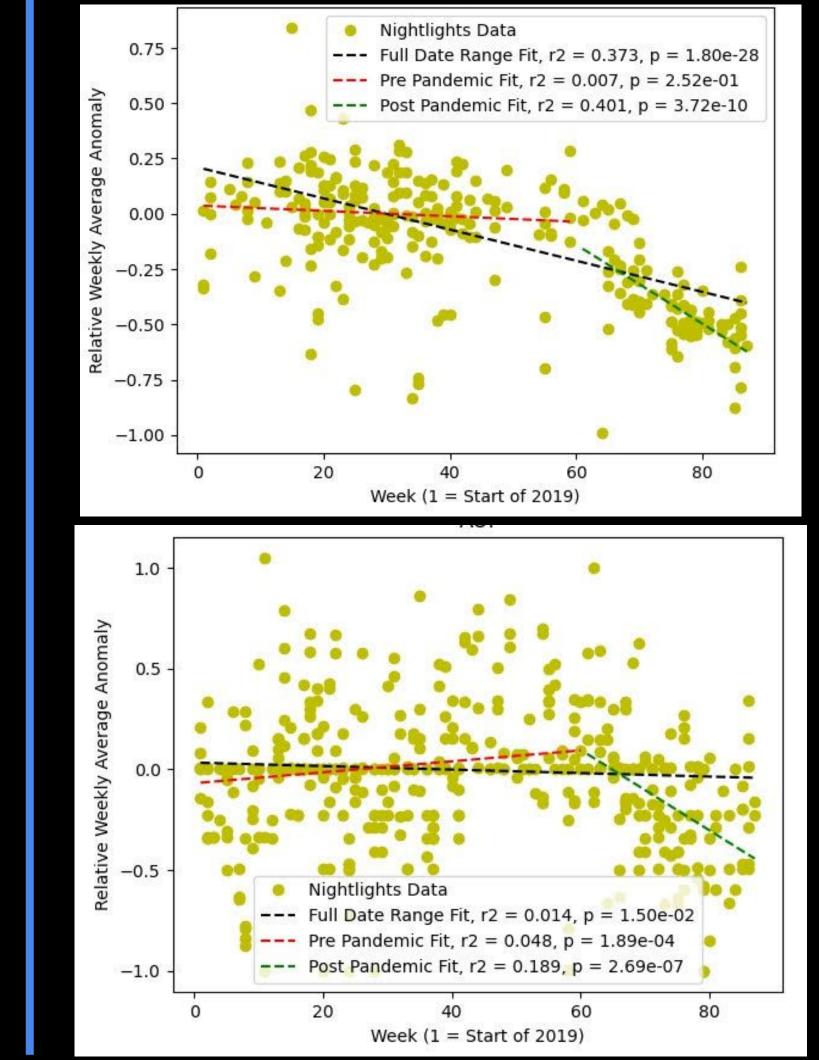
## Rio de Janeiro, Brazil

### Santos Dumont Airport

Ipanema



## Bali, Indonesia



### Ngurah Rai Airport



14

**Jack Reid** 

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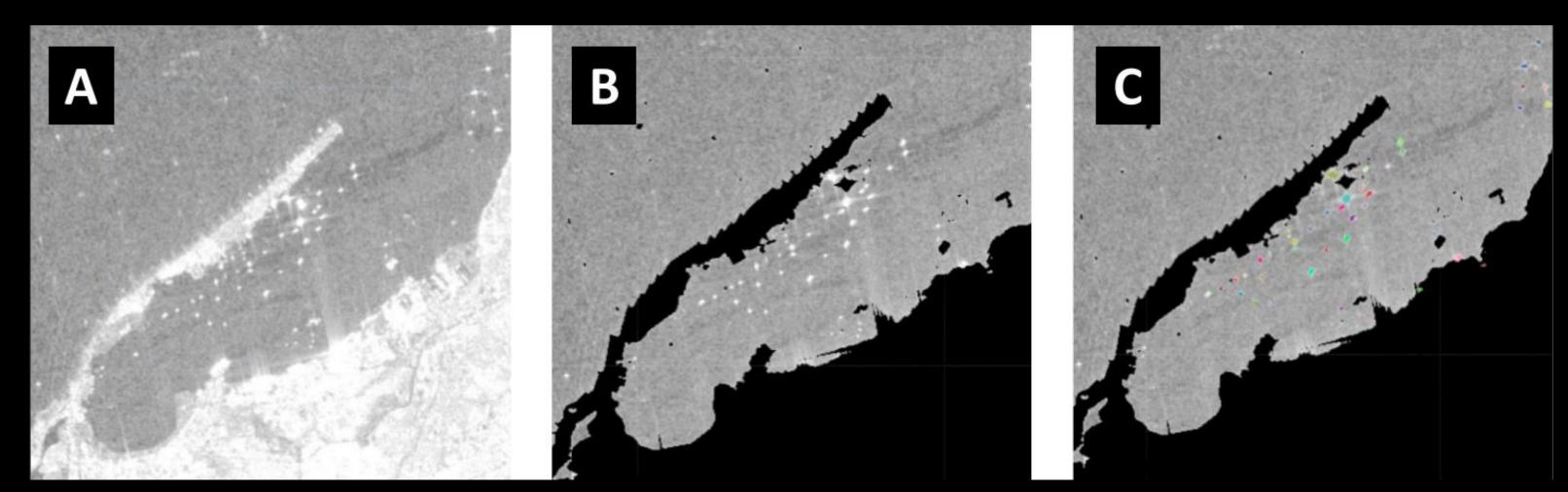
## **Ex) Rio de Janeiro PM10 Changes**

- Relatively small changes in air quality detected once seasonal and long-term trends are taken into account
- What changes do exist point to an increase in PM10

Barrio	Code	Туре	Pre vs Post T-Test P-Value	Anderson Darling P-Value	Change Mean (F vs Post COVID)
Copacabana	AV	Tourist	0.956	0.1438	-0.
Bangu	BG	Mixed Use/Residential	0.2645	0.001	0.
Centro	CA	Downtown/Business District	0.0119	0.00002	0.
Campo Grande	CG	Mixed Use/Residential	0.3806	0.0217	0.
Irajá	IR	Urban/Residential	0.6295	0.0023	0.
Pedra de Guaratiba	PG	Rural	0.7844	0.0801	0.
São Cristóvão	SC	Downtown/Recreational	0.3913	0.0015	0.
Tijuca	SP	Mixed Use/Residential	0.0839	0.00003	0.



## **Data & Methods: Vulnerability**



- Socioeconomic Data
  - Ex) Poverty Rates, Employment Rates, GDP
  - Sources: Local government authorities, NASA SEDAC
- Mobility & Transit Data
  - Telecoms-based mobility data (as reported by Google and local authorities)
  - Public transit usage (as reported by local authorities)
  - Airline Flights (as reported by local authorities)
  - Ship counts and wait periods (as detected in Sentinel radar imagery)

# 

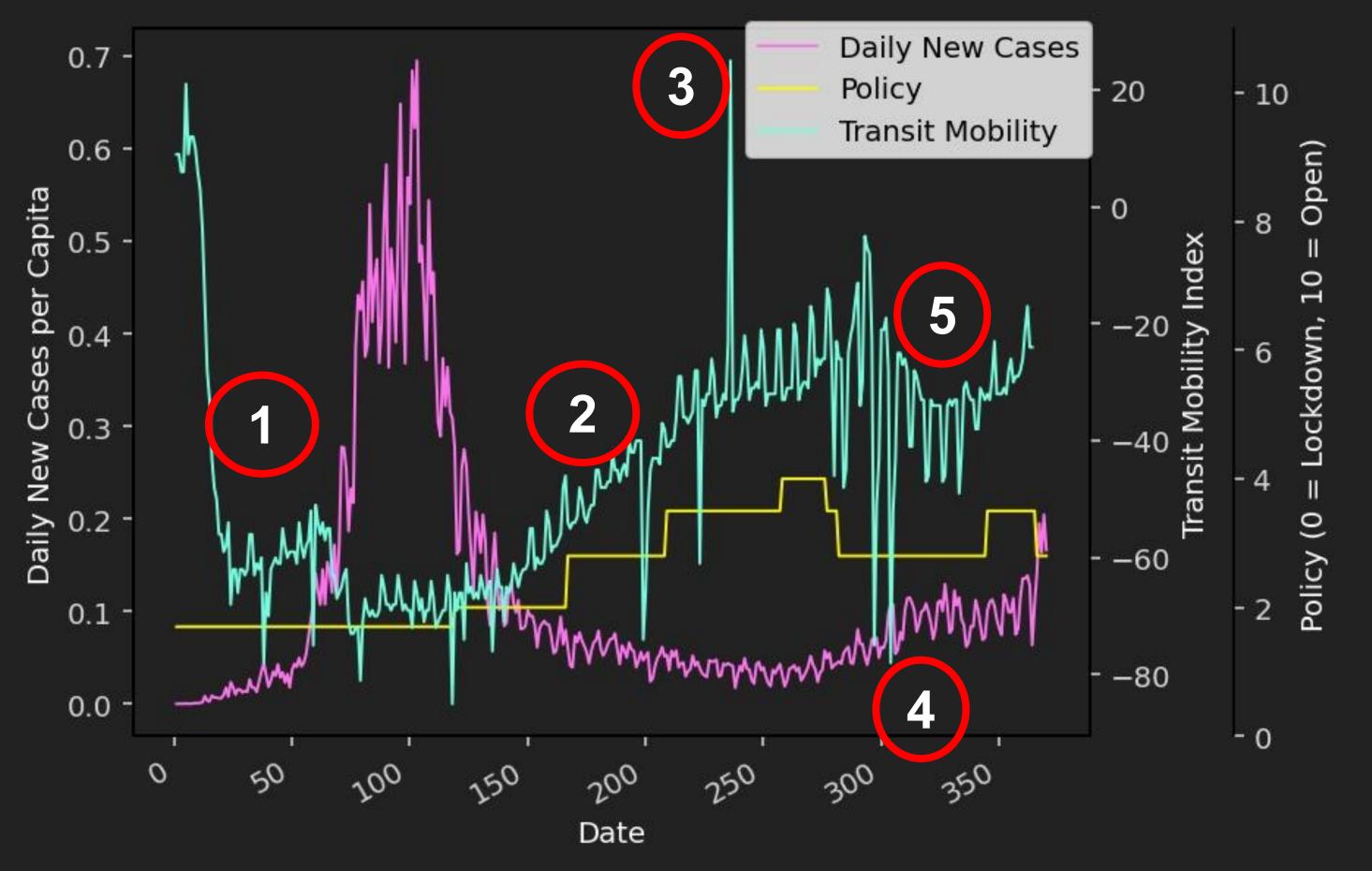
Images and analysis done by Amanda Peyton

**Jack Reid** 

Graduate Student, MIT Media Lab Space Enabled research group



## **Ex)** Metropolitana, Chile Mobility Changes



- 1. Mobility falls, notably *after* the initial wave of policy restrictions went into effect
- 2. As New Cases decline and policy relaxes, mobility rises
- 3. Chile has a constitutional referendum
- 4. Christmas & New Years
- 5. A rise in new cases prompts a policy restriction, decreasing mobility temporarily



## Data & Methods: Decision-making COVID-19 Social Distancing Requirements & Closures Announcements, histories, definitions, and conditions created by local

- authorities
- comparisons based on the CoronaNet Research Project

Recovery P	lan Ind	icators
------------	---------	---------

Reference Date < 07/29/2020		Comparison with previous days 07/16/2020 7/29/202						7/29/2020	WE ARE IN PHASE 6B (Since 01/10/2020)							
GROUP	ANALYSIS PARAMETERS	PRIMARY INDICATORS	F-1	D-5	D-4	D-3	D-2	D-1	Ref. Previous Phase	Result	PHASE 1	PHASE 2	PHASE 3	PHASE 4	PHASE 5	PHASE 6
HEALTH SYSTEM RESPONSE CAPACITY beds		<ul> <li>Percentage of occupancy of dedicated adult ICU beds COVID (ICU SRAG) METRO I SUS bed (7-day moving average)</li> </ul>	8	0	0	0	0	Ø	69.4	71.2	Favorable	Favorable	Favorable	Favorable	Favorable	Favorable
	Capacity of ICU	2 Occupancy rate of supplementary sector ICU beds (moving average 7 days) (a)	8	8	8	×	×	×	67.9	70.0	Favorable	Favorable	Favorable	Favorable	Favorable	Favorable
	beds	<ul> <li>Percentage of occupancy of life support beds REDE SUS Territory of the municipality (moving average 7 days)</li> </ul>	8	0	0	0	0	0	76.0	77.0	Favorable	Favorable	Favorable	Favorable	Favorable	Favorable
		4 ICU COVID beds (REDE SUS) per 100k inhabitants (b)	×	×	×	×	×	0	6.59	6.41	Favorable	Favorable	Favorable	Favorable	Favorable	Favorable
TRANSMISSION LEVEL	Variation of deaths	5 Death Variation Rate by COVID19 in each period (Information released at 6 pm on the day, referring to the previous day) (c)	8	θ	8	θ	0	0	0.92	0.95	Favorable	Favorable	Favorable	Favorable	Favorable	Favorable
	Growth of hospitalized cases	6 Rate of Variation of Inpatients (Clinical + ICU) in each period (Information released at 6 pm on the day, referring to the previous day) (c)	8	0	0	0	0	0	0.92	0.95	Favorable	Favorable	Favorable	Favorable	Favorable	Favorable
	Variation of new cases	7 Number of cases reported by Influenza Syndrome (SG) in the last two epidemiological weeks of notification (d)	0	Ø	0	θ	θ	θ	16,554	13,931	Favorable	Favorable	Favorable	Favorable	Favorable	Not Favorable
OPINION FOR OPENING PHASE ACCORDING TO PRIMARY INDICATORS								Favorable	Favorable	Favorable	Favorable	Favorable	Not Favorable			

# - Ongoing effort to compare policies using standardized, quantitative

updated 01/10/2020

### For more information, see https://riocontraocorona.rio/ and http://inteligencia.rio/planoretomada

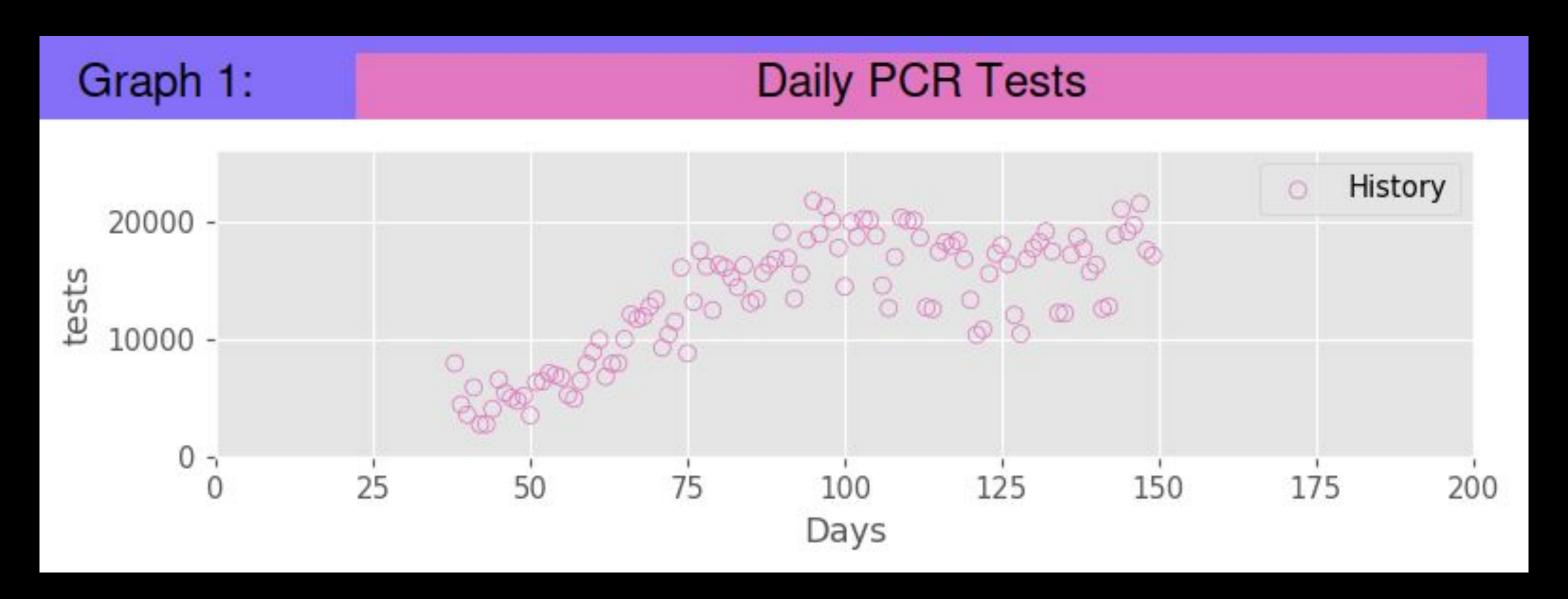
**Jack Reid** 





## Data & Methods: Technology • Earth observation systems are still relevant!

- Various public health sensing technologies and regimes
  - PCR and other tests to identify the actively infected
  - Antibody tests to identify those previously infected





- Additional relevant platforms like VIIRS, MODIS, Planet, Maxar, etc.





## **User Interface**



## **User Interface**



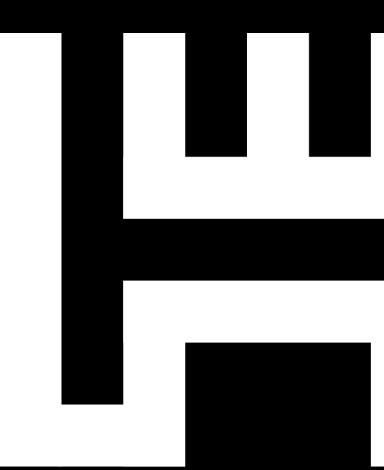
## **Ongoing and Future Work**

- Automating data updates and ingestion
- Standardizing architecture to facilitate reuse
- Add simulation capabilities to the online version
- Improving visualizations

- Adding a spatial component to the epidemiological model
- Continue air quality and nightlight analysis

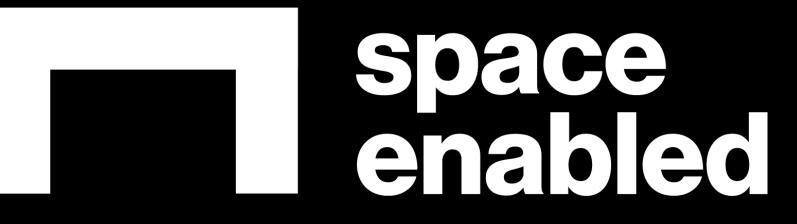


## **Contact Information:**



## **Project Page:** https://www.media.mit.edu/projects/vida-decision-support-system/overview/

- jackreid@mit.edu
- https://twitter.com/Jack B Reid
  - **Acknowledgements:**
  - Soffen Memorial Fund
  - All of our collaborators



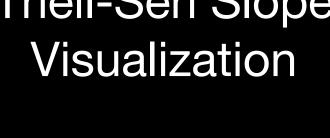
# Methodology

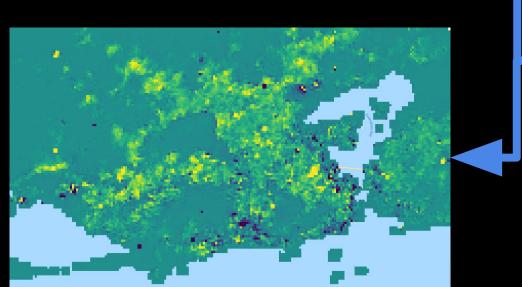
### VNP46A2 Raw Image

<u>Filter</u> Area of Interest Clouds Water

Theil-Sen Slope Visualization

Statistical Analysis













## Weekly Averages

## Percent Change Relative to 2019 Annual Average



Start of Pandemic - 1/Aug/21

### Split Data into Temporal Categories

Select specific geographic subunits

**Jack Reid** 

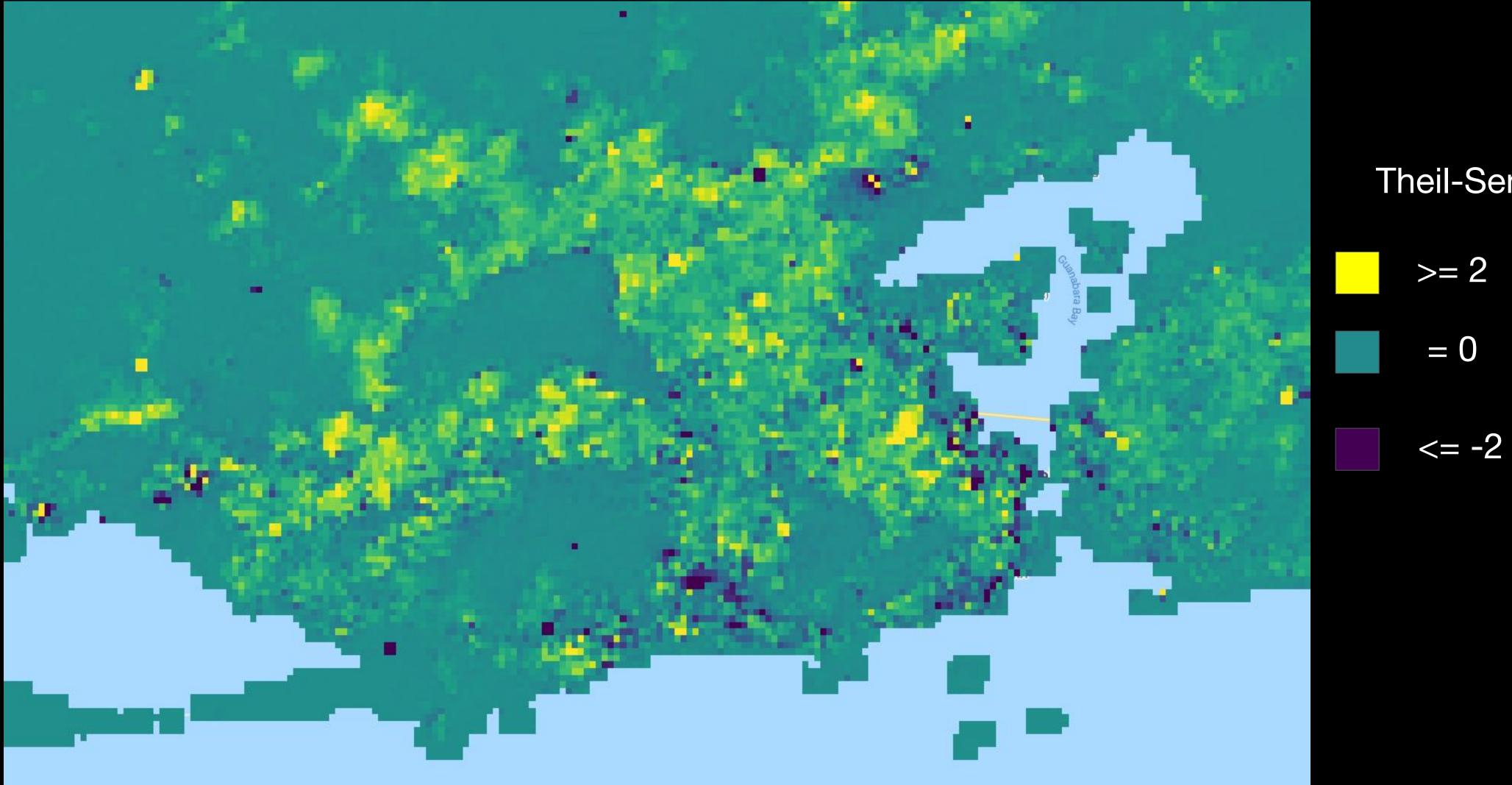
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## BRAZIL Visualization - Rio de Janeiro Changes (March - July, 2020)





### Theil-Sen Slope

**Jack Reid** 

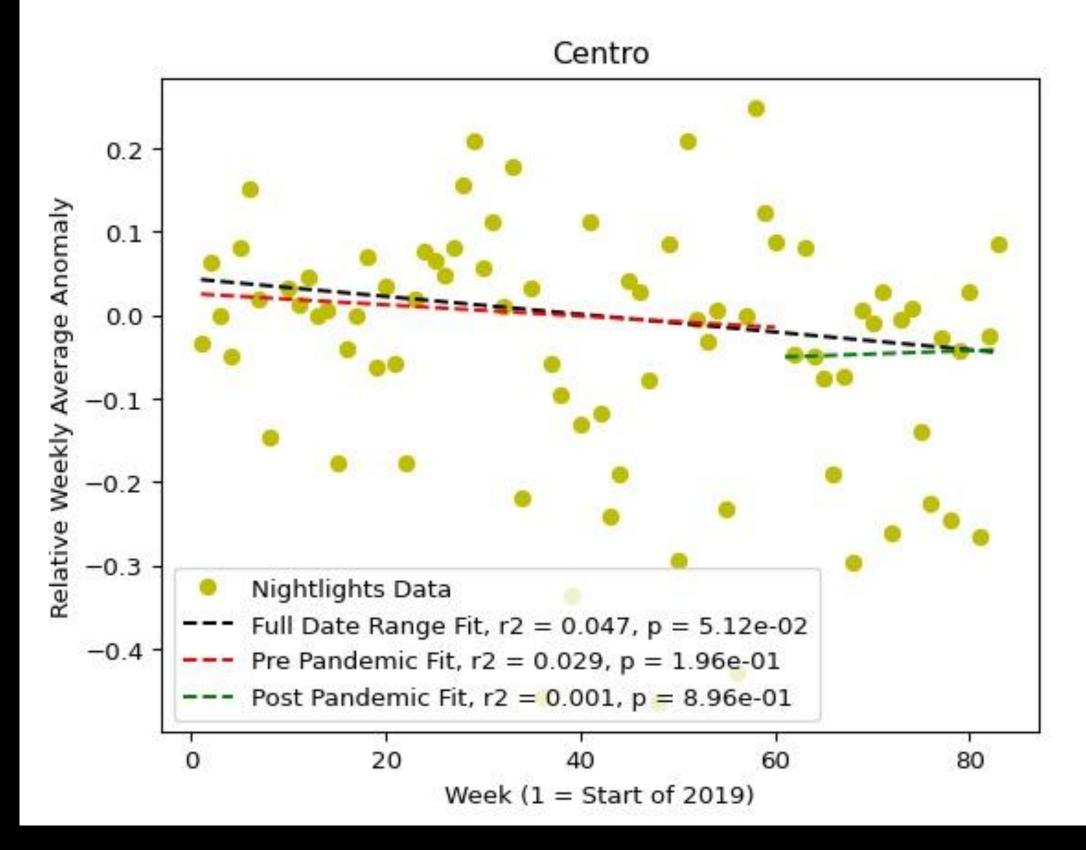
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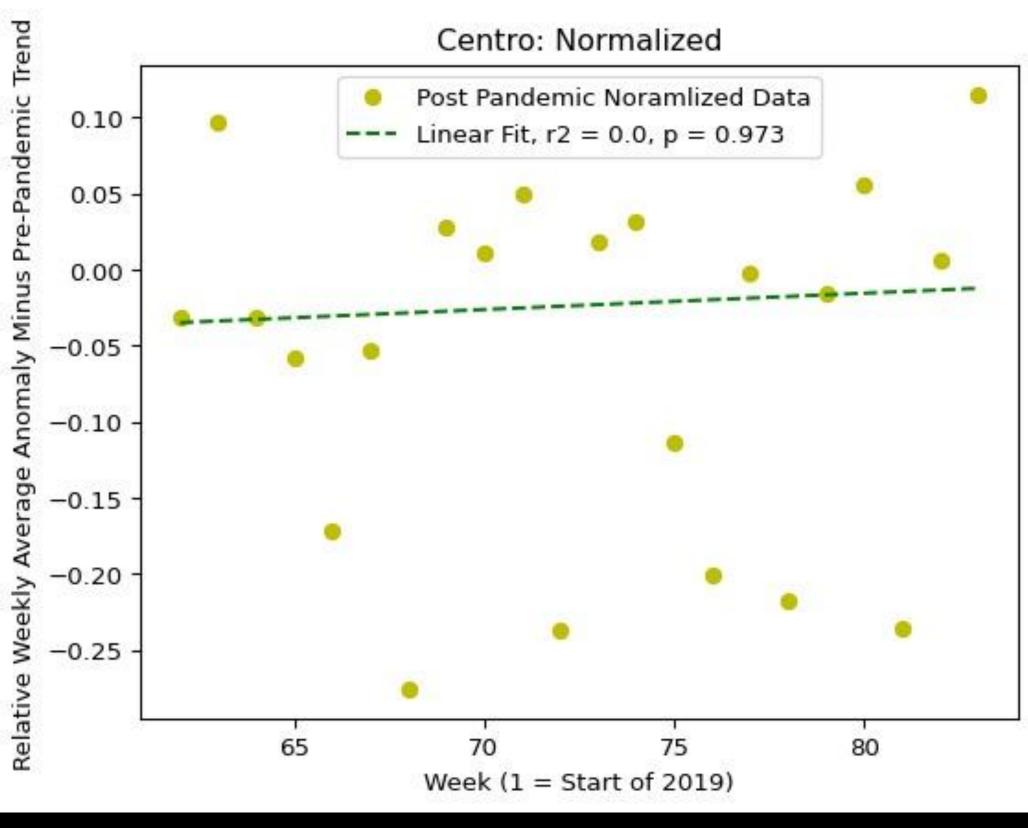


## **Statistics - Rio Changes**



Percent Change Relative to 2019 Annual Average

BRAZIL

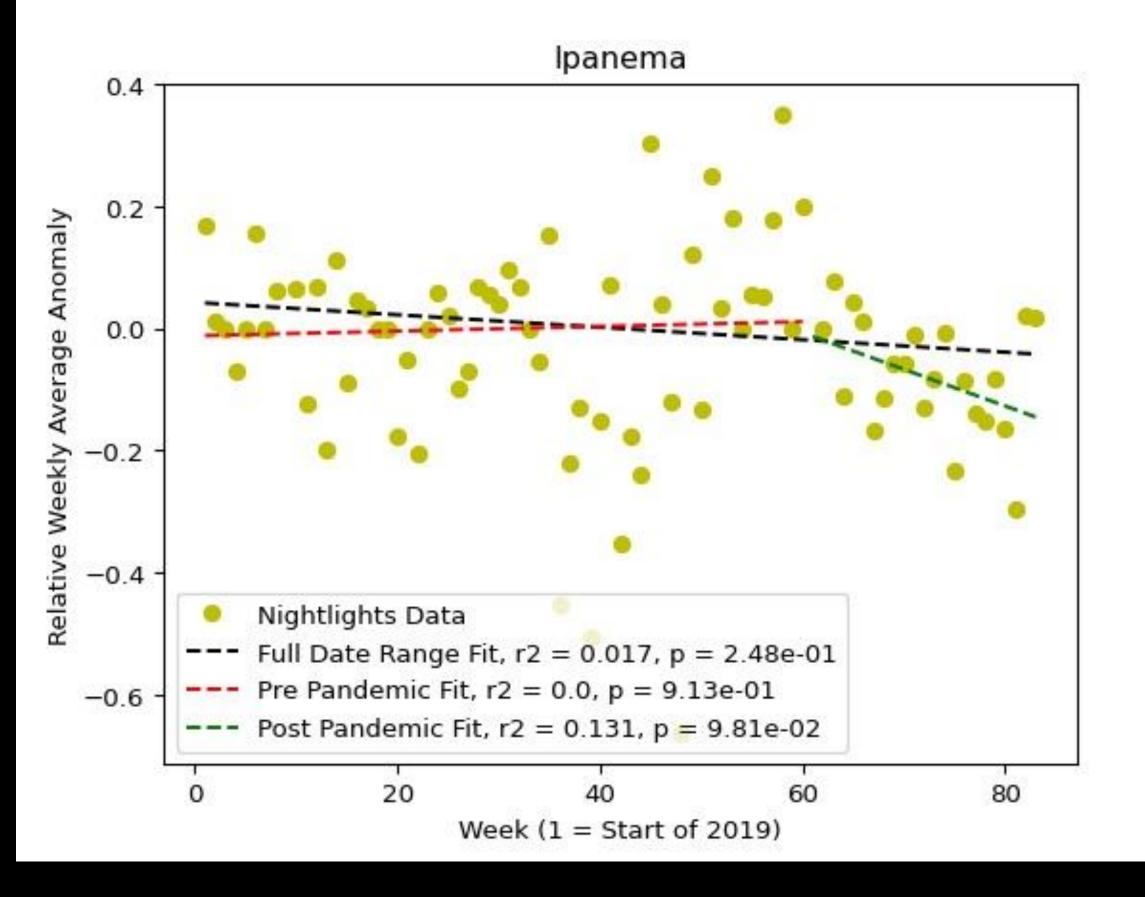


Normalized: (Post Pandemic Data - Red Line Trend)

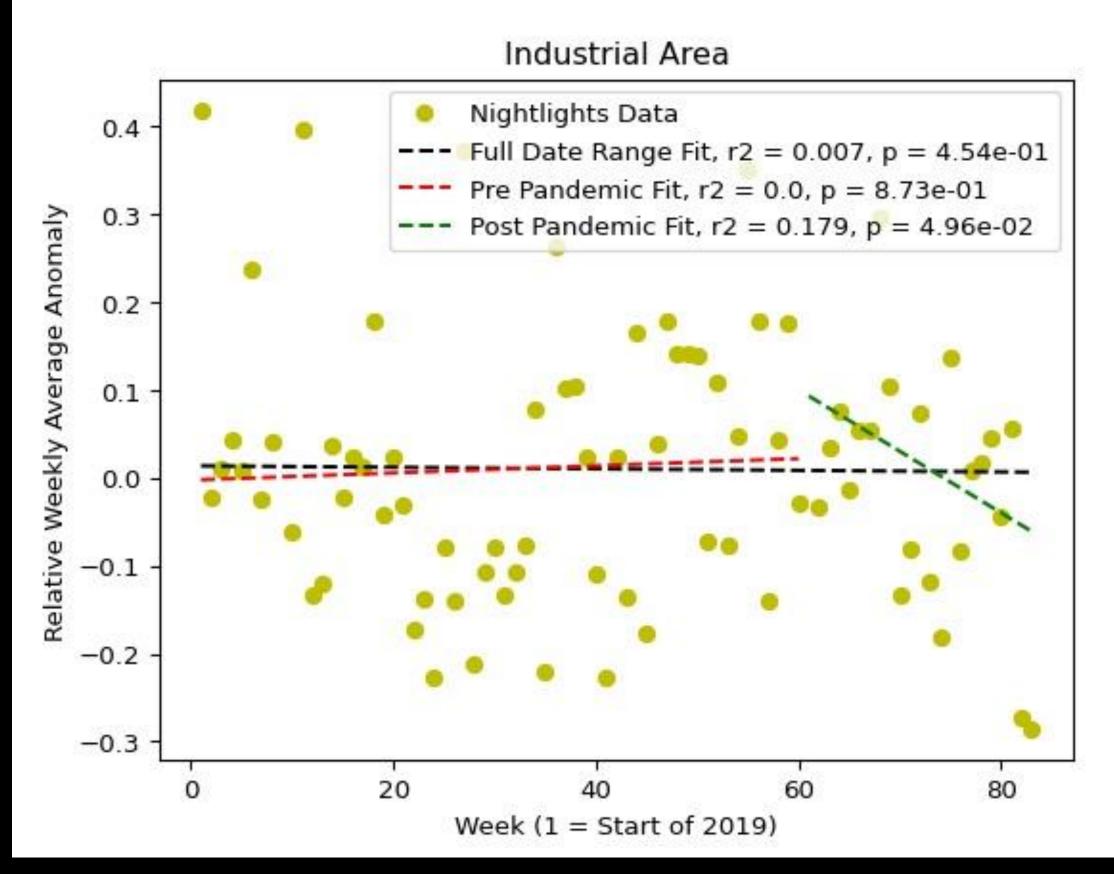
**Jack Reid** 



## **Statistics - Rio Changes**







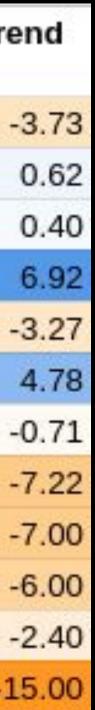


## **Rio de Janeiro Changes**

Area	Туре	Pre vs Post T-Test P-Value	Normalized Data Linear Fit P-Value	Pre Pandemic Trend (*1000)	Post Pandemic Tre (*1000)
Barra da Tijuca	Tourist	0.000	0.11	-0.64	-
Campo Grande	Suburb	0.503	0.93	0.25	
Centro	Downtown	0.115	0.97	-0.67	
Cidade de Deus	Mixed Use / Residential	0.433	0.01	-0.50	
Cidade Nova	Downtown	0.604	0.88	-3.76	
City	Entire City	0.347	0.45	0.58	
Copacabana	Tourist	0.769	0.90	-1.44	-
Galeao Airport	Airport	0.000	0.24	-2.57	-
Industrial Area	Heavy Industry	0.395	0.04	0.41	
Ipanema	Tourist	0.063	0.08	0.38	-
Pedra de Guaratiba	Rural / Residential	0.052	0.70	-0.76	-
Santos Dumont Airport	Airport	0.005	0.12	-3.38	-1



### BRAZIL

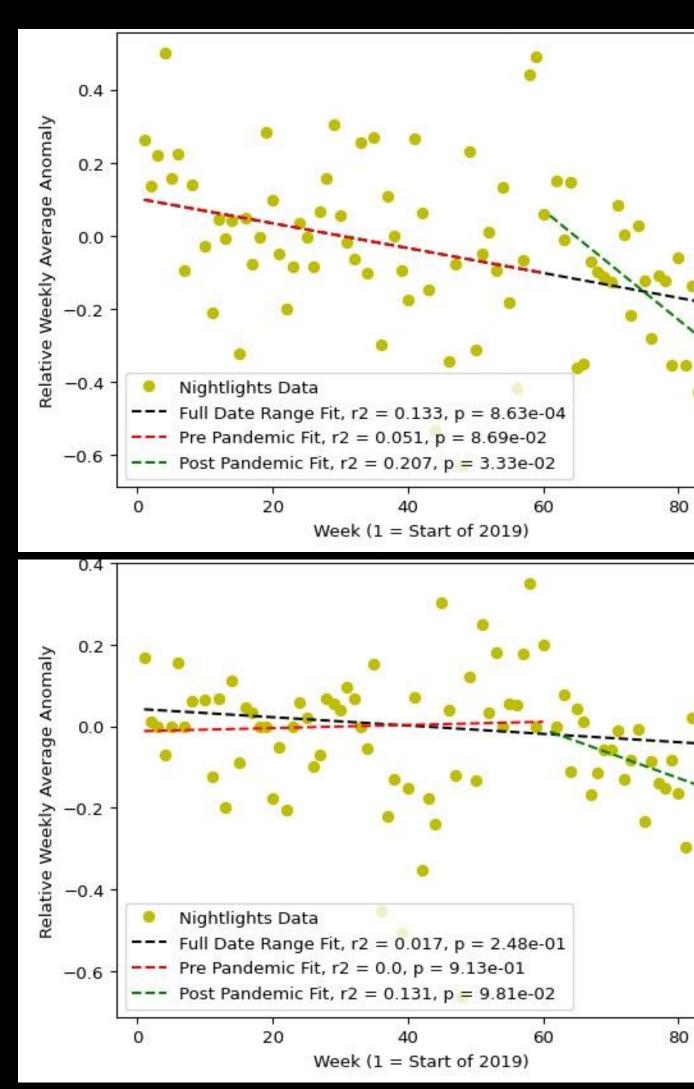




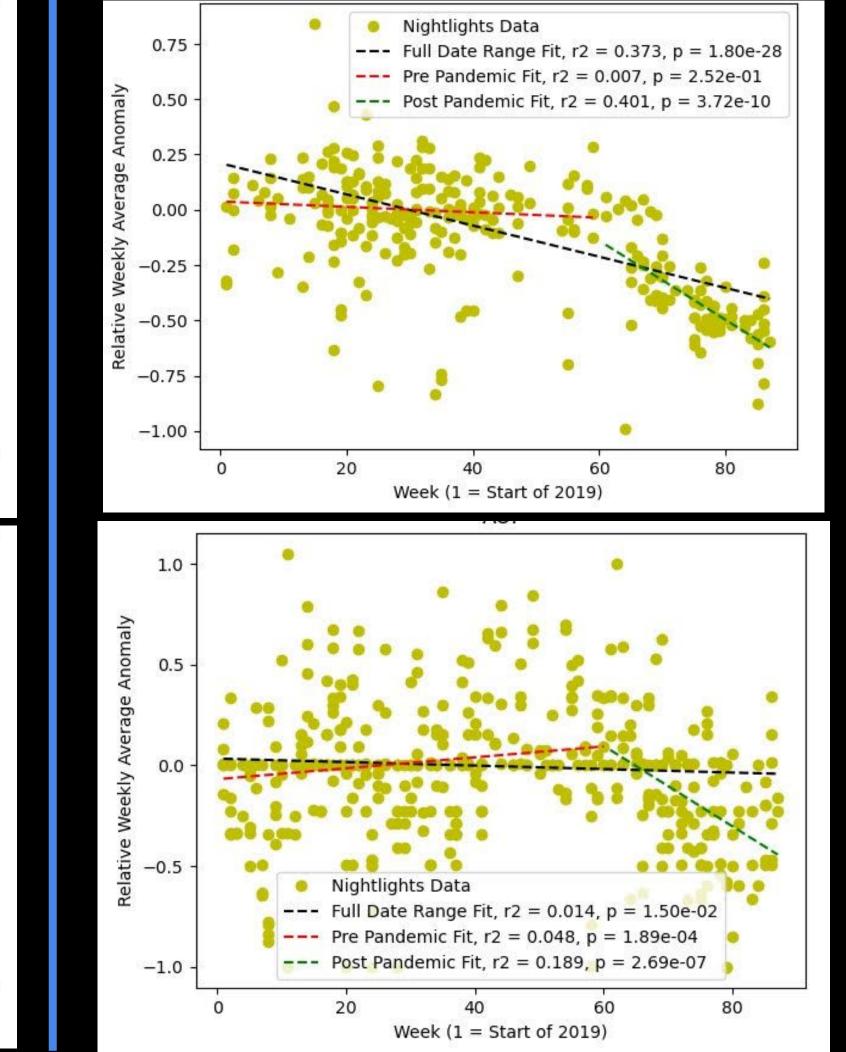
## Rio de Janeiro, Brazil

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## Bali, Indonesia



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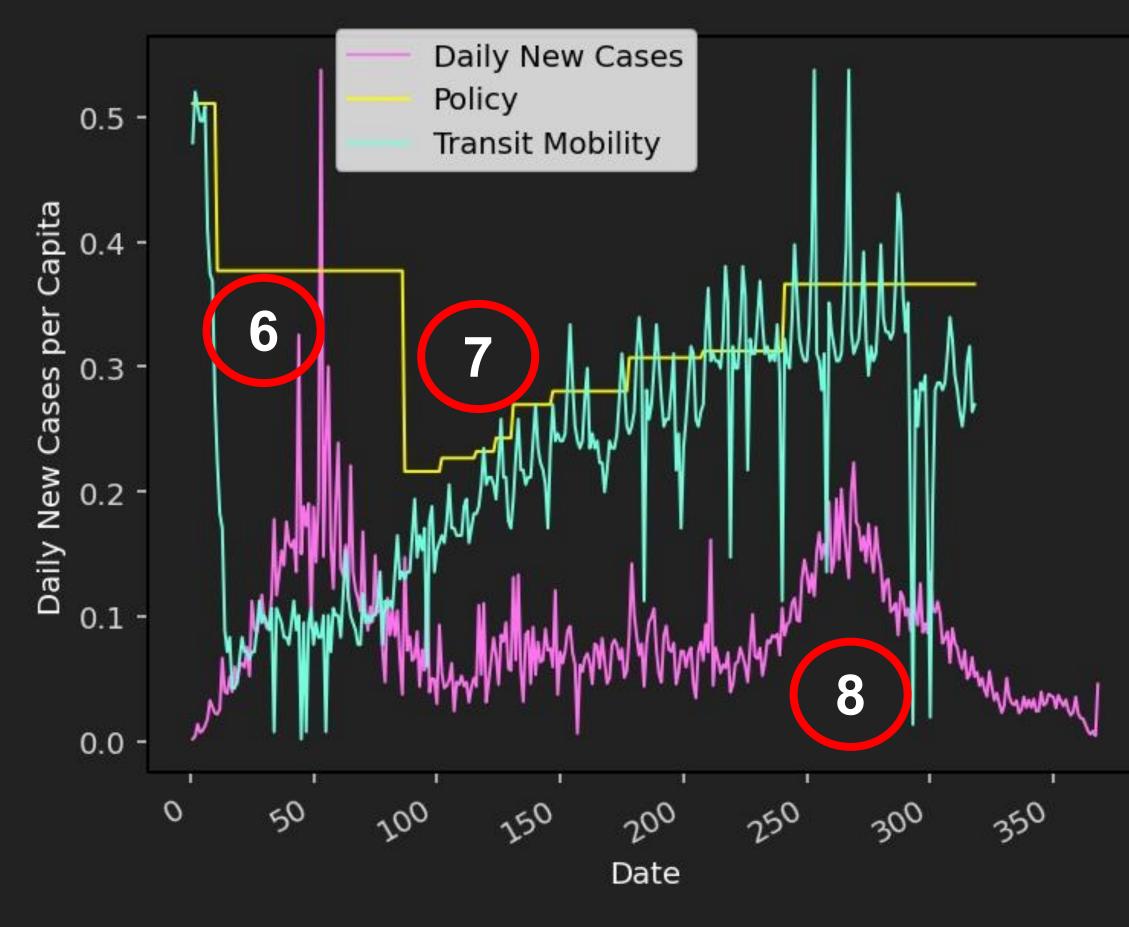
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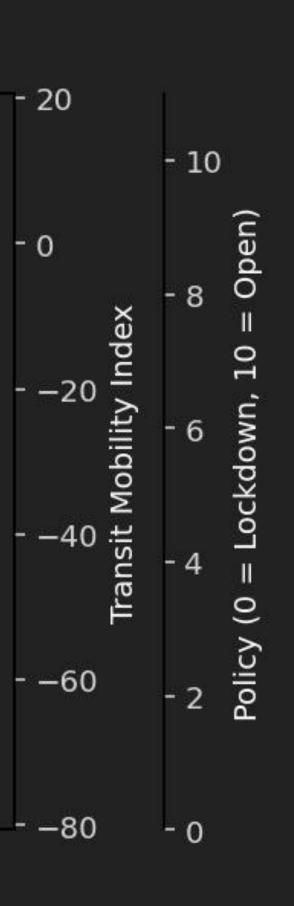
**Jack Reid** 





## **Rio de Janeiro**





6. Mobility falls, matching or even leading actual policy changes 7. Mobility rises, leading policy changes upwards as case counts fall 8. Mobility drops starkly for Christmas and New Years, then returns to a lower level than previously, following a rise in cases and a new government with different priorities.



