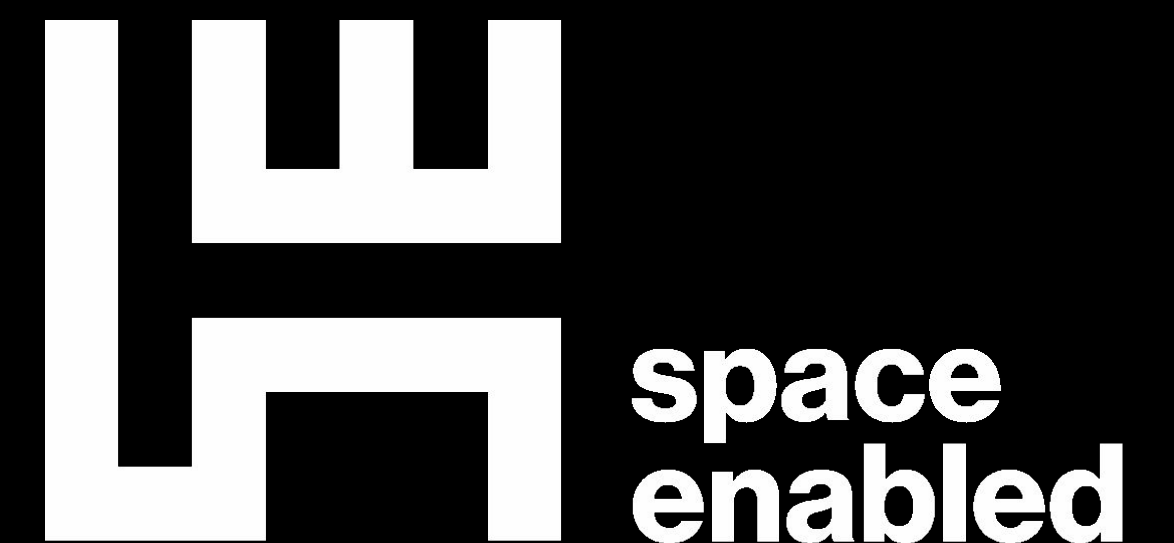
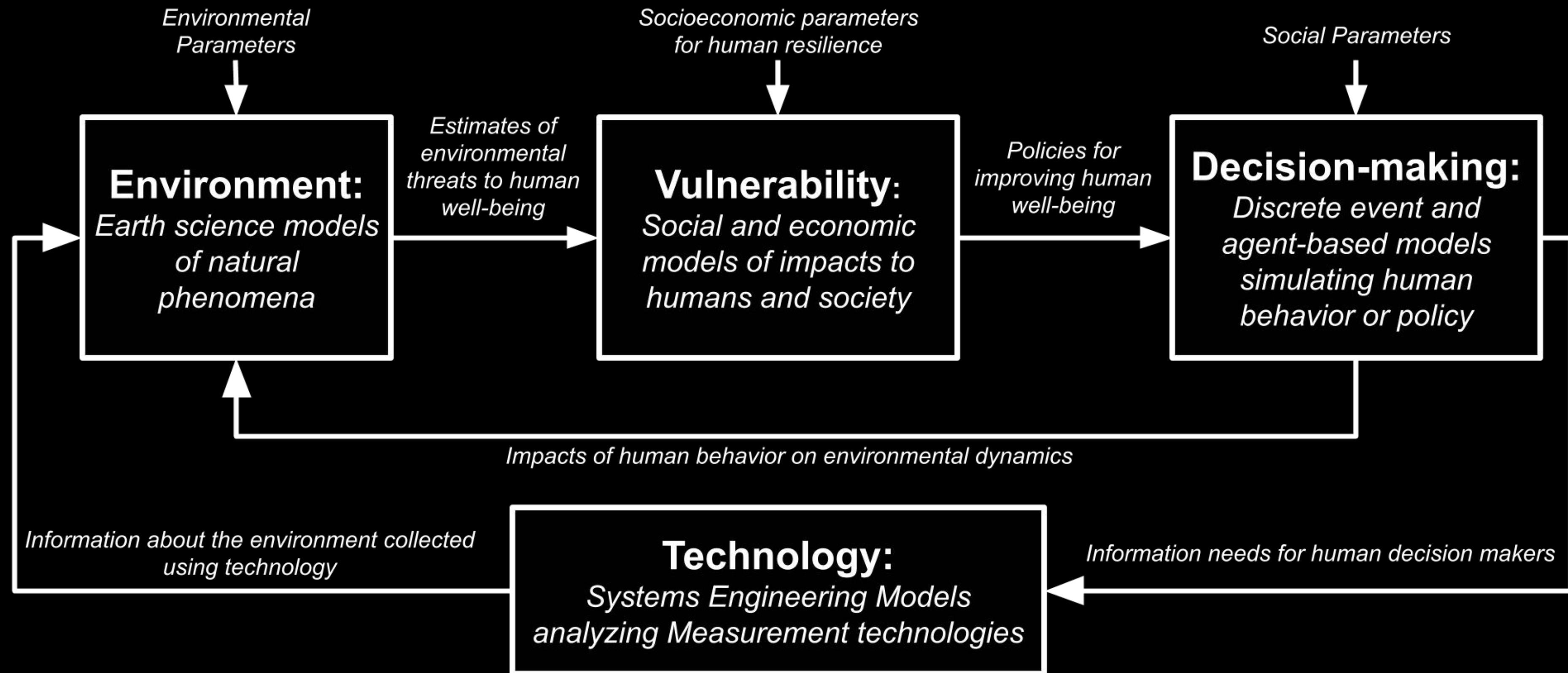


International Collaboration Aimed at Identifying Relevant Social, Policy, and Environmental Factors in the Progression of SARS-CoV2/COVID-19 in Six Metropolitan Areas

Jack Reid, Seamus Lombardo, David Lagomasino, Eric Ashcroft, Paulina Assmann, Carlos Ávila, Mary Bracho, Joana Caetano, Hanifa Denny, Mohammad Jalali, Felipe Mandarino, Amanda Payton, Joaquin Salas, Gilson Santos, Joga Setiawan, Eduina Teodoro, Katlyn Turner, Maggie Zheng, Danielle Wood



EVDT Framework

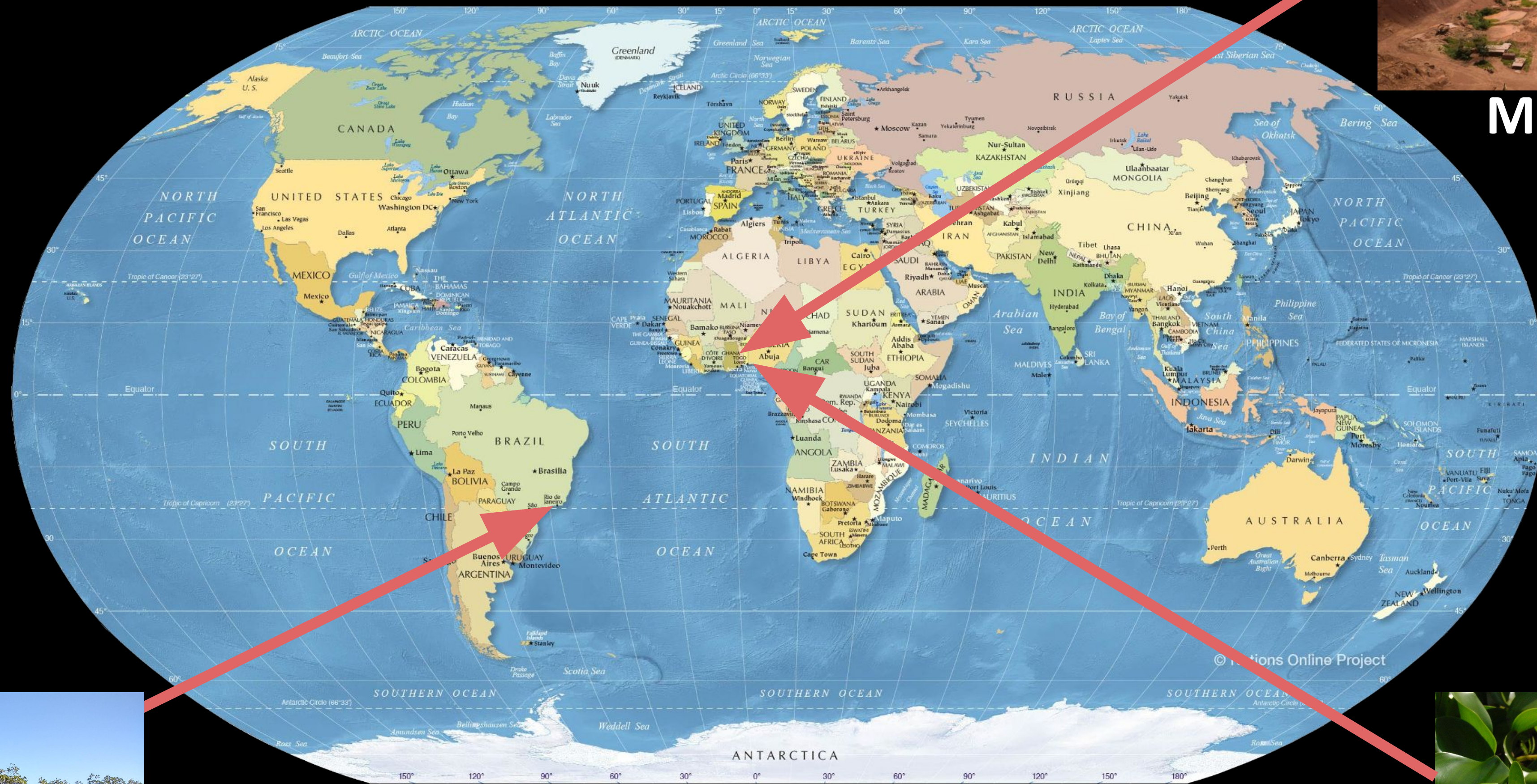


- What is happening in **the natural environment**?
- How will **humans be impacted** by what is happening in the natural environment?
- What **decisions are humans making** in response to environmental factors and why?
- What **technology system** can be designed to provide high quality information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

Some Pre-Pandemic EVDT Applications



Mining in Ghana



Map adapted from the Nations Online Project.

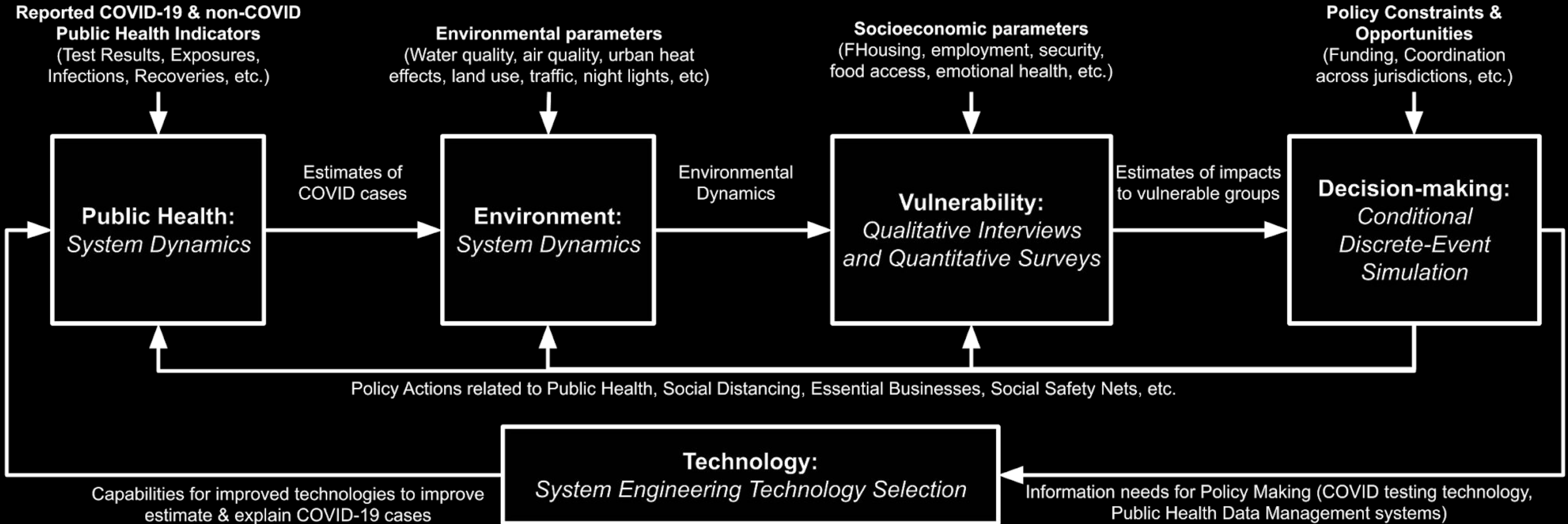


Mangroves in Rio de Janeiro



Water Hyacinth in Benin

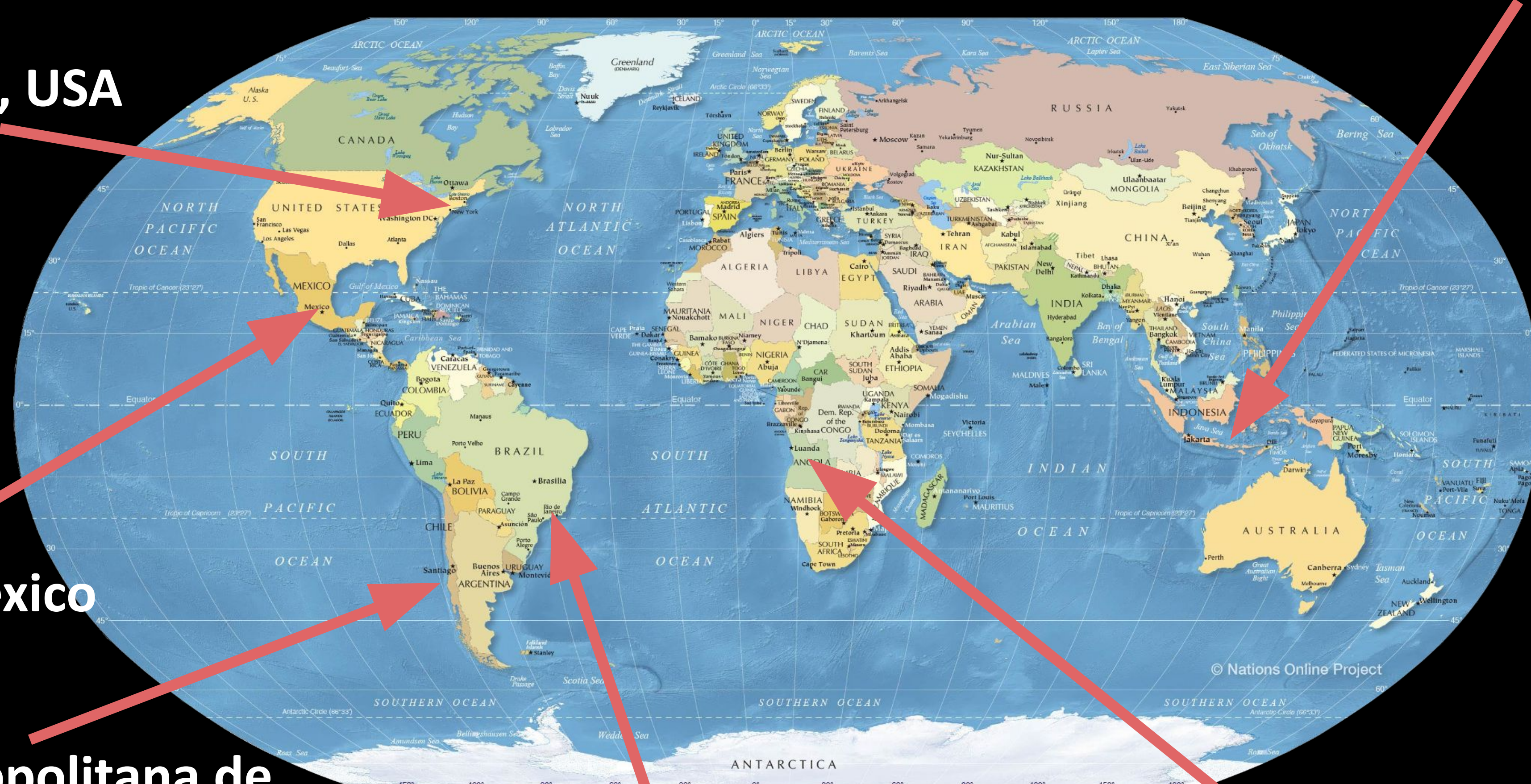
Vida Decision Support System



Vida DSS International Network

Java & Sulawesi, Indonesia

Boston, USA



Querétaro, México

Región Metropolitana de Santiago, Chile

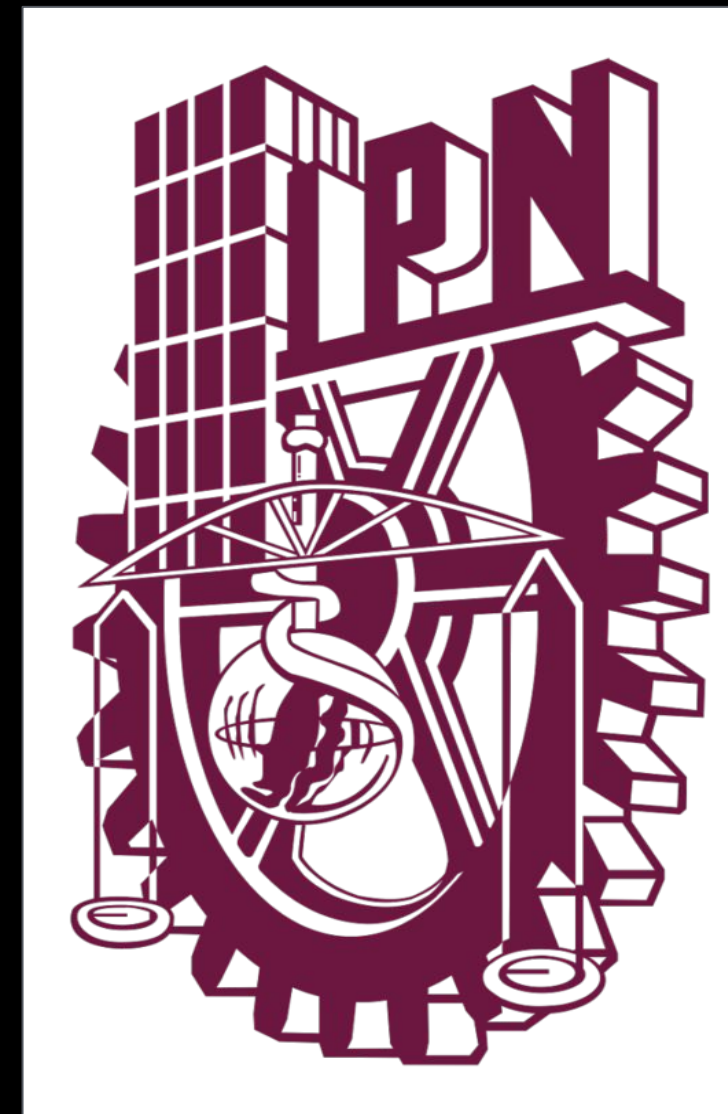
Rio de Janeiro, Brasil

Luanda, Angola

Brasil



México



Indonesia



Chile

Angola



Datos-COVID19

El objetivo de la Mesa de Datos COVID-19 liderada por el Ministerio de Ciencia, Tecnología, Conocimiento e Innovación es disponer de información de nuestro país durante la pandemia para promover el uso de datos para investigación científica, clínica y para soluciones innovadoras que contribuyan a la toma de decisiones de las autoridades y la ciudadanía frente a esta pandemia. Se disponen los datos epidemiológicos provenientes del Ministerio de Salud (MINSAL) y datos de otras fuentes, documentados y abiertos para el análisis de la comunidad, en concordancia con la Ley N° 19.628.

Ver <https://minciencia.gob.cl/covid19/> para más información, incluyendo actas de las reuniones de la mesa y también los informes publicados a la fecha.

Data Products

Data Product 1 - Casos totales por comuna incremental: Archivo con valores separados por coma (csv) que concatena historia de publicaciones de MINSAL sobre casos confirmados totales por comuna. Contiene las columnas 'Región', 'Código Región', 'Comuna', 'Código comuna', 'Población', múltiples columnas correspondientes a '[fecha]', y una columna 'Tasa'. Incluye versión con serie de tiempo. Ver más.

Data Product 2 - Casos totales por comuna: Archivos con valores separados por coma (csv) con la información de casos confirmados notificados a nivel comunal por cada informe publicado. Cada archivo contiene las columnas 'Región', 'Código Región', 'Comuna', 'Código comuna', 'Población' y 'Casos Confirmados'. Ver más.

Data Product 3 - Casos totales por región incremental: Archivo con valores separados por coma (csv) que concatena historia de publicaciones de casos totales por parte de MINSAL. El archivo contiene una columna 'Región', seguida por columnas correspondientes a '[fecha]'. Estas últimas columnas, '[fecha]', contienen los 'Casos Confirmados' reportados por el Ministerio de Salud de Chile en cada una de las fechas que se indican en las respectivas columnas. Incluye versión con serie de tiempo. Ver más.

Open In Colab



Bienvenido al jupyter notebook demo del MinCiencia

Para empezar

Este documento te permite interactuar con los datos que se encuentran en el [repositorio](#) del Ministerio de Ciencia, Tecnología, Conocimiento e Innovación. Estos datos son recopilados de las cifras oficiales publicadas por el [Ministerio de Salud](#) sobre la pandemia del COVID-19 en Chile.

Este notebook es un ejemplo y puedes usarlo como base para generar tus propios gráficos y/o productos. Los datos publicados están en <https://colab.research.google.com/github/MinCiencia/output/blob/master>

Para trabajar con los productos

Este notebook está escrito utilizando el lenguaje de programación Python versión 3.x, cuya lectura se facilita a programadores no expertos. Cada bloque de ejecución está separado en distintas celdas, es necesario "ejecutar" cada una en secuencia haciendo click en botón "play" que aparece al posicionar el mouse sobre el recuadro [] al inicio de cada celda. Una vez que la celda se ejecuta, aparece un número que indica el orden en que se ha ejecutado.

Una manera común de manipular los datos, es usando [pandas](#). Para cargar uno de los archivos en un dataframe. También es necesario utilizar [numpy](#) para los distintos cálculos

```
In [0]: import numpy as np
import pandas as pd
```

Los datos están almacenados como tablas en formato csv. Algunas tablas están almacenadas por día. Es importante mantener el formato 'año-mes-día' (las comillas simples o dobles indican que es texto y no se ejecutará como una operación matemática)

```
In [0]: # el símbolo # al inicio de la línea nos indica que esto es un comentario, no se ejecuta con el código
# date indica la fecha de los datos que queremos utilizar

date = '2020-04-20'
```

Para leer la tabla correspondiente al día escogido y se le asigna el nombre dataTotalRegion:

```
In [0]: dataTotalRegion = pd.read_csv("https://raw.githubusercontent.com/MinCiencia/Datos-COVID19/master/output/producto4/"+date+"-CasosConfirmado")
```

Utilizamos read_csv de pandas con el parámetro

```
# index_col = 'Region'
```

para indicar que el índice de la tabla son los nombres de las regiones. Los nombres de las columnas se listan a continuación

```
In [0]: columnas = list(dataTotalRegion.columns.values)
columnas
```

```
Out[0]: ['Casos nuevos', 'Casos totales', '% Casos totales**', 'Fallecidos']
```

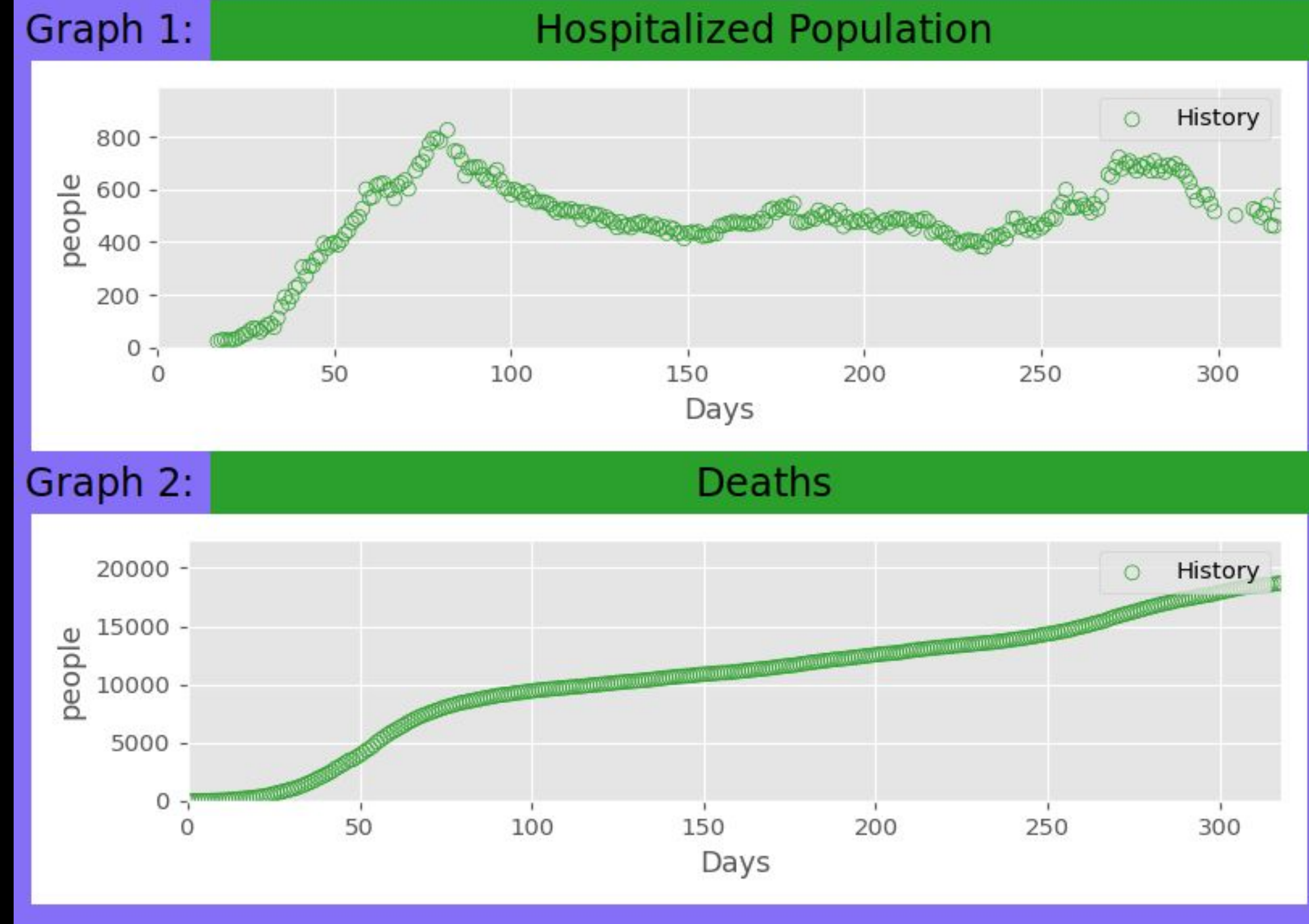
Podemos ver los valores de las columnas ejecutando en una celda

```
# dataTotalRegion
```



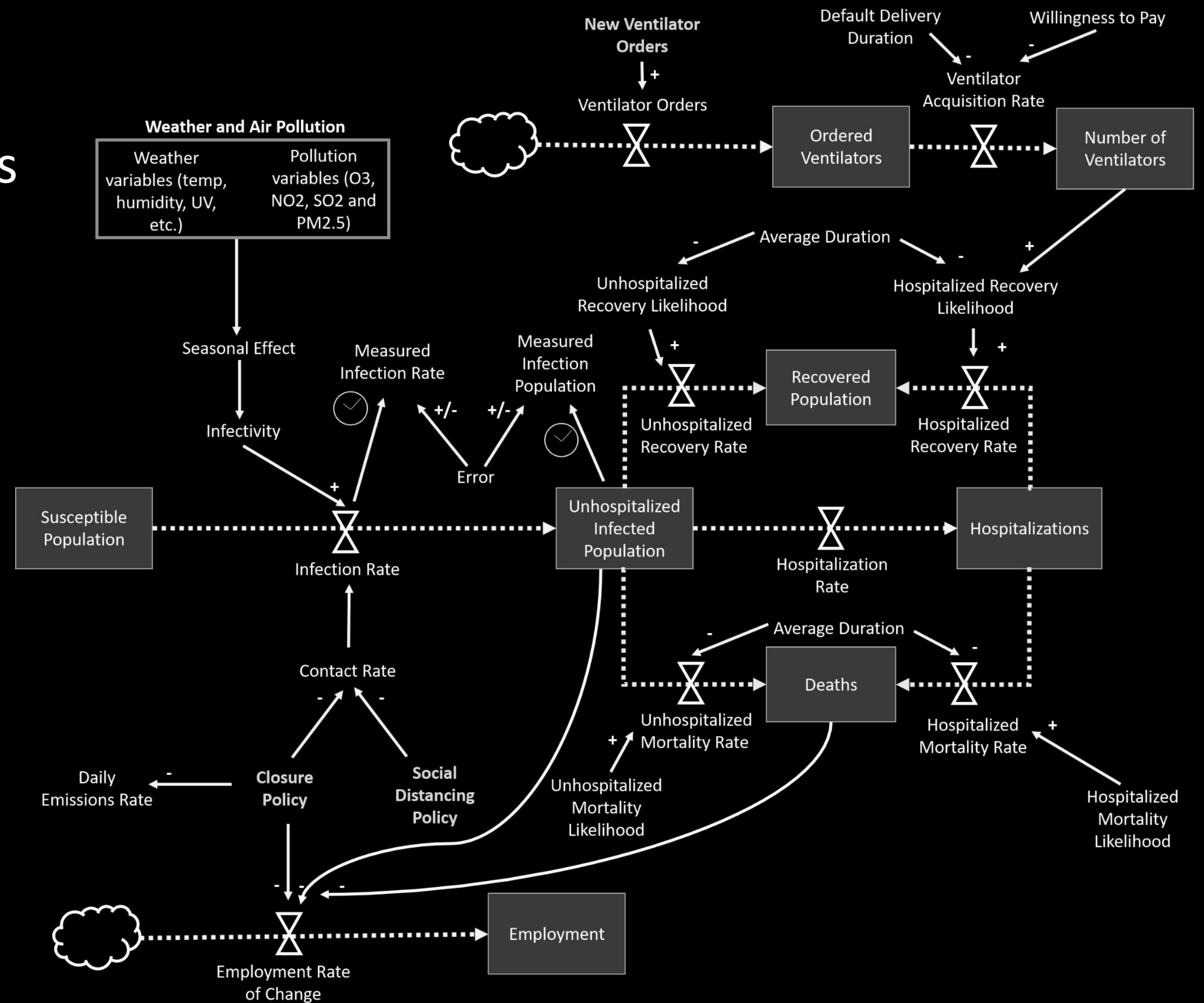
Data & Methods: Public Health

- 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



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



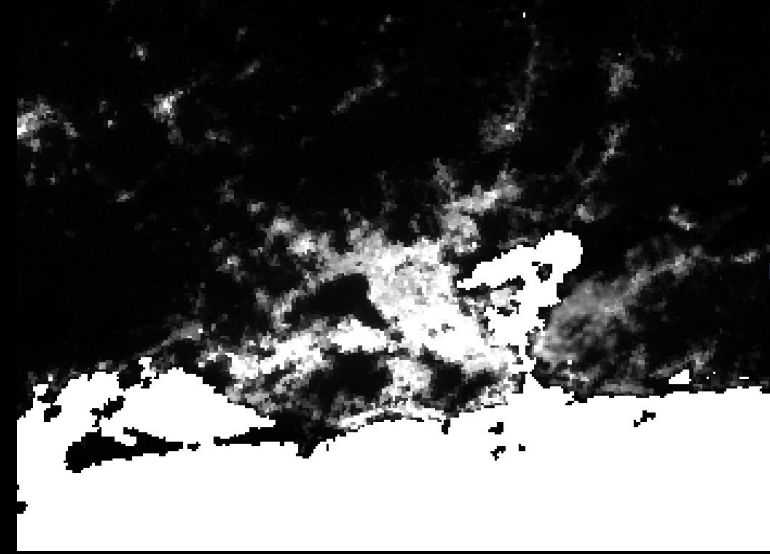
Data & Methods: Environment

- Air Quality (O3, 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

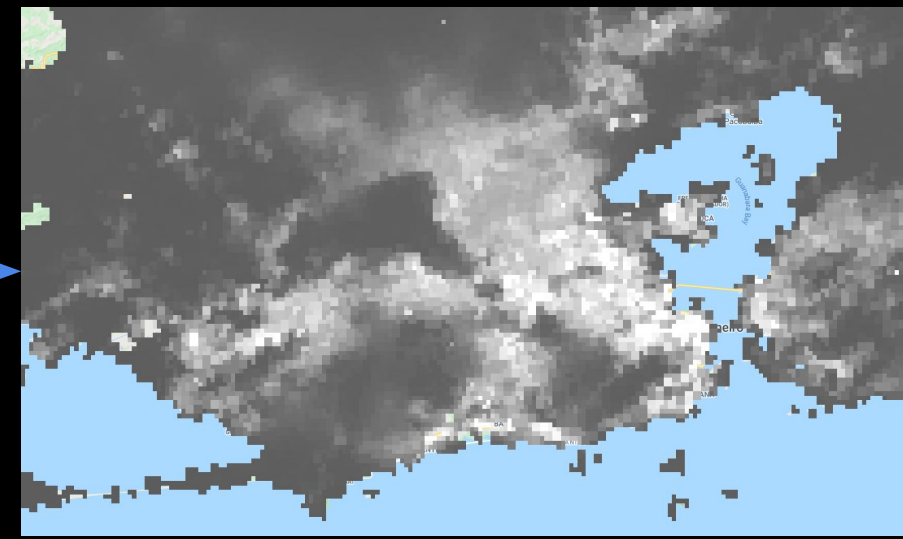


Nightlights Methodology

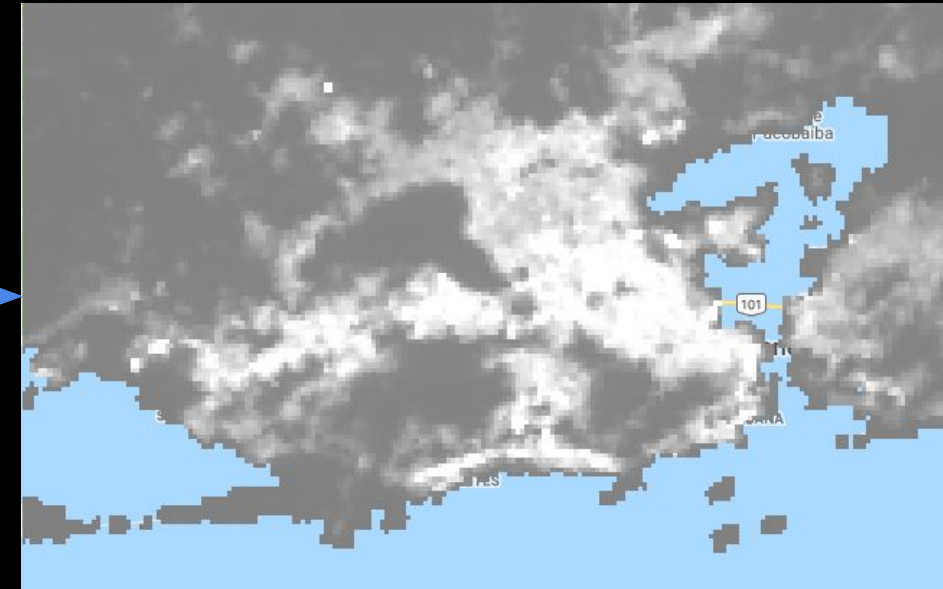
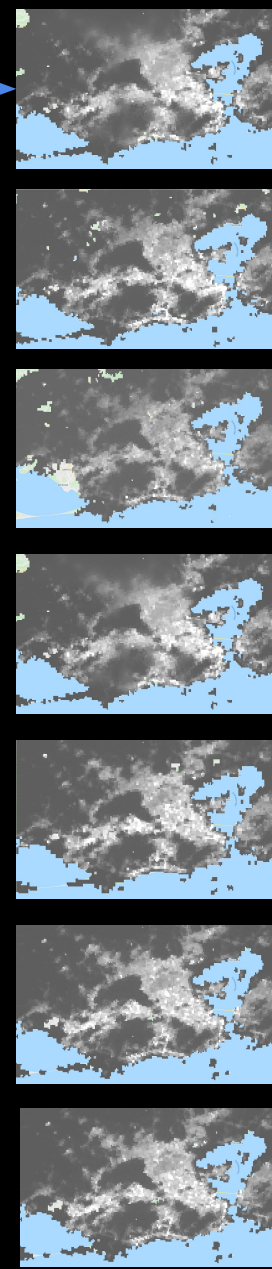
VNP46A2 Raw Image



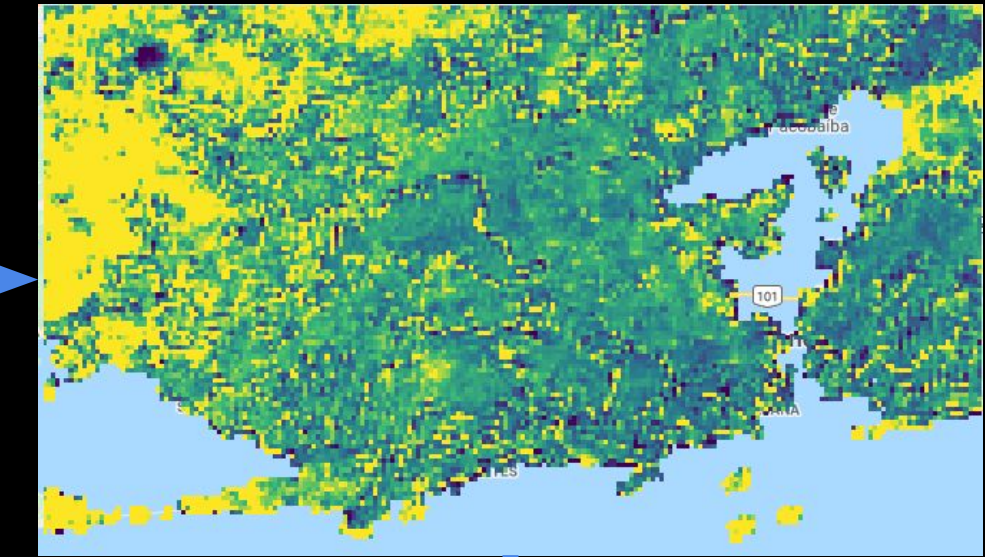
Filter
Area of Interest
Clouds
Water



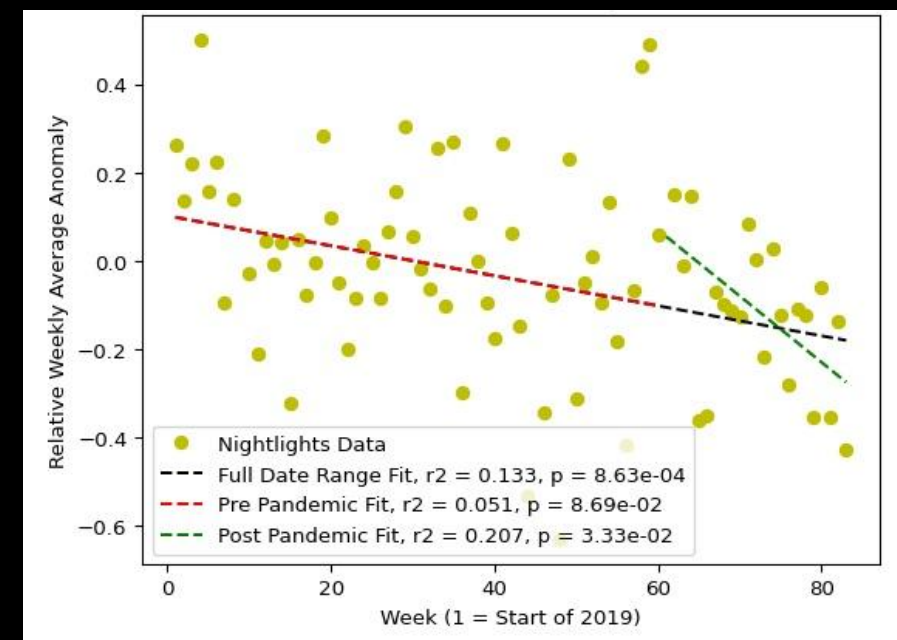
Weekly Averages



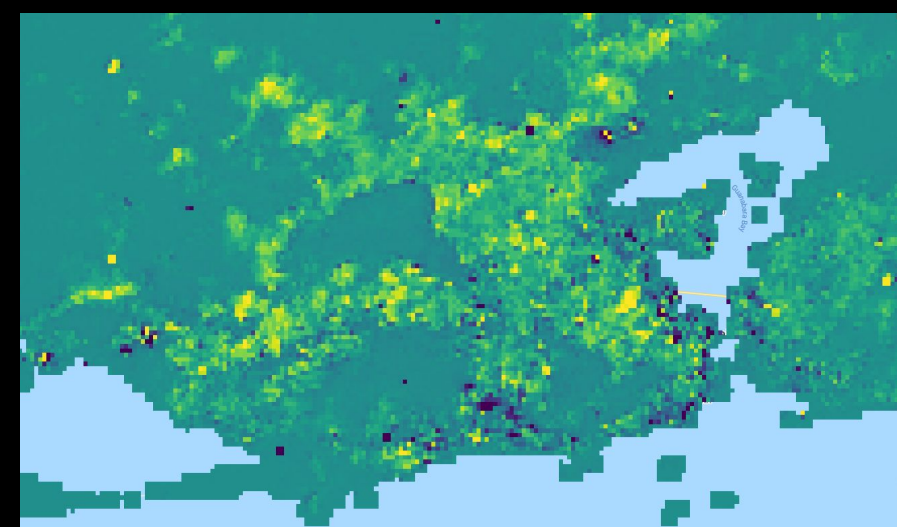
Percent Change Relative to 2019 Annual Average



Statistical Analysis



Theil-Sen Slope Visualization



2019 - Start of Pandemic

Start of Pandemic - 1/Aug/21

Split Data into Temporal Categories

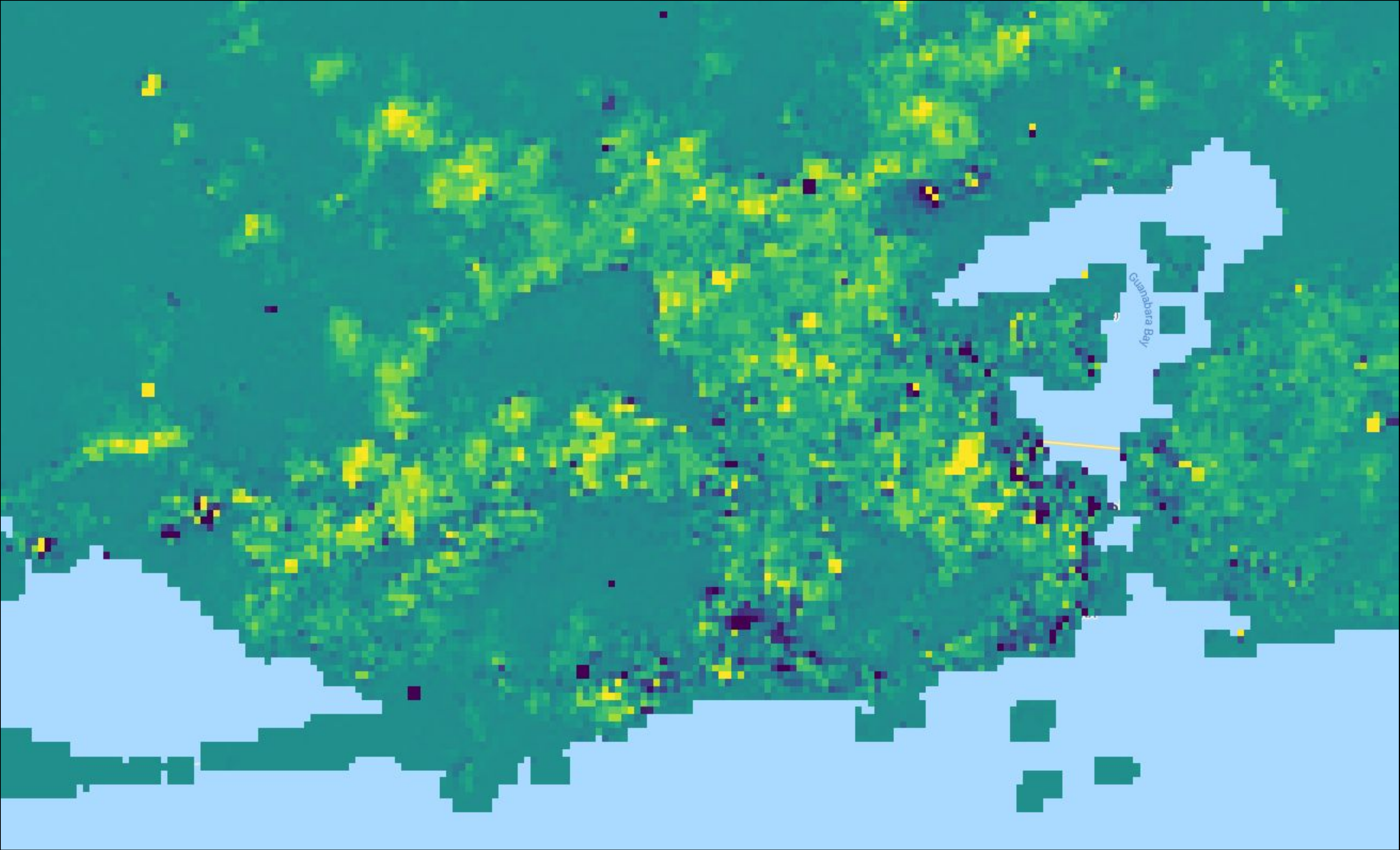
Select specific geographic subunits



Jack Reid

Graduate Student, MIT Media Lab
Space Enabled research group

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

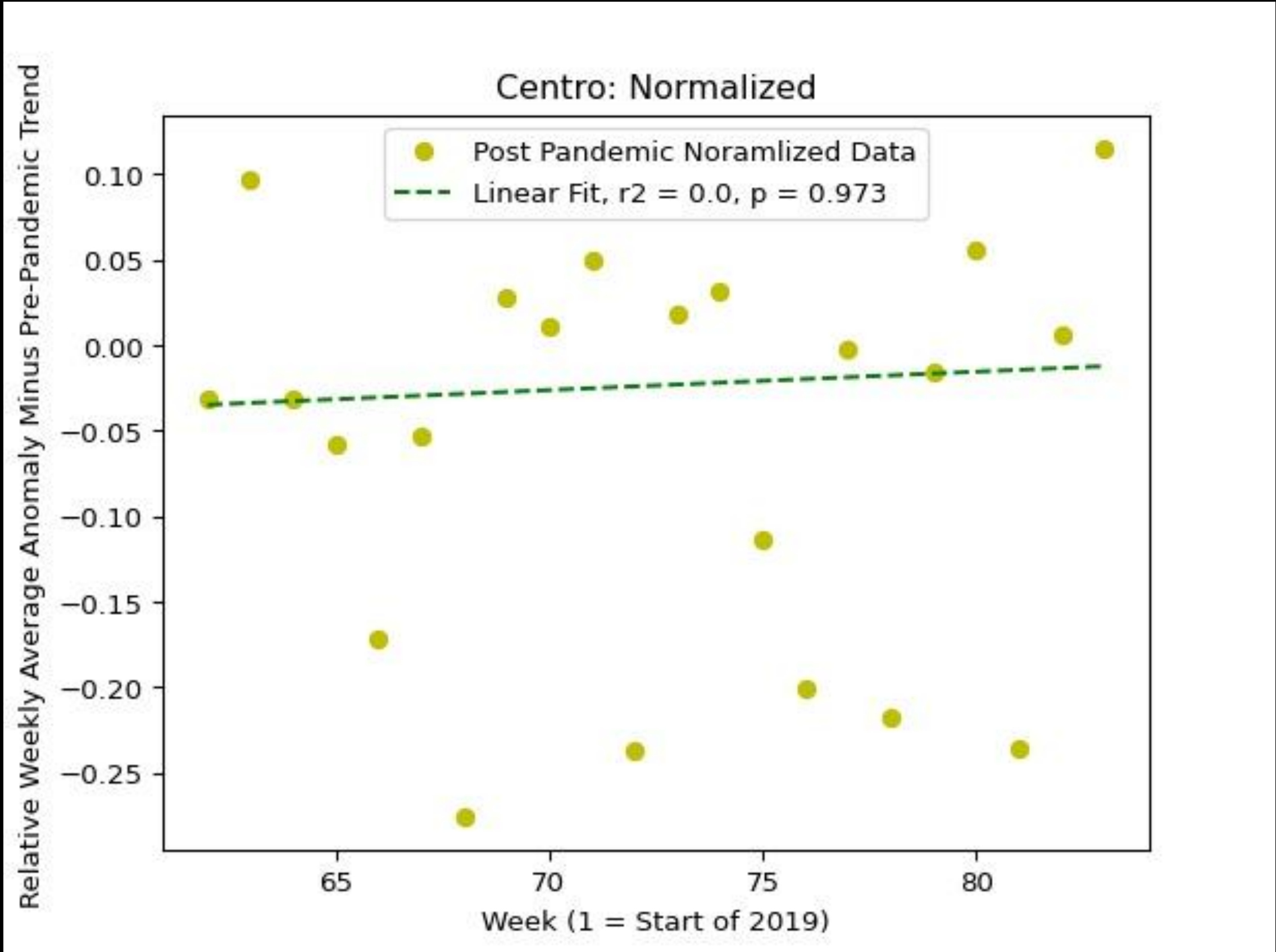
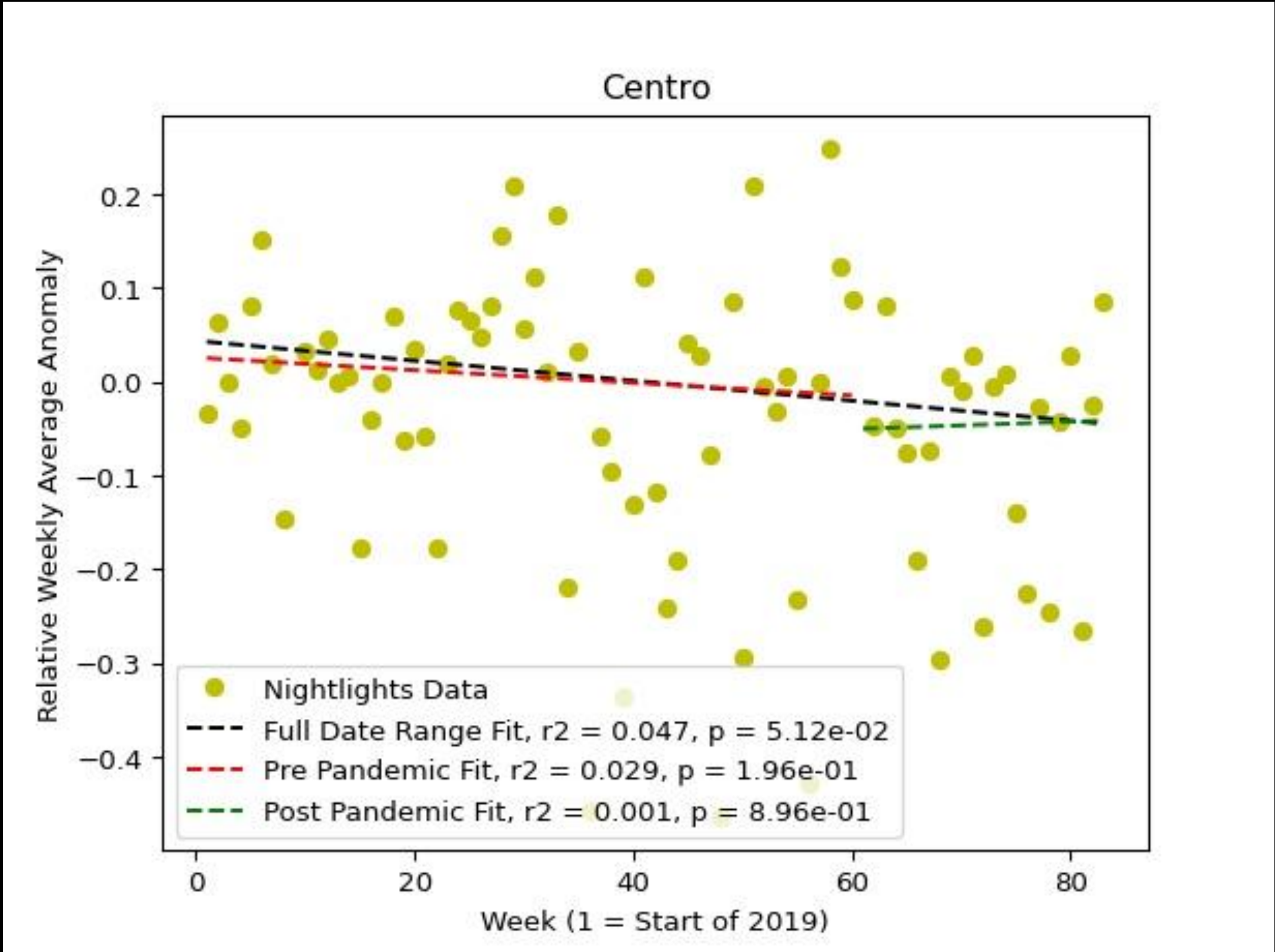


Theil-Sen Slope

- ≥ 2
- $= 0$
- ≤ -2



Nightlights Statistics - Rio Changes



Percent Change Relative to 2019 Annual Average

Normalized:
(Post Pandemic Data - **Red Line Trend**)



Nightlights Rio de Janeiro Changes

Green = Evidence of Change
 Red = No Evidence of Change
 Orange = Darkening Trend
 Blue = Brightening Trend

Darkened

Area	Type	Pre vs Post T-Test P-Value	Normalized Data Linear Fit P-Value	Pre Pandemic Trend (*1000)	Post Pandemic Trend (*1000)
Ipanema	Tourist	0.063	0.08	0.38	-6.00
Barra da Tijuca	Tourist	0.000	0.11	-0.64	-3.73
Santos Dumont Airport	Airport	0.005	0.12	-3.38	-15.00
Galeao Airport	Airport	0.000	0.24	-2.57	-7.22

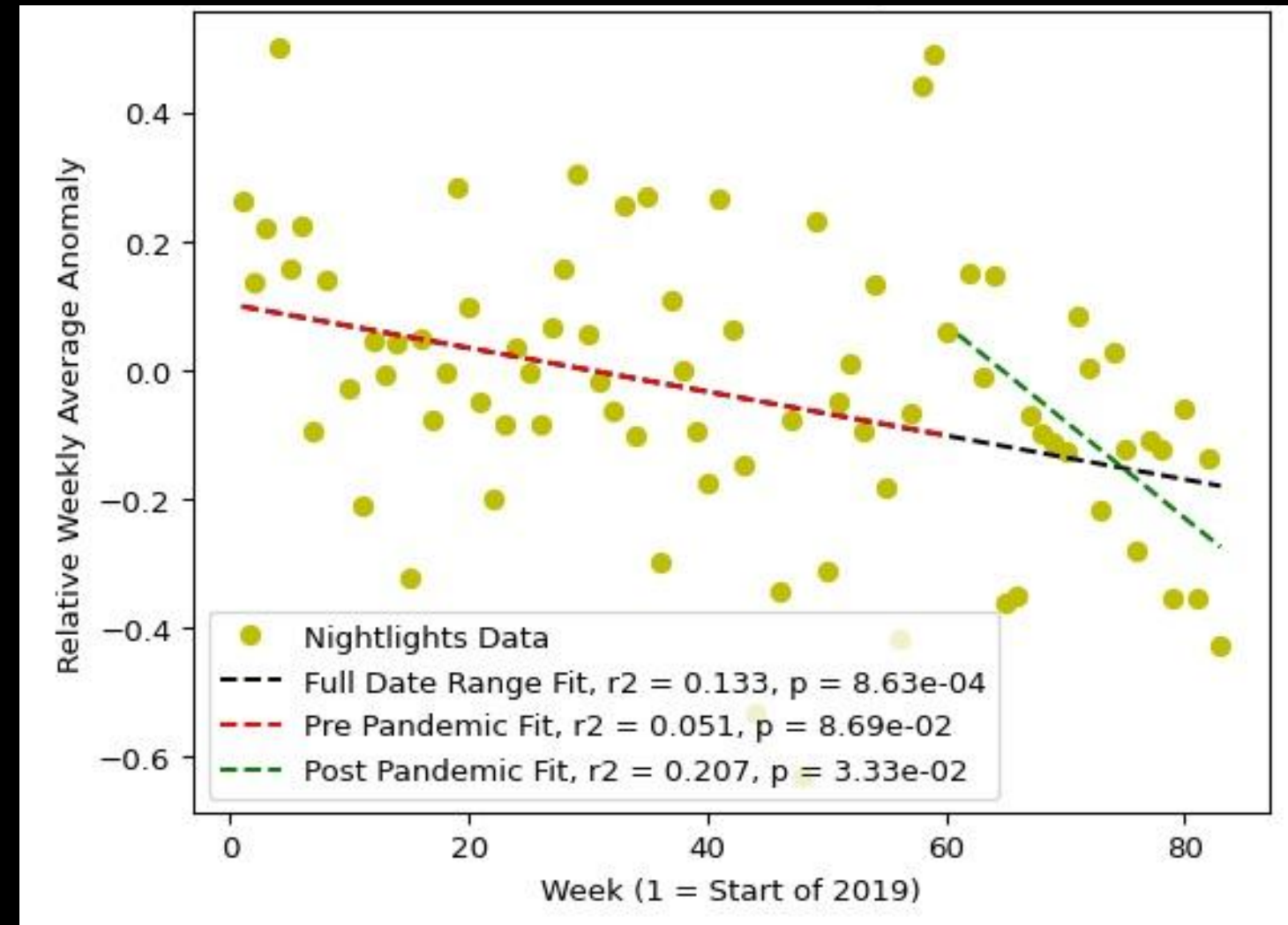
No Change / Ambiguous

Cidade de Deus	Mixed Use / Residential	0.433	0.01	-0.50	6.92
City	Entire City	0.347	0.45	0.58	4.78
Cidade Nova	Downtown	0.604	0.88	-3.76	-3.27
Campo Grande	Suburb	0.503	0.93	0.25	0.62
Centro	Downtown	0.115	0.97	-0.67	0.40
Copacabana	Tourist	0.769	0.90	-1.44	-0.71

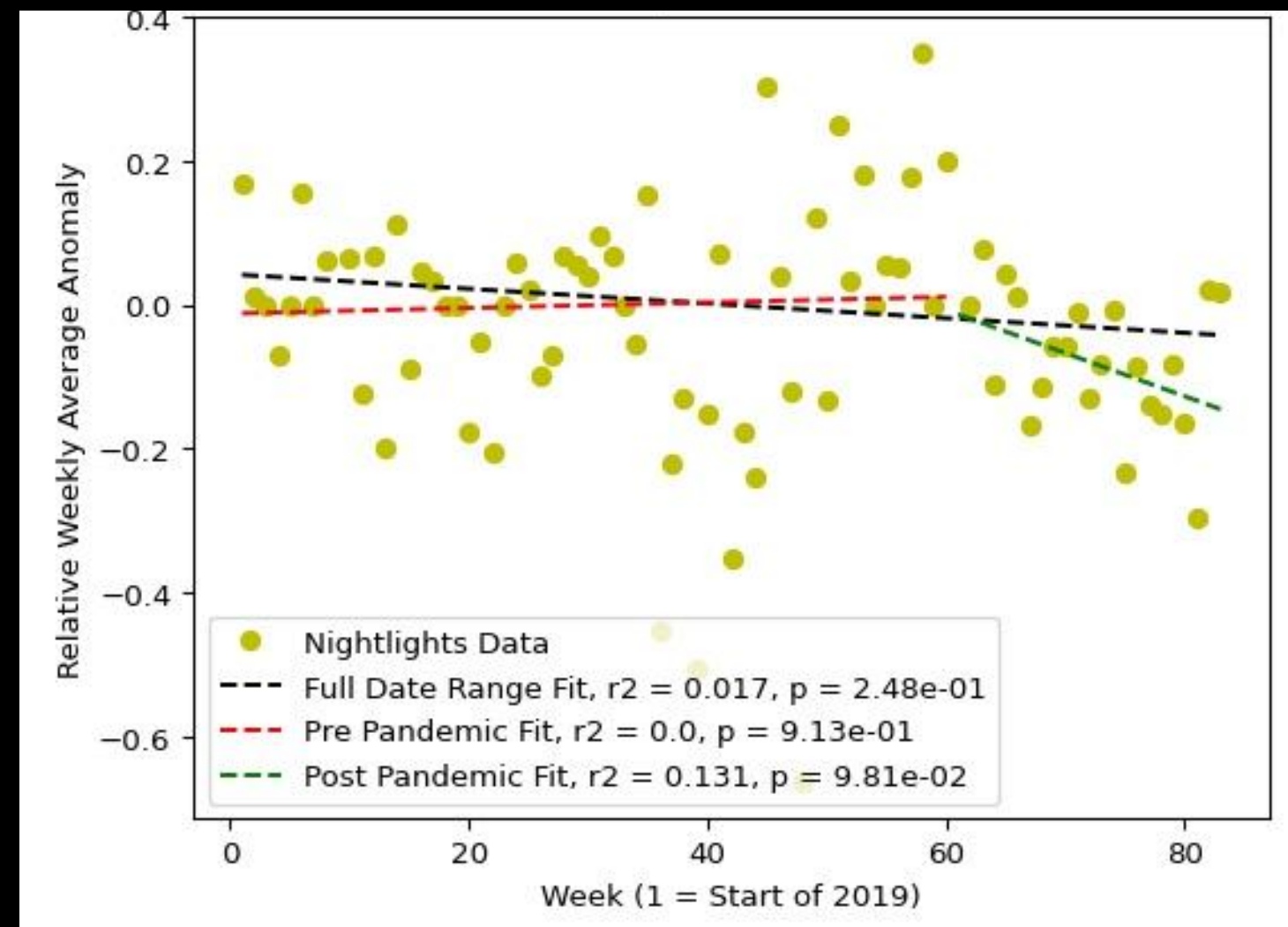


Rio de Janeiro, Brazil

Santos Dumont Airport

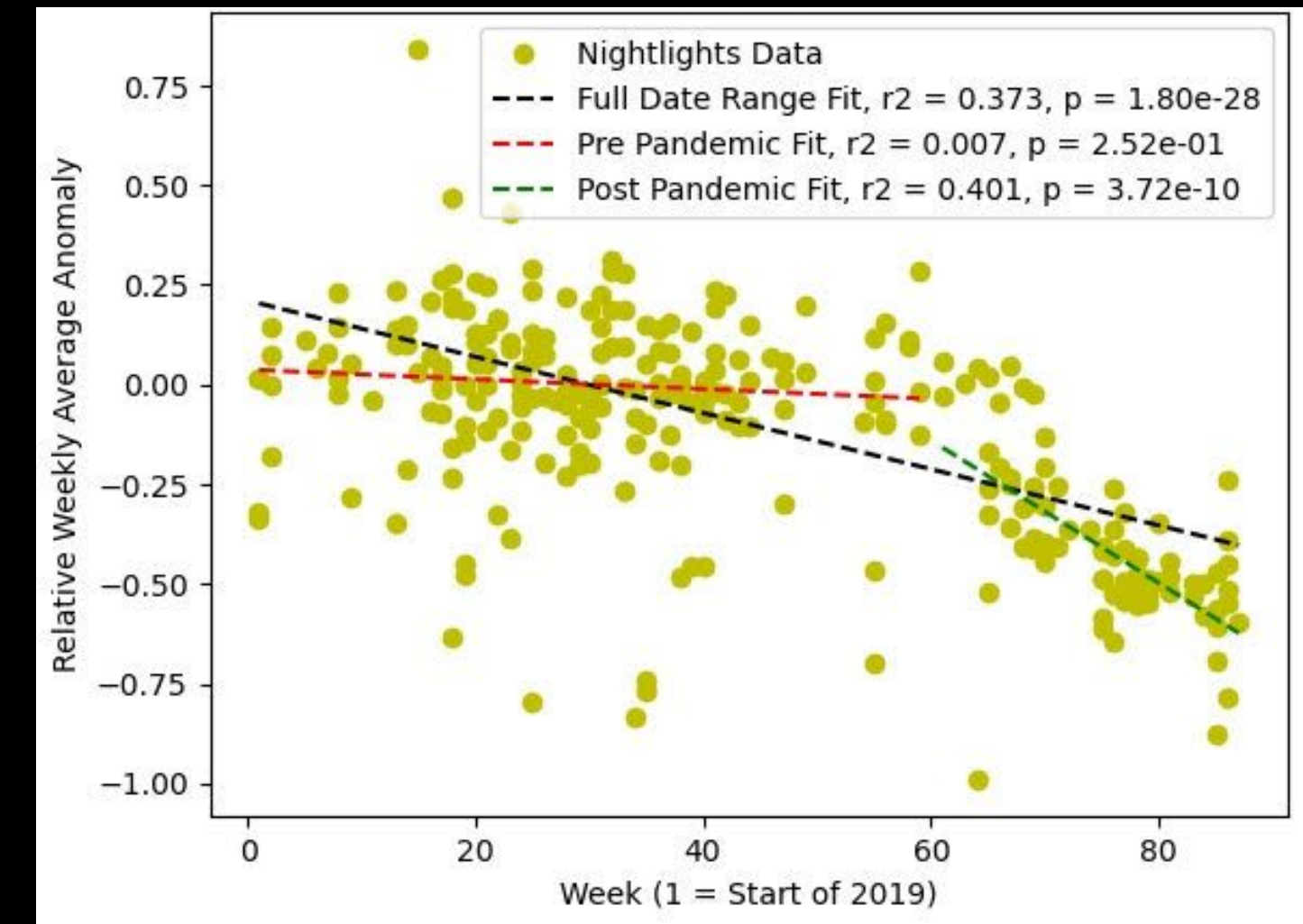


Ipanema

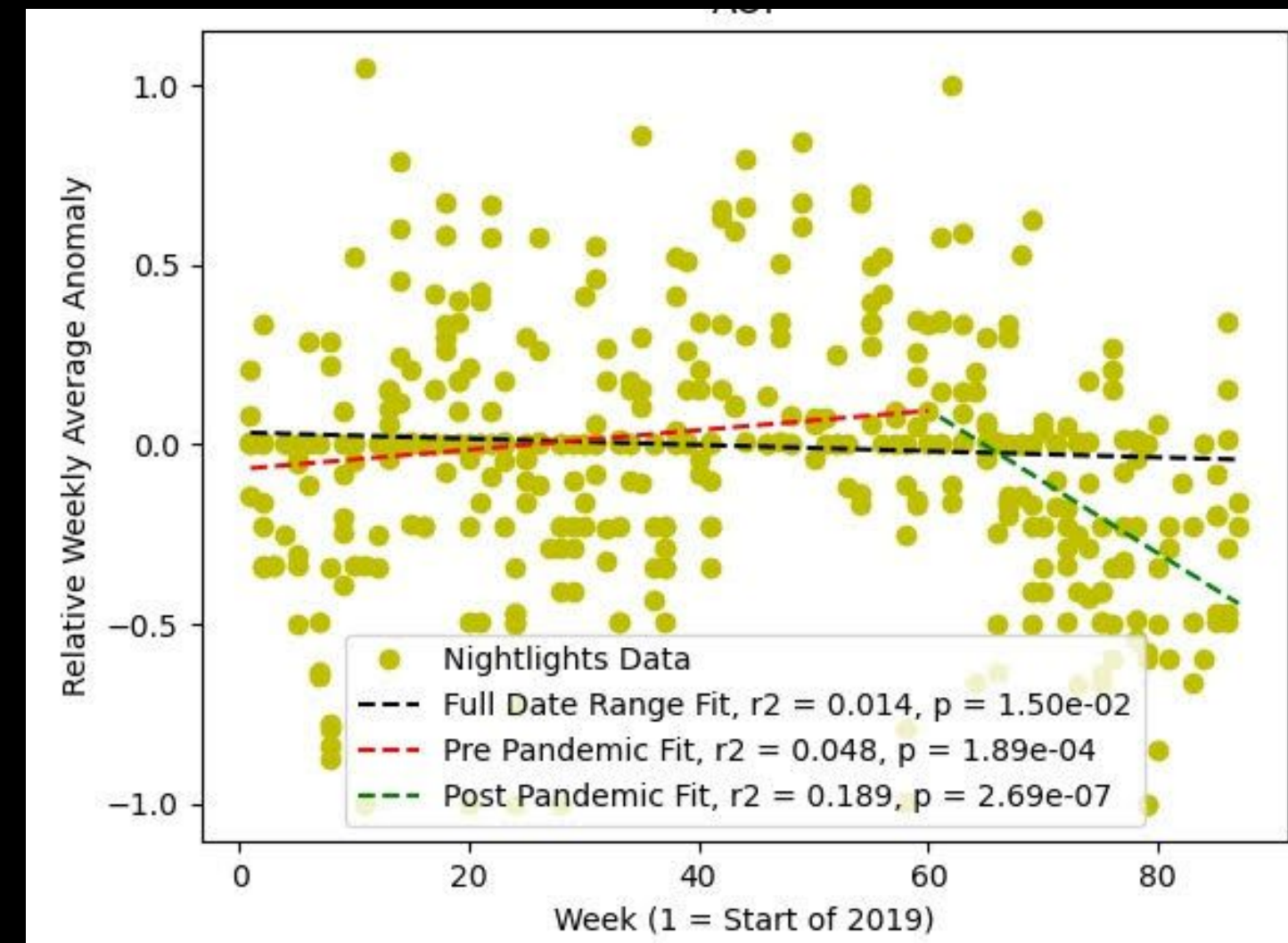


Bali, Indonesia

Ngurah Rai Airport



Island



In-Situ Air Quality Methodology

BRAZIL

Raw Hourly Data from
MonitorAr (Rio de Janeiro)
SINCA (Chile)

Weekly Averages

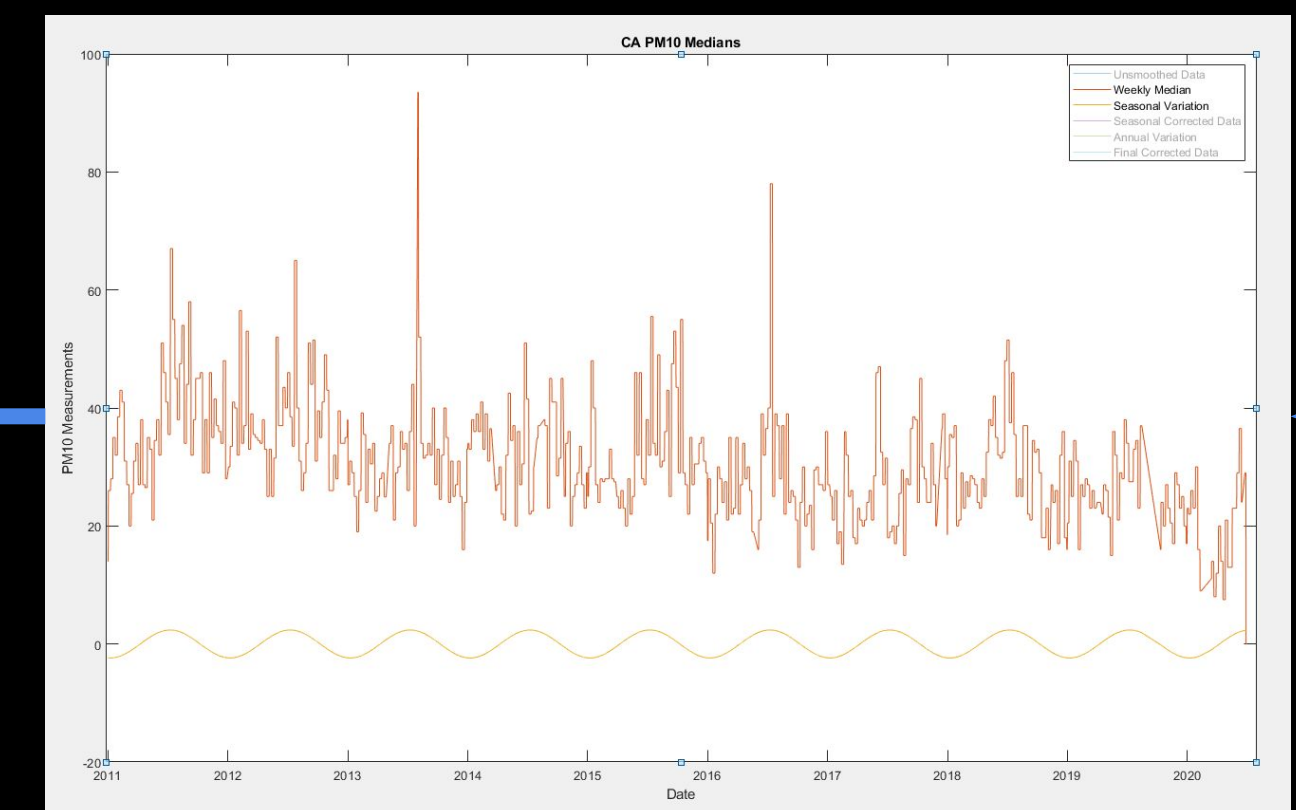
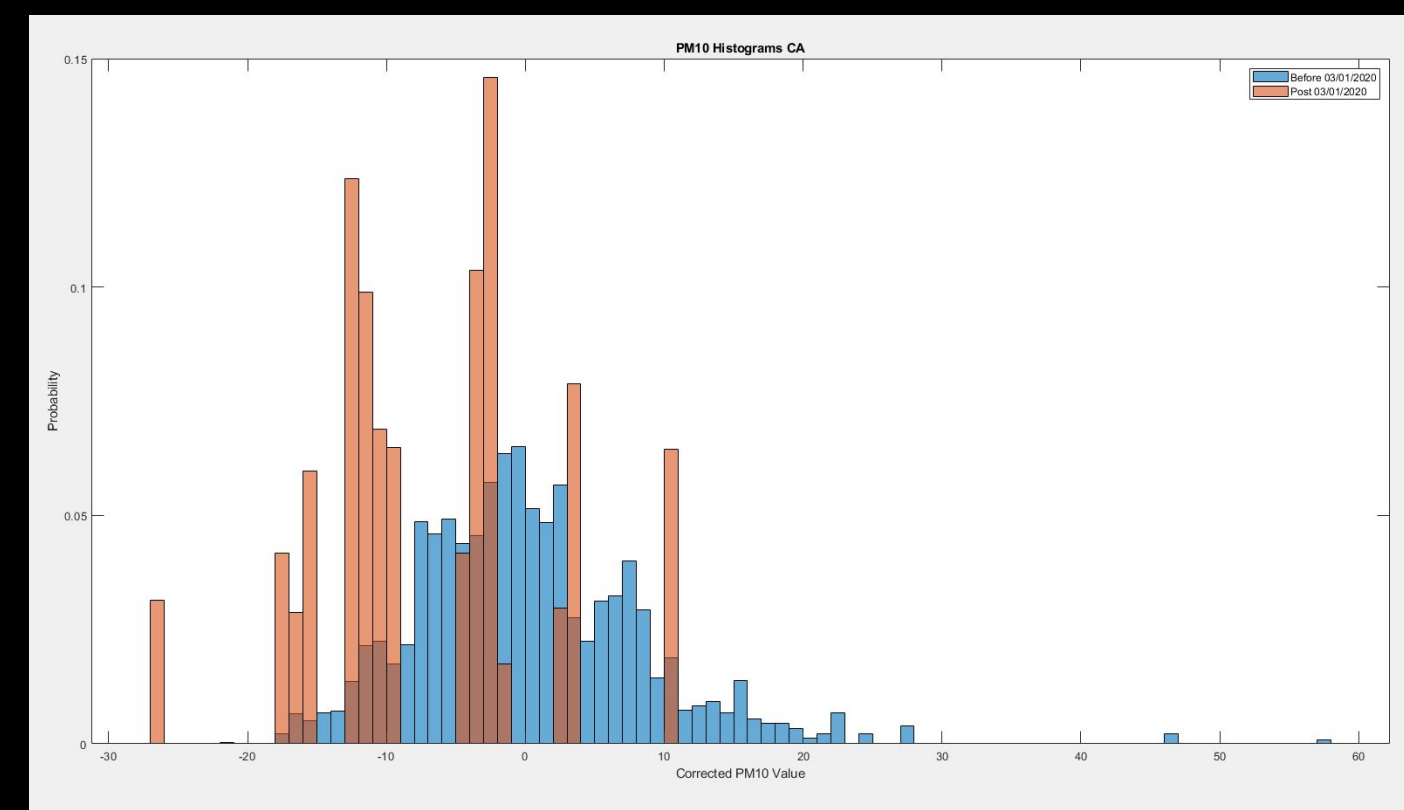
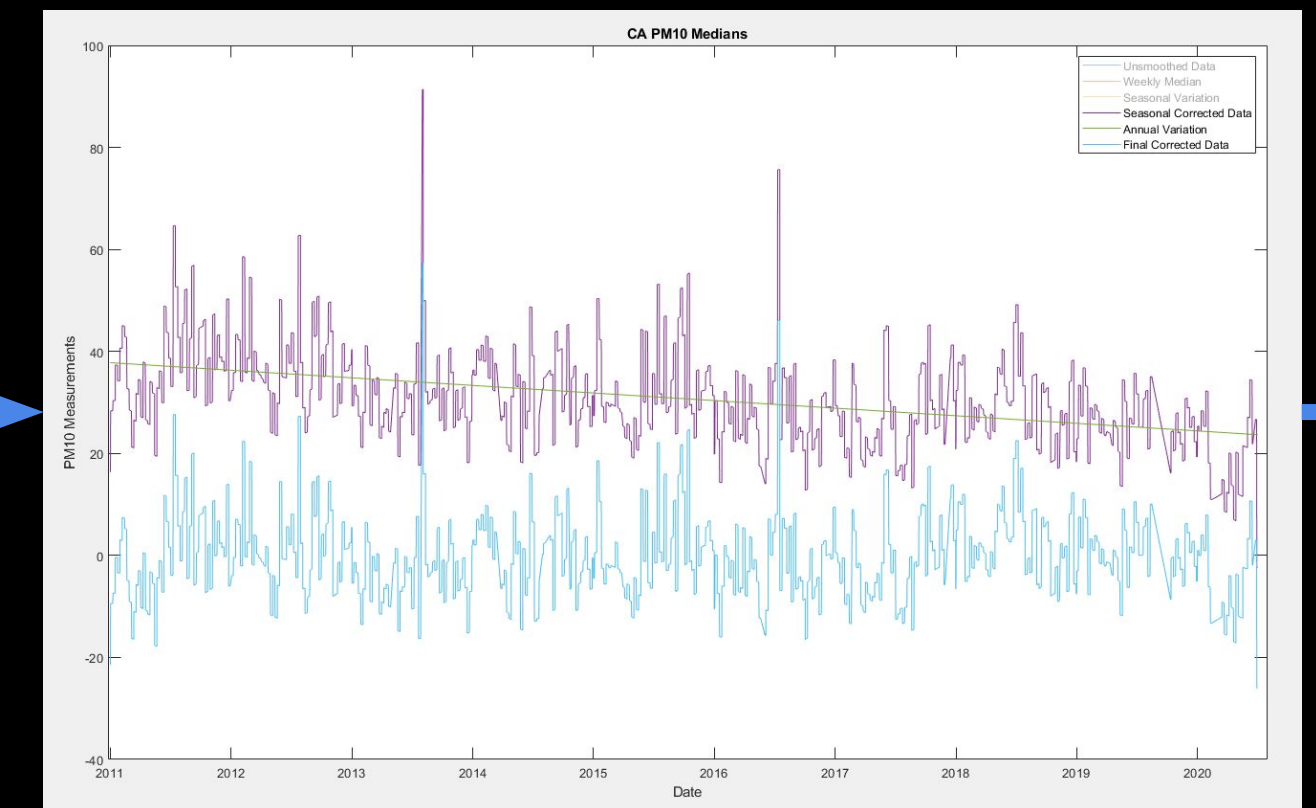
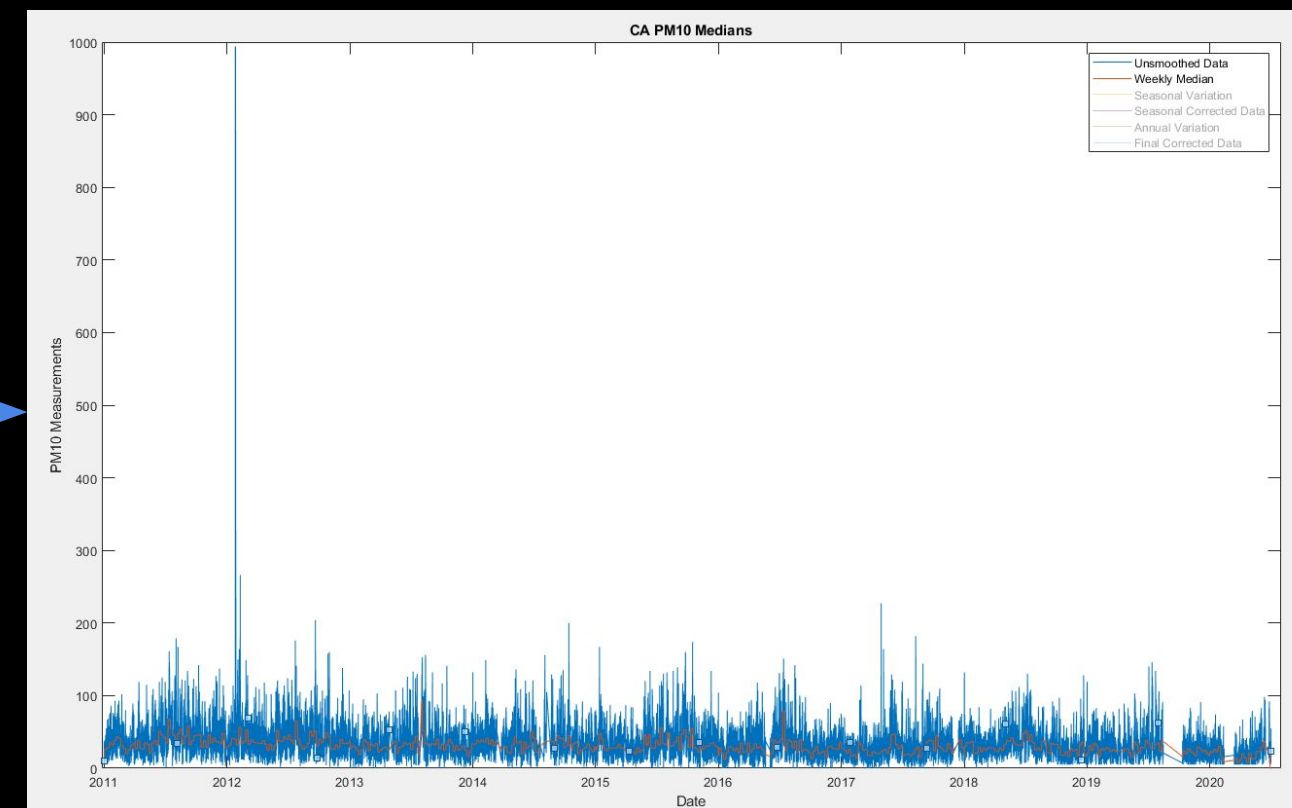
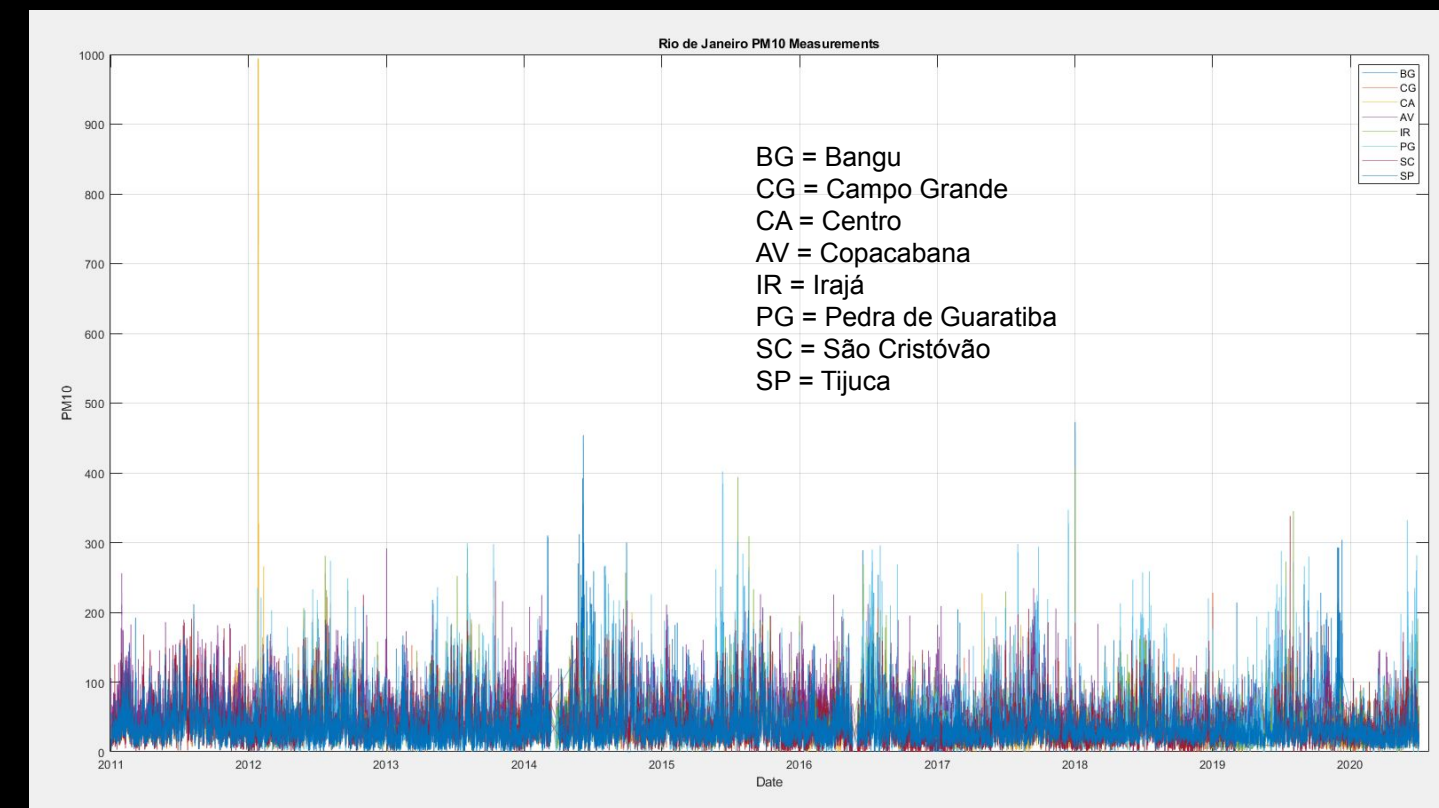
Normalize for secular trends

Statistical Analysis

Normalize for periodic
(seasonal) trends

Jack Reid

Graduate Student, MIT Media Lab
Space Enabled research group



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

Area	Type	Pre vs Post T-Test P-Value	Anderson Darling P-Value	Change in Mean (*1000)
Copacabana	Tourist	0.956	0.14	-3.00
Bangu	Mixed Use / Residential	0.265	0.00	4.20
Centro	Downtown	0.012	0.00	1.38
Campo Grande	Mixed Use / Residential	0.381	0.02	5.10
Irajá	Urban / Residential	0.630	0.00	2.20
Pedra de Guara	Rural	0.784	0.08	1.40
São Cristóvão	Tourist	0.392	0.00	4.10
Tijuca	Mixed Use / Residential	0.084	0.00	9.70

Next Step: Linking with Remote Sensing Data

In-Situ Measurements* (Brazil & Chile)

CH4

CO

Non-Methane Hydrocarbons

Total Hydrocarbons

NO

NO2

NOx

O3

PM2.5

PM10

SO2

Weather (temp, pres., radiance, etc.)

Sentinel 5-P

CH4

CO

HCHO

NO2

O3

SO2

*Not all measurement types are available at all in-situ sites.

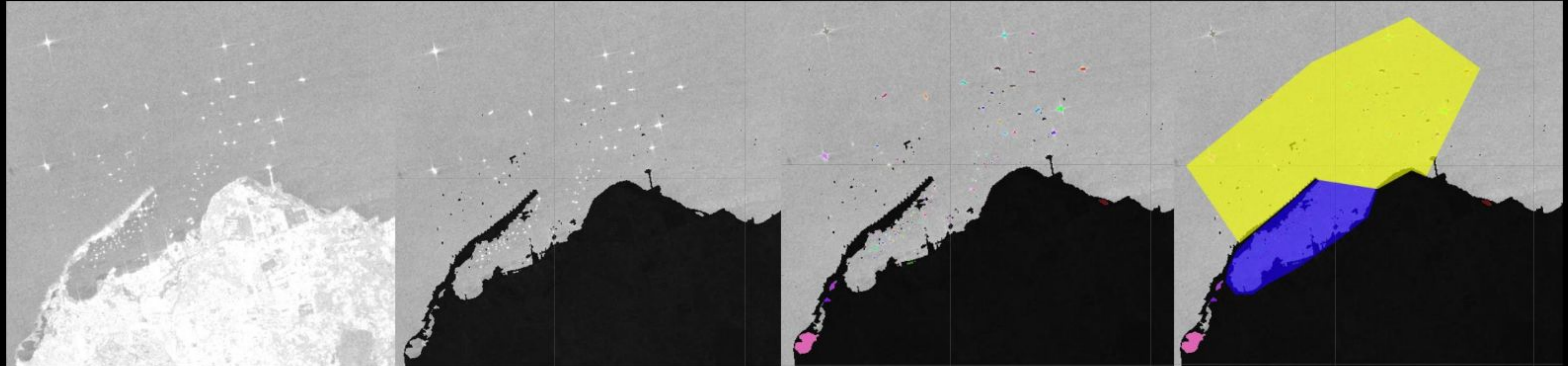


Data & Methods: Vulnerability

- 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)



Ship Tracking Methodology



Sentinel Radar Imagery

- Reference: 2018 & 2019
- Observation: 2020

Mask out land/ permanent structures

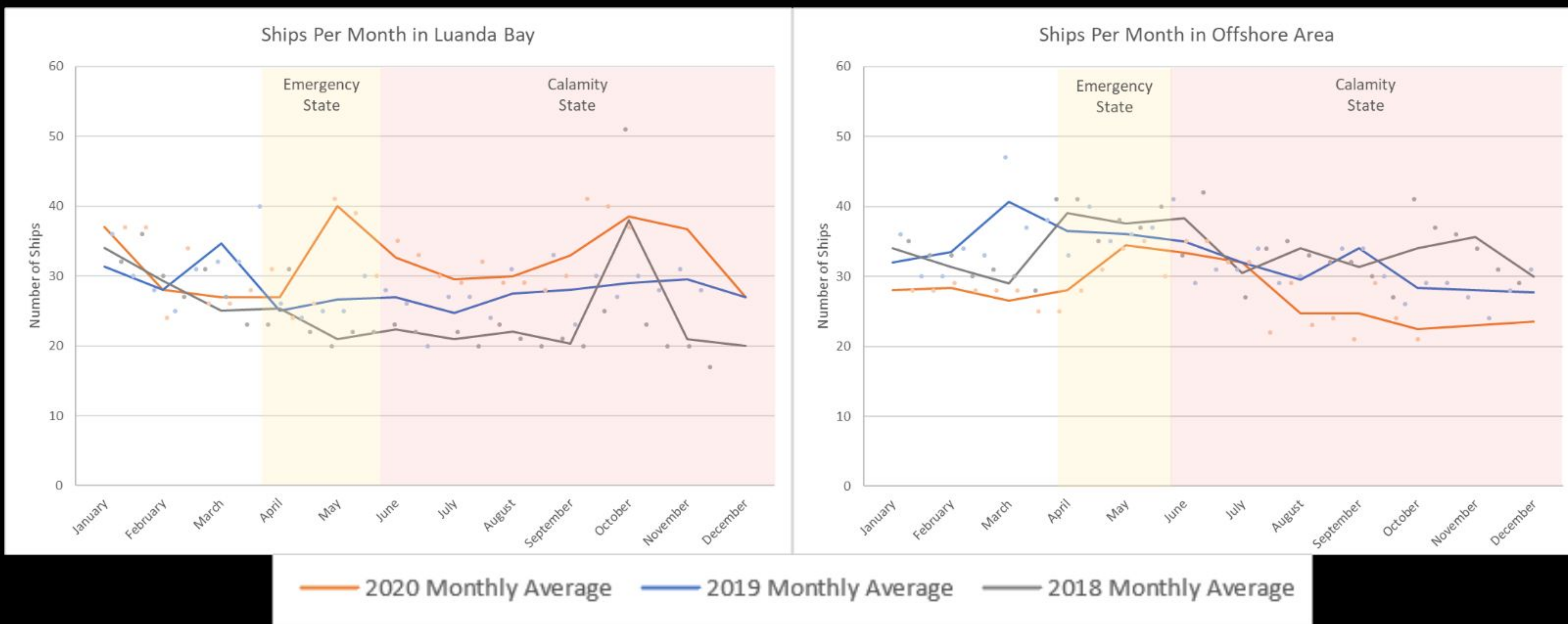
Identify the individual ships

Count the number of ships in and outside the bay

Images and analysis done by Amanda Peyton

Ship Tracking Results

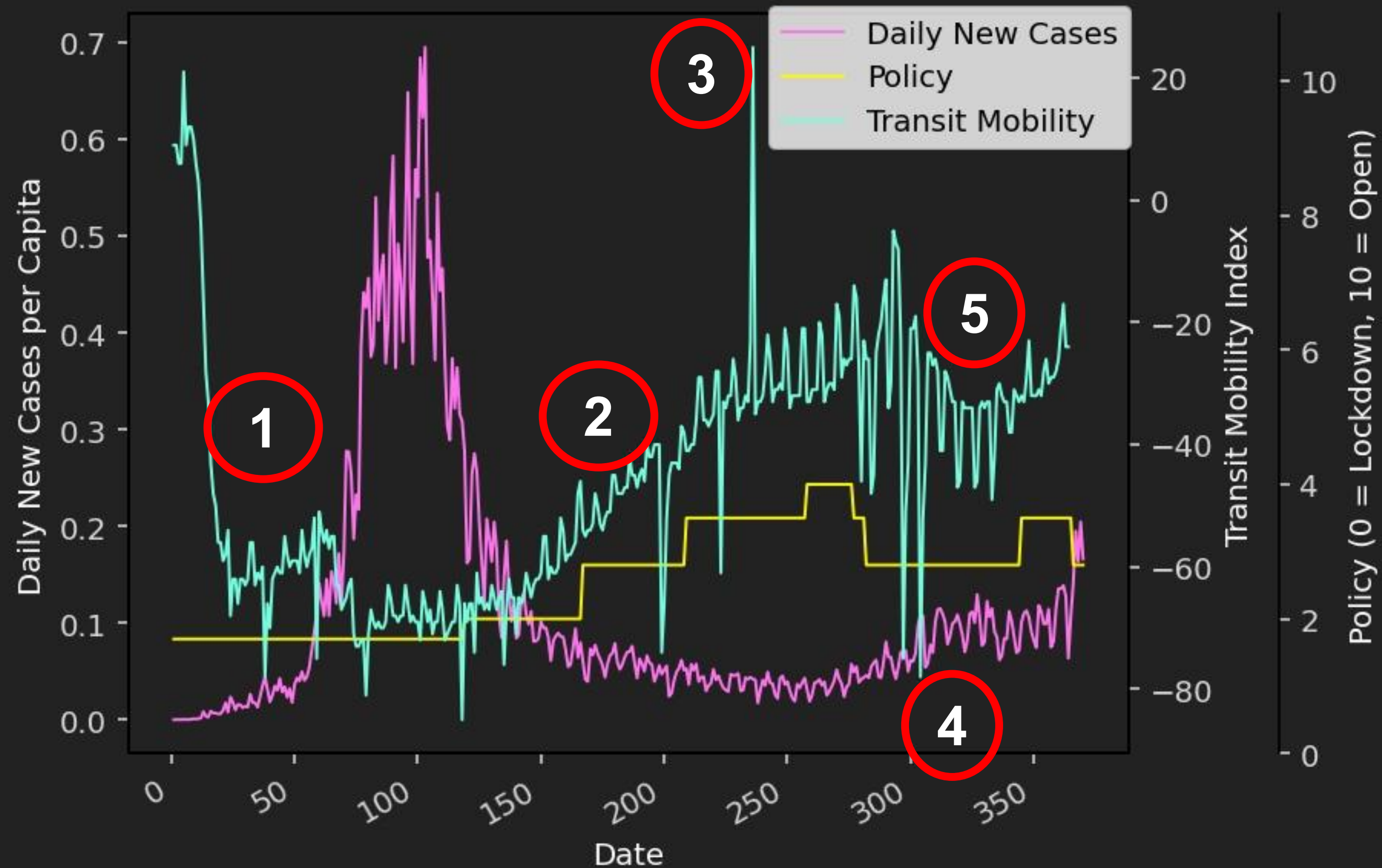
Measuring Ship Presence Over Time



Images and analysis done by Amanda Peyton

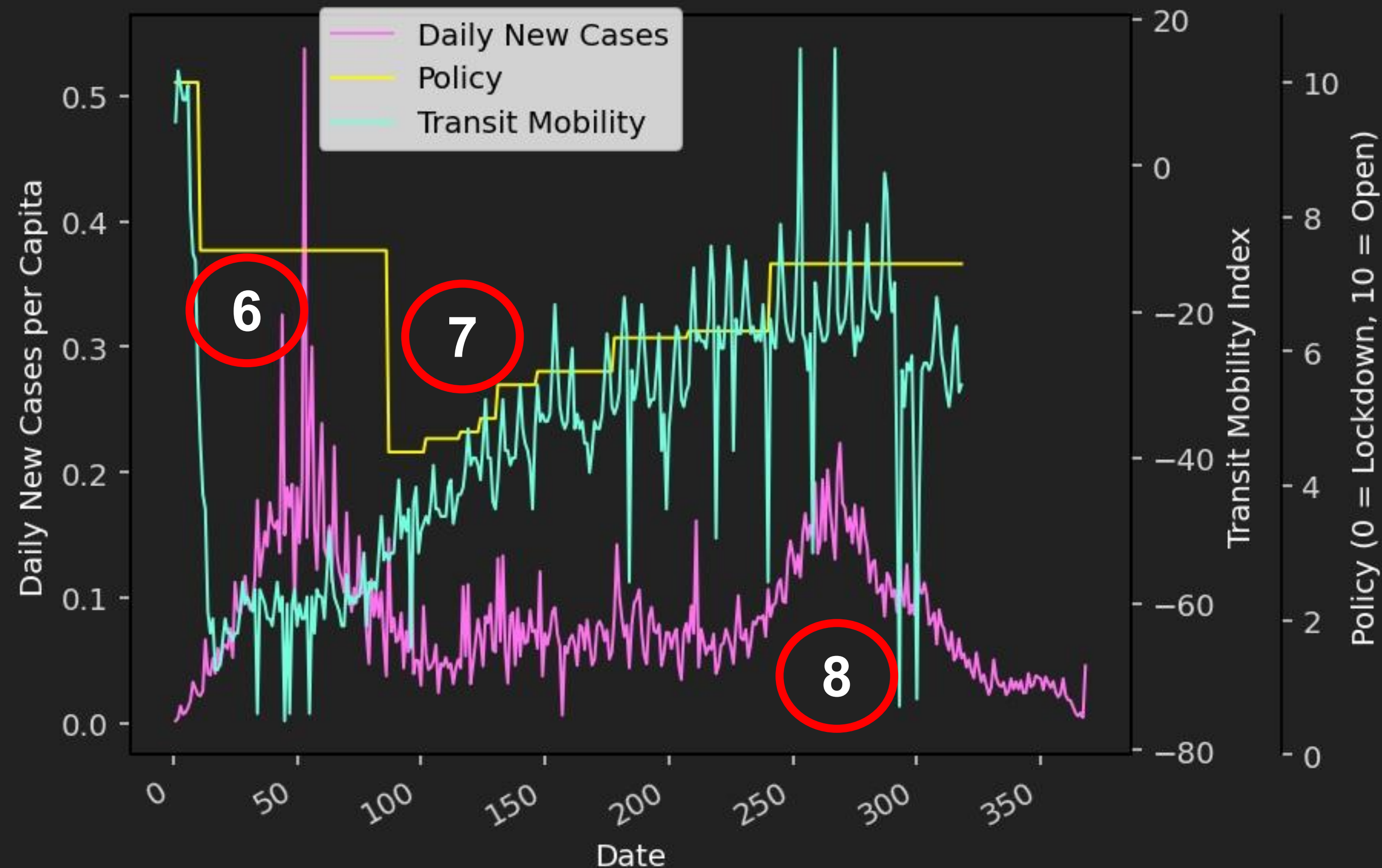


Ex) Metropolitana, Chile Mobility Changes



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

Ex) Rio de Janeiro Mobility Changes



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.

Data & Methods: Decision-making

- COVID-19 Social Distancing Requirements & Closures
 - Announcements, histories, definitions, and conditions created by local authorities
 - Ongoing effort to compare policies using standardized, quantitative comparisons based on the CoronaNet Research Project



Data & Methods: Decision-making

Recovery Plan Indicators updated 01/10/2020

Reference Date: < 07/29/2020 >

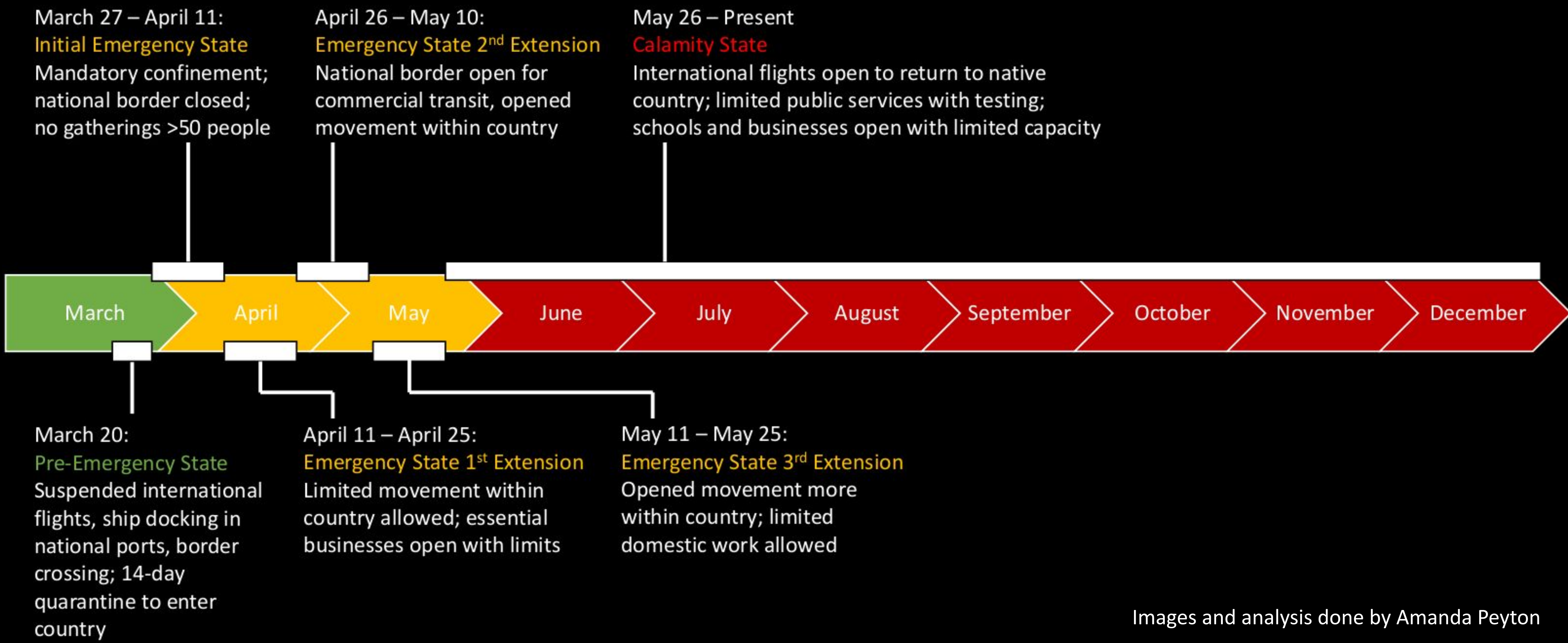
GROUP	ANALYSIS PARAMETERS	PRIMARY INDICATORS	Comparison with previous days						07/16/2020	7/29/2020	WE ARE IN PHASE 6B (Since 01/10/2020)					
			F-1	D-5	D-4	D-3	D-2	D-1	Ref. Previous Phase	Result	PHASE 1	PHASE 2	PHASE 3	PHASE 4	PHASE 5	PHASE 6
HEALTH SYSTEM RESPONSE CAPACITY	Capacity of ICU beds	1 Percentage of occupancy of dedicated adult ICU beds COVID (ICU SRAG) METRO I SUS bed (7-day moving average)	✗	✓	✓	✓	✓	✓	69.4	71.2	Favorable	Favorable	Favorable	Favorable	Favorable	Favorable
		2 Occupancy rate of supplementary sector ICU beds (moving average 7 days) (a)	✗	✗	✗	✗	✗	✗	67.9	70.0	Favorable	Favorable	Favorable	Favorable	Favorable	Favorable
		3 Percentage of occupancy of life support beds REDE SUS Territory of the municipality (moving average 7 days)	✗	✓	✓	✓	✓	✓	76.0	77.0	Favorable	Favorable	Favorable	Favorable	Favorable	Favorable
		4 ICU COVID beds (REDE SUS) per 100k inhabitants (b)	✗	✗	✗	✗	✗	✓	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)	✗	=	✗	=	✓	✓	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)	✗	✓	✓	✓	✓	✓	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)	✓	✓	✓	=	=	=	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

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



Data & Methods: Decision-making

Timeline of 2020 COVID-19 Response in Angola



Images and analysis done by Amanda Peyton



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, connect to mobility data



Project Page:

<https://www.media.mit.edu/projects/vida-decision-support-system/overview/>

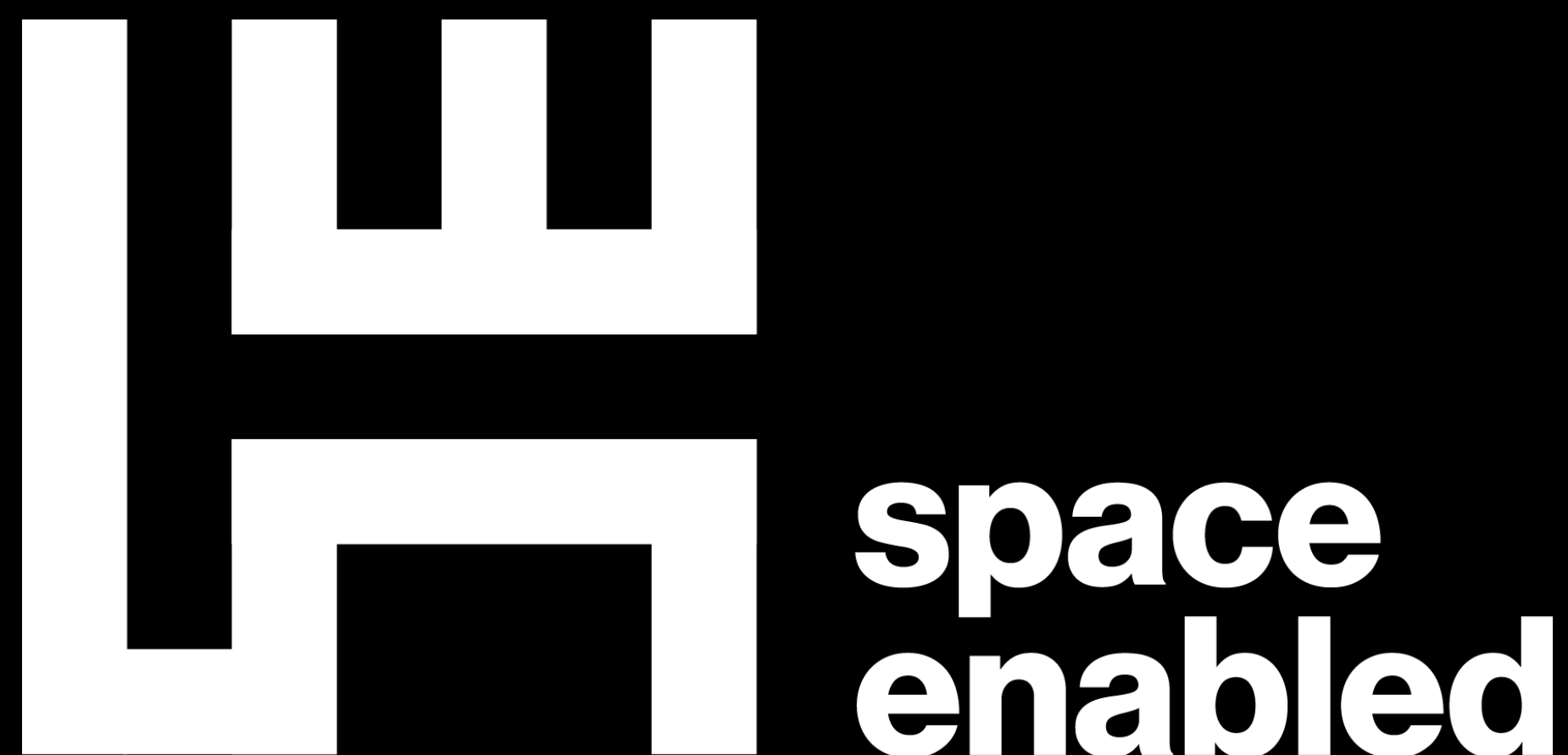
Code Repository:

https://github.com/mitmedialab/Vida_Modeling

Contact Information:

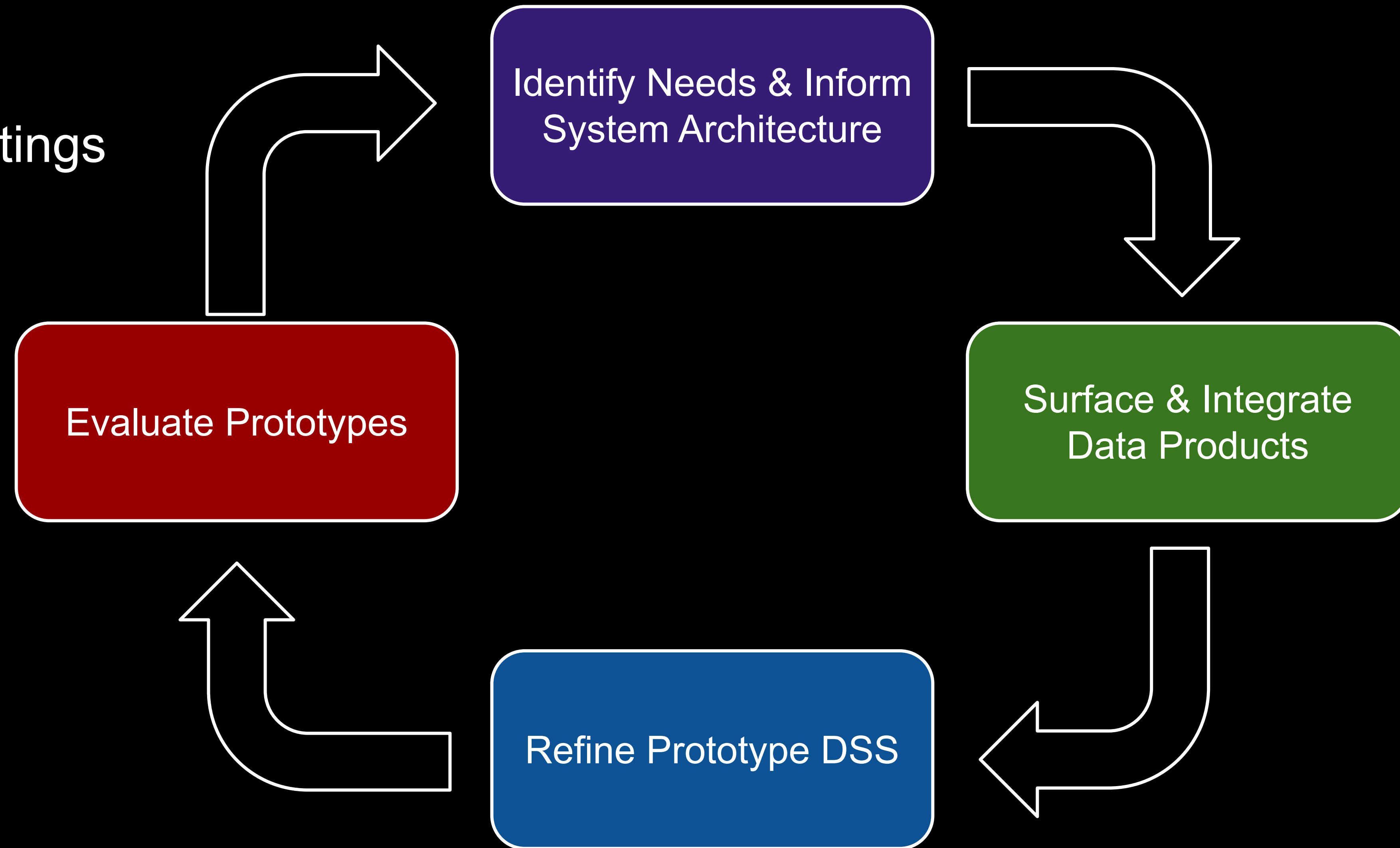
jackreid@mit.edu

https://twitter.com/Jack_B_Reid



Stakeholder Involvement

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



External Context: The COVID-19 pandemic and related societal factors

Inputs

System Boundary

Outputs

Constraints or Opportunities:

- 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 government 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

System Objectives

- Proof-of-concept for integrated data visualization and modeling tool
- Collaborators will use this version as a basis for developing their own, locally managed versions.

Emergent Properties:

- Understanding of the relationships between the pandemic's effects on Public Health, the Environment, Socioeconomic Factors, Public-Sector decision making, and Technology Design
- DSS accessible to decision makers without technical expertise in modeling and data analysis

Allocate

Express

Execute

Meet

System Forms

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

System Functions

- Visualize data from five integrated models
- Simulate different potential policy scenarios

Transform