

# An evaluation of Earth Observation as a potential tool to forecast and manage resources during the Covid-19 pandemic

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# C-19 pandemic

## COVID-19

Disease caused by the SARS-CoV-2 virus



### Novel coronavirus

Coronaviruses are viruses that circulate among animals but some of them are also known to affect humans.

The 2019 novel coronavirus was identified in China at the end of 2019 and is a new strain that has not previously been seen in humans.



### Prevention

#### When visiting affected areas

Avoid contact with sick people



Wash your hands with soap and water



If you develop cough, use a medical face mask



Wherever you travel apply general hygiene rules

### Symptoms

FEVER

COUGH

DIFFICULTY BREATHING

MUSCLE PAIN

TIREDFNESS

### Transmission

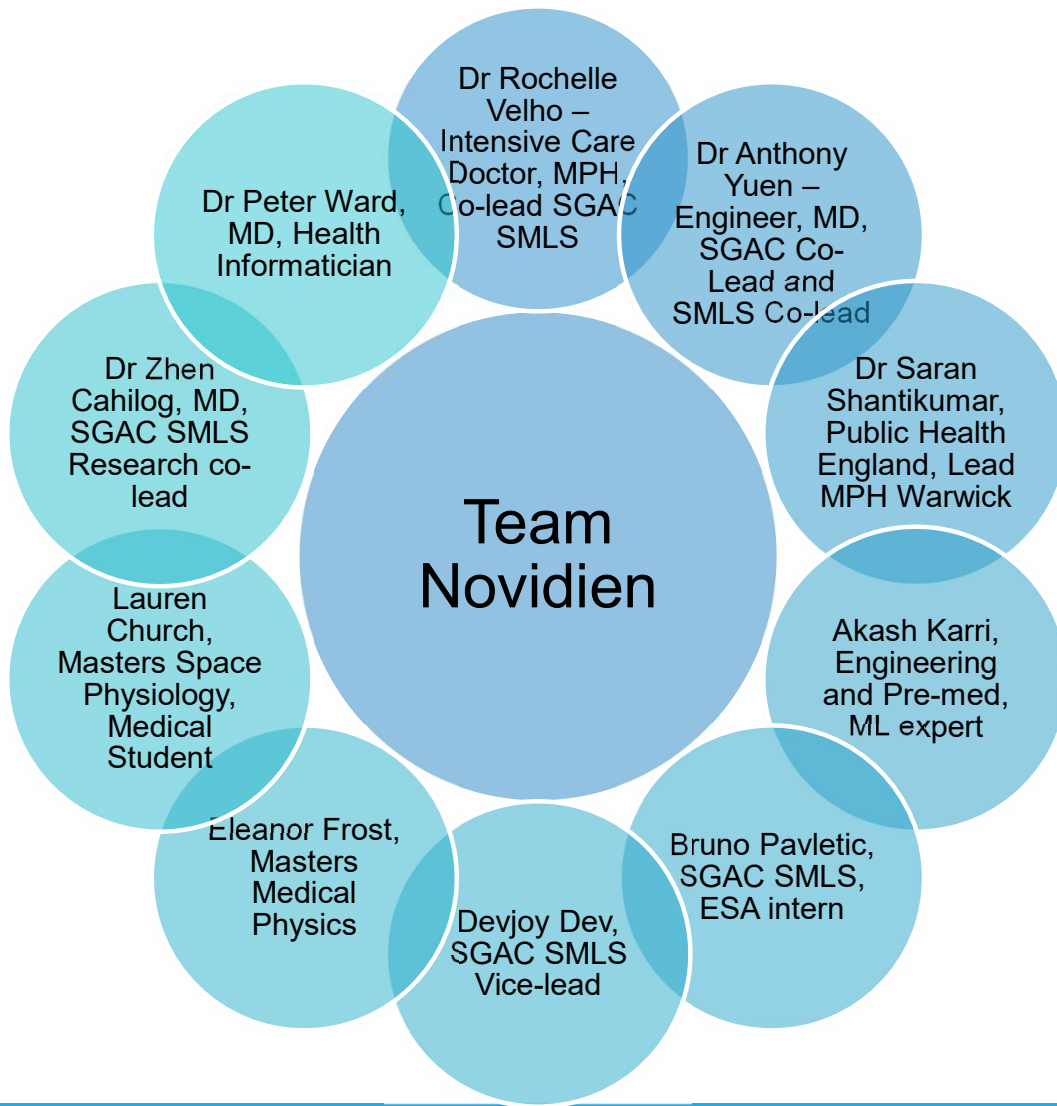
VIA RESPIRATORY DROPLETS

2-14 days  
estimated incubation period



[ecdc.europa.eu/en/novel-coronavirus-china](https://ecdc.europa.eu/en/novel-coronavirus-china)

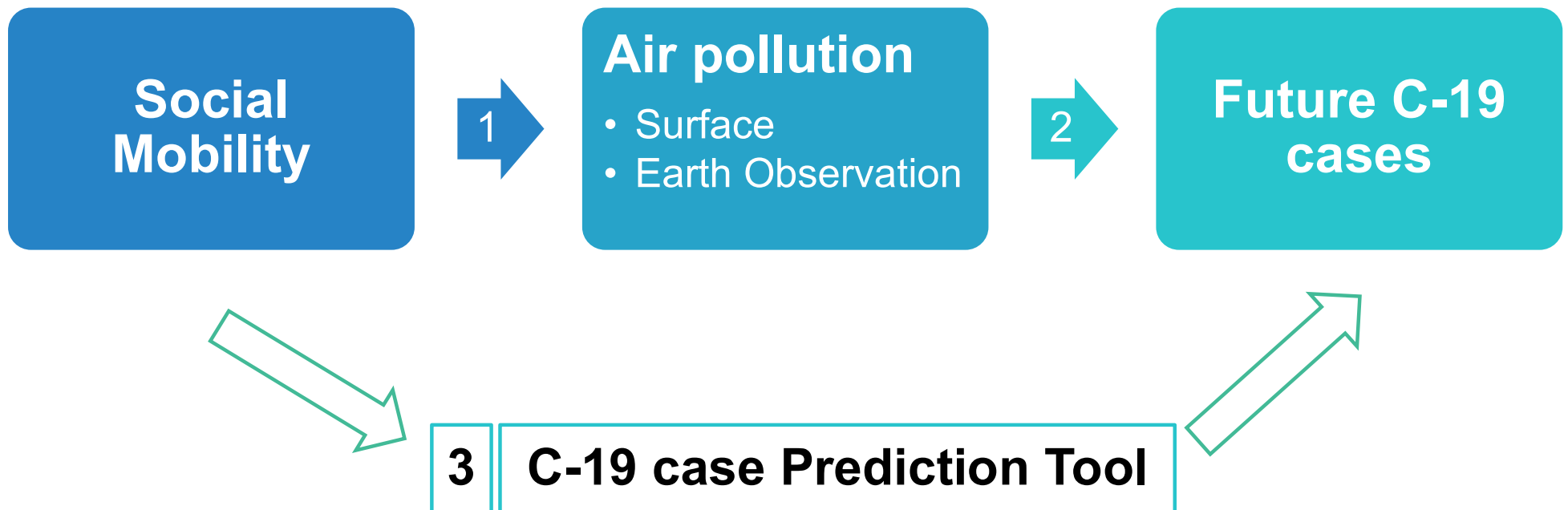




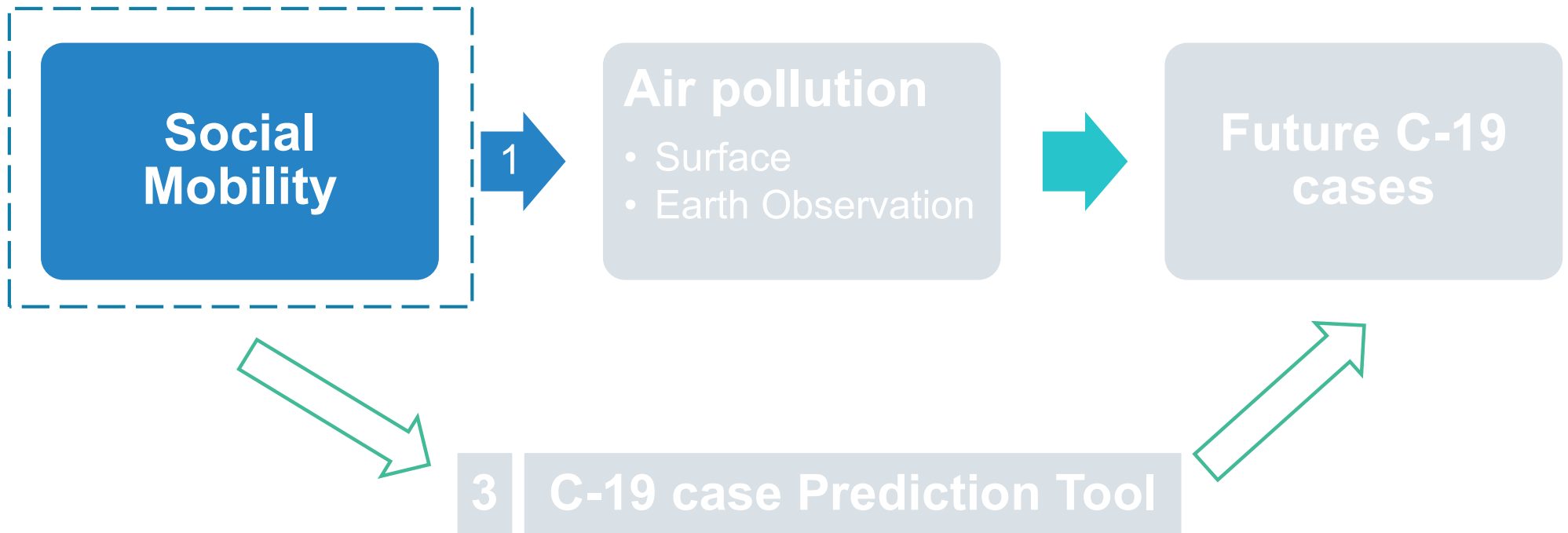
*Additional affiliations:*



# Can Earth Observation data be used to forecast the number of C-19 cases in an area?



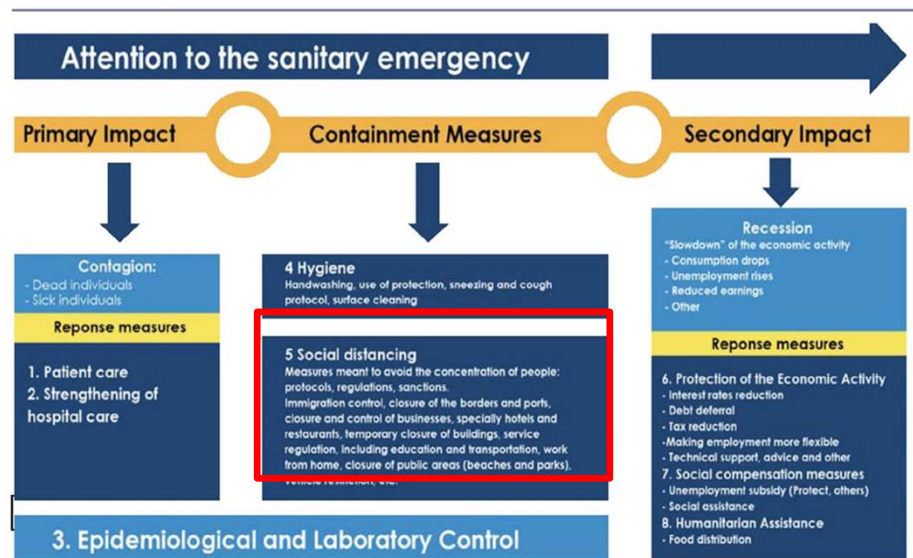
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# Social distancing minimizes C-19 transmission

Figure 2  
COVID-19 impact and care actions



UN/DESA Policy Brief #85: Impact of COVID-19: perspective from Voluntary National Reviews

- Part of the public health solution for C-19
- Lockdown and uplift timelines geographically different
- Public health guidance different across borders
- Public and workplace adherence measures variable

# Mobile phone mobility data indirectly measures social mobility

Article | Open Access | Published: 18 February 2021

## Mining Google and Apple mobility data: temporal anatomy for COVID-19 social distancing

Corentin Cot, Giacomo Cacciapaglia & Francesco Sannino

Scientific Reports 11, Article number: 4150 (2021) | Cite this article

1754 Accesses | 1 Citations | 4 Altmetric | Metrics

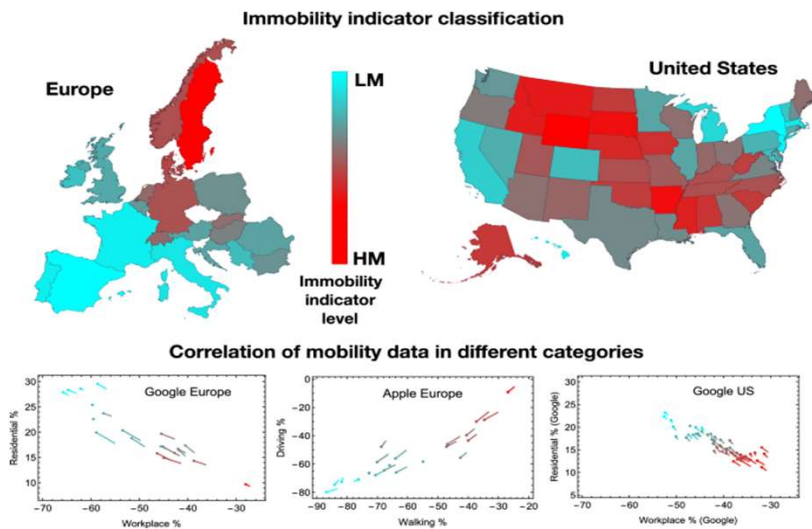
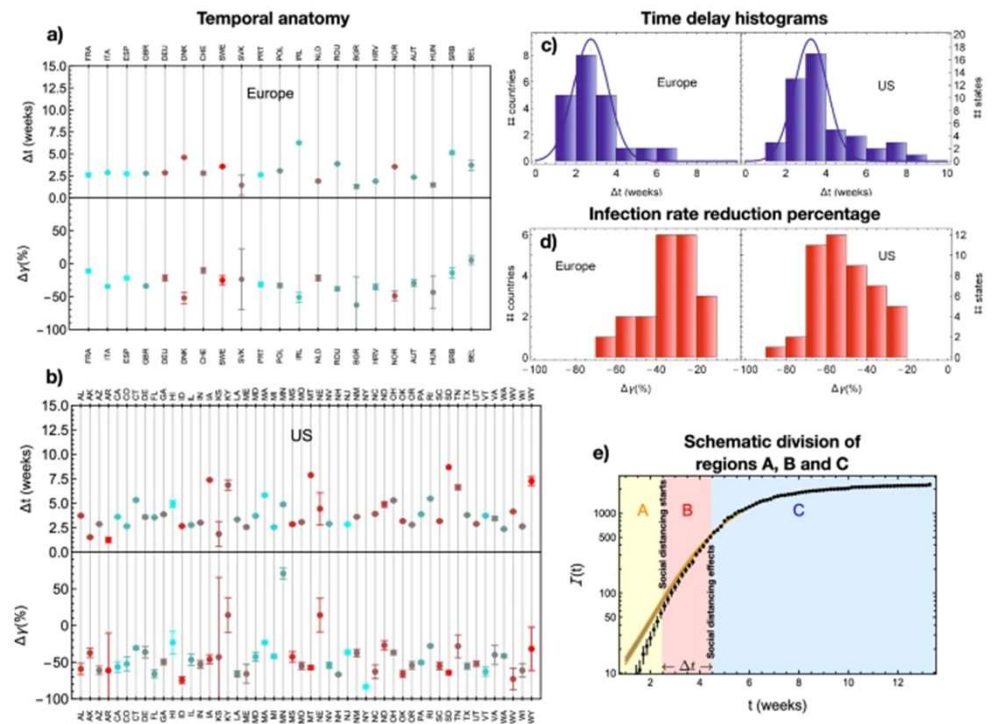
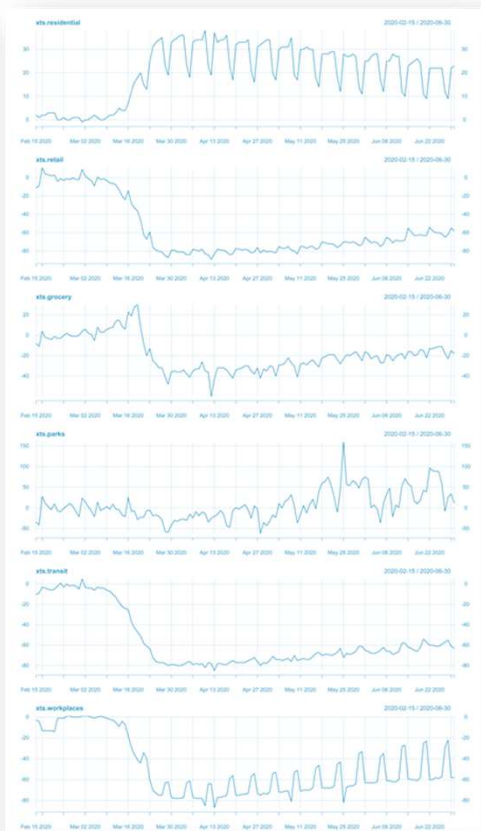


Figure 2



# Our analysis → Google social mobility data good indirect measure



Home



Retail



Grocery



Parks



Transit



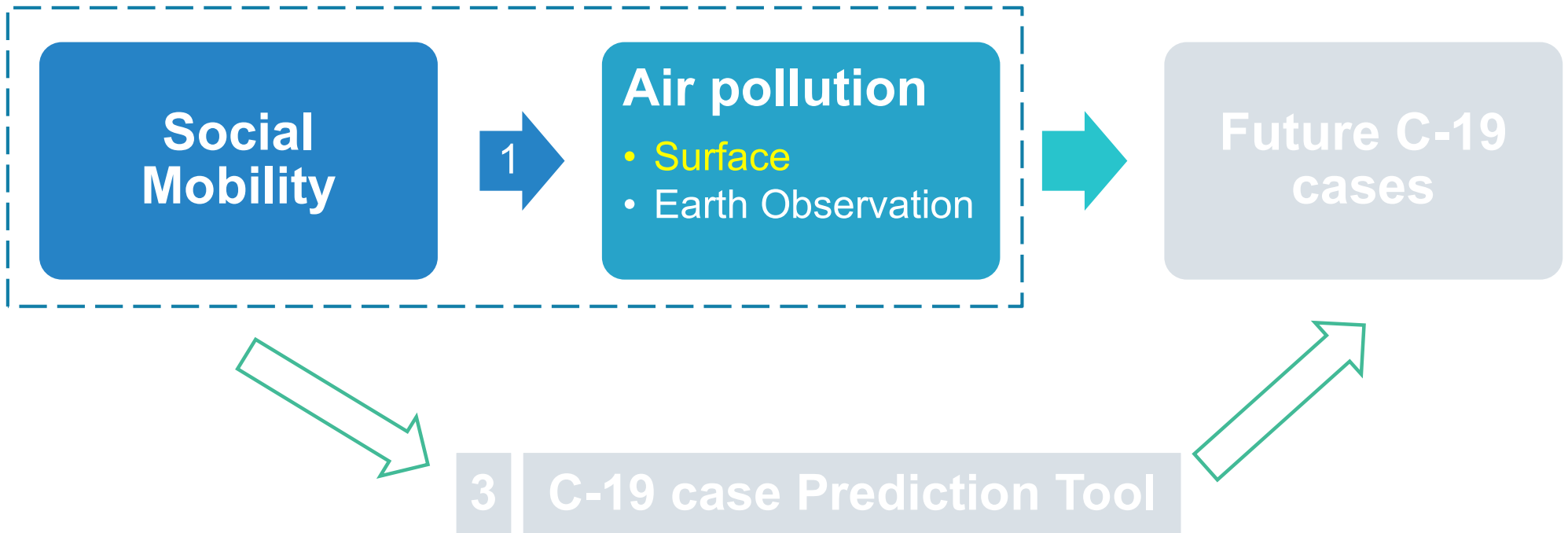
Workplace

- London – primary case study: Google Mobility data for London Feb 2020 to June 2020

- Same trend across European capitals



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# Social mobility vs Air pollution (surface)

THE LANCET  
Planetary Health

ARTICLES | VOLUME 4, ISSUE 10, E474-E482, OCTOBER 01, 2020

Short-term and long-term health impacts of air pollution reductions from COVID-19 lockdowns in China and Europe: a modelling study

Paolo Giani, MSc • Stefano Castruccio, PhD • Alessandro Anav, PhD • Prof Don Howard, PhD • Wenjing Hu, MSc • Paola Crippa, PhD

Open Access • Published: September 22, 2020 • DOI: [https://doi.org/10.1016/S2542-5196\(20\)30224-2](https://doi.org/10.1016/S2542-5196(20)30224-2)

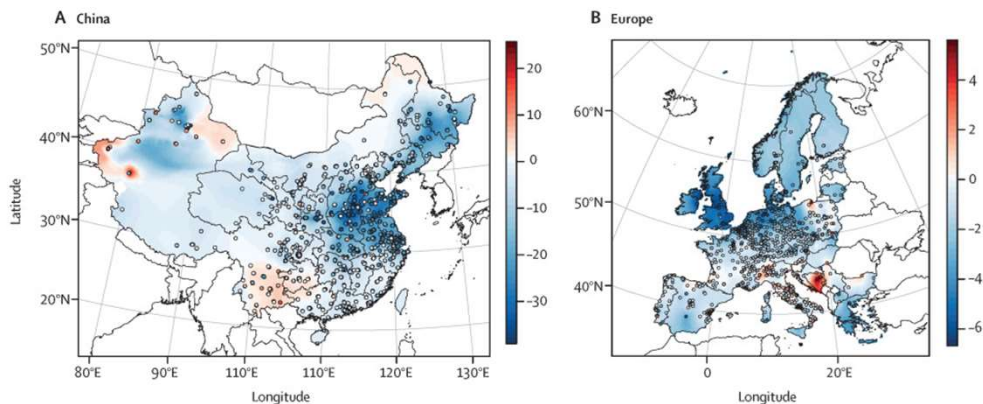


Figure 1 Effect of lockdown on surface PM<sub>2.5</sub> concentrations

- Observed by surface pollution stations
- Social mobility limitations (Lockdowns) = decrease in air pollution across regions

# Our analysis → Surface Pollution vs Google social mobility data

## NITROGEN DIOXIDE

Region	Residential time	Retail and recreation time	Grocery and pharmacy time	Parks time	Transit time	Workplaces time
Amsterdam	-0.03	0.19	0.26	0.16	0.16	-0.03
Ankara	-0.61	0.62	0.41	0.55	0.62	0.64
Athens	-0.06	0.15	0.19	0.18	0.11	0.1
Belfast	-0.37	0.59	0.58	0.33	0.51	0.38
Bratislava	0.19	0	0.08	0.04	-0.05	-0.21
Brussels	0.26	0.01	0.17	-0.01	-0.15	-0.23
Bucharest	-0.51	0.69	0.66	0.61	0.63	0.49
Budapest	-0.08	0.29	0.42	0.23	0.24	0.1
Copenhagen	-0.13	0.3	0.14	0.23	0.26	0.05
Dublin	-0.01	-0.05	0.03	0.16	-0.03	0.07
Edinburgh	-0.17	0.47	0.5	0.41	0.42	0.19
Helsinki	-0.12	0.51	0.45	-0.19	0.42	0.02
Lisbon	-0.31	0.5	0.5	0.49	0.46	0.33
London	0.05	0.15	0.21	0.12	0.06	-0.03
Madrid	-0.44	0.62	0.65	0.51	0.59	0.46
Oslo	-0.05	0.46	0.43	0.25	0.31	0.03
Paris	-0.07	0.27	0.32	0.23	0.23	0.07
Rome	-0.24	0.38	0.43	0.31	0.37	0.26
Sofia	-0.1	0.24	0.28	0.14	0.26	0.13
Tallinn	-0.12	0.32	0.27	0.26	0.27	0.04
Warsaw	0.25	0.03	0.02	-0.11	-0.09	-0.18
Zagreb	-0.21	0.54	0.61	0.09	0.46	0.34

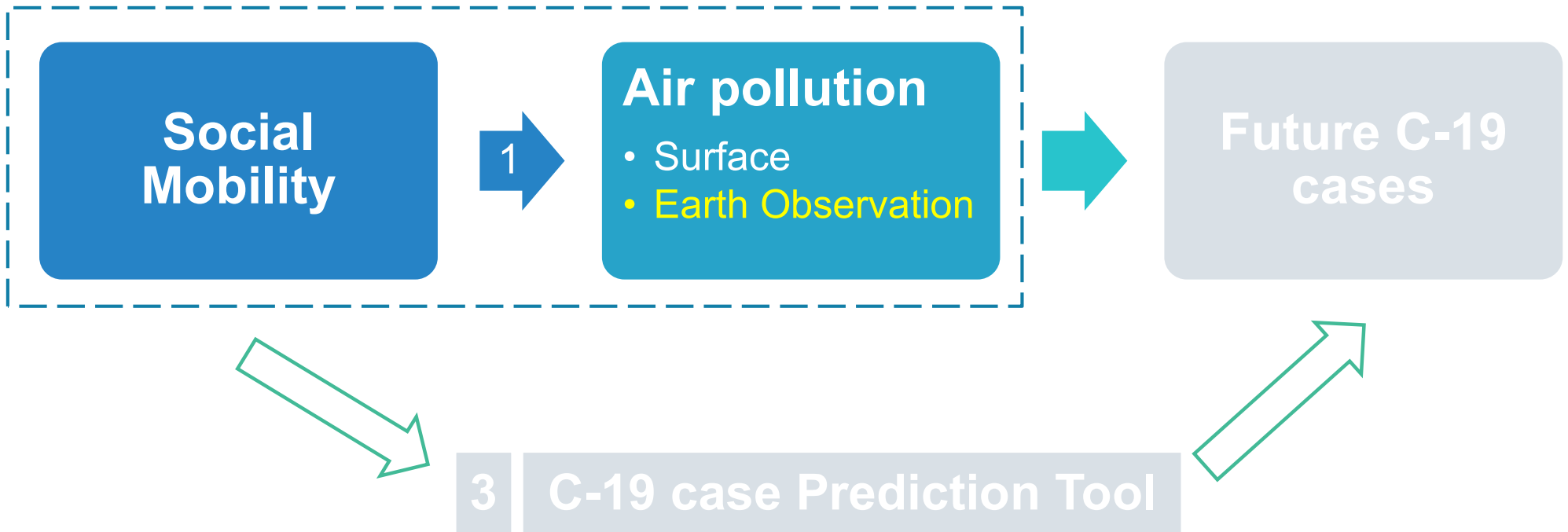
- Residential time causes significant drop in surface NO<sub>2</sub>
- Across 22 European capital cities
- Feb 2020 to June 2020 (start of the C-19 pandemic)
- Similar trend with CO

# Other surface pollutants and social mobility

Pollutant	Residential time	Retail and recreation time	Grocery and pharmacy time	Parks time	Transit time	Workplaces time
CO	Orange	Blue	Blue	Blue	Blue	Blue
O <sub>3</sub>	Blue	Orange	Orange	Blue	Orange	Orange
SO <sub>2</sub>	Blue	Orange	Orange	Blue	Orange	Orange
PM2.5	Blue	Orange	Orange	Blue	Orange	Orange

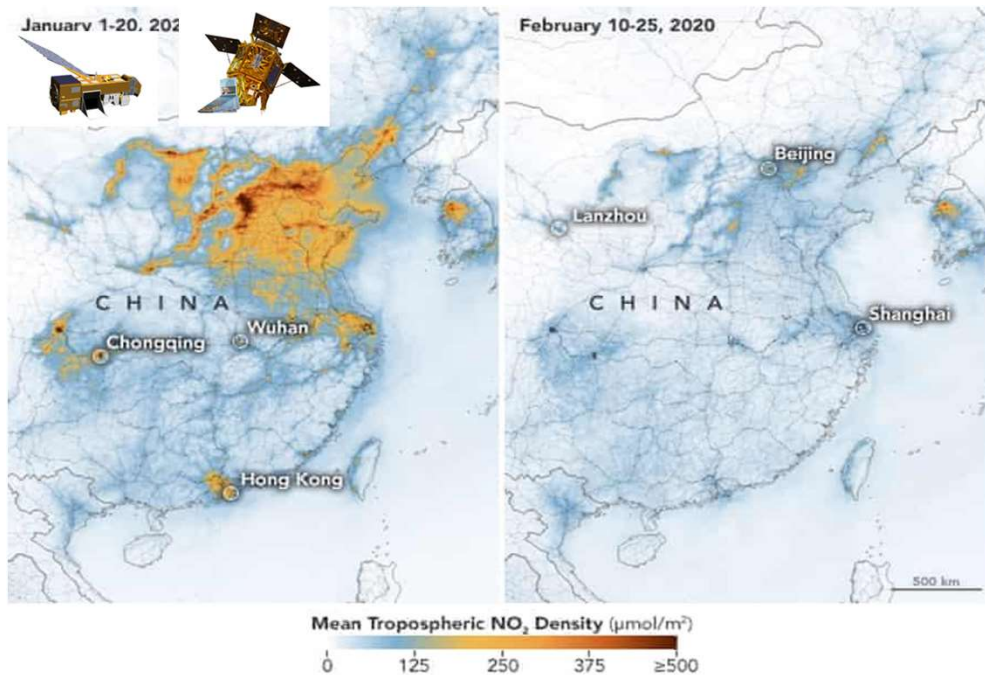
- Similar trends for surface CO = less CO more time at home
- Not all surface pollutant markers correlate with social mobility in the same way
- Opposite trends for O<sub>3</sub>, and mixed trends for SO<sub>2</sub> and PM<sub>2.5</sub>

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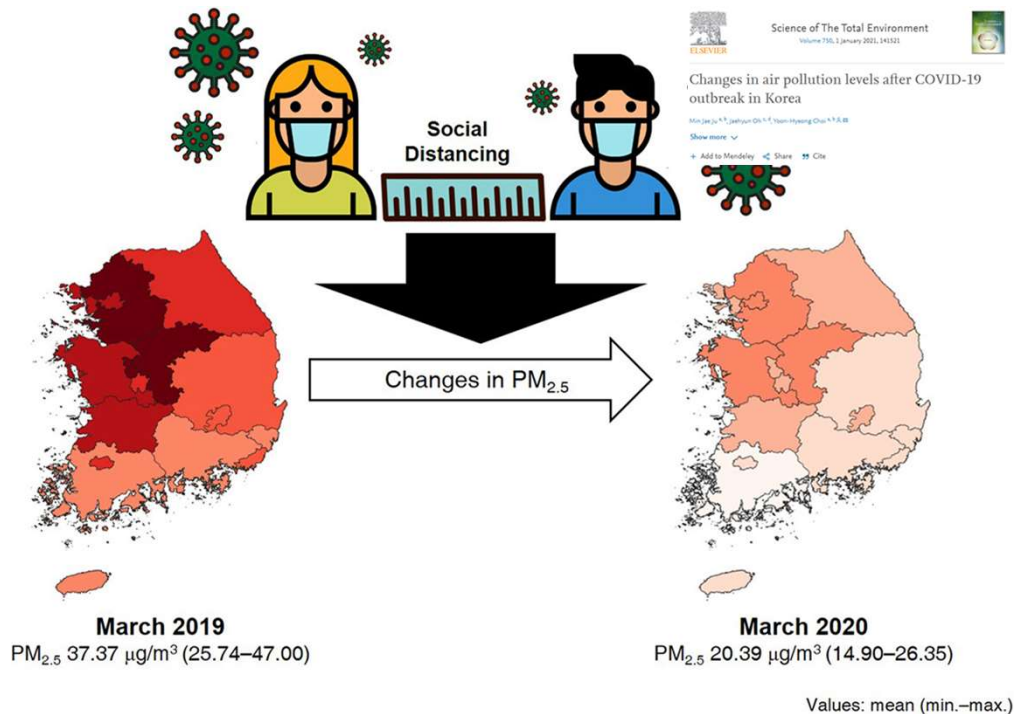


# Social mobility vs Air pollution (EO)



- Social mobility limitations (Lockdowns) = decrease in air pollution across regions
- Observed on satellite Earth Observation (ESA Sentinel-5 and NASA Aura)
- Focus – Wuhan, China

# Social mobility vs Air pollution (regional variation)



- Social mobility limitations (Lockdowns) = decrease in air pollution across regions
- Observed across different regions – map of Korea last year in March 2020

# Our analysis → EO Pollution and Google social mobility data in Europe

EO pollutants that were lower with more time at home? **NO<sub>2</sub>** and **SO<sub>2</sub>**

## NITROGEN DIOXIDE

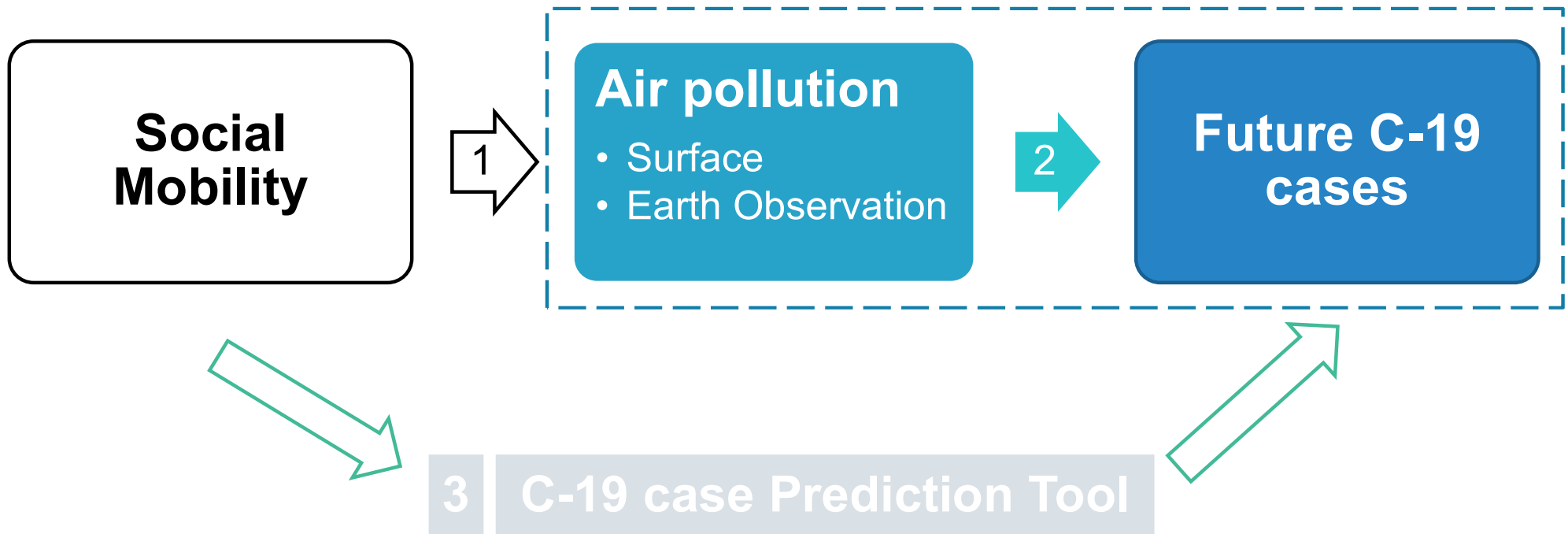
Region	Residential time	Retail and recreation time	Grocery and pharmacy time	Parks time	Transit time	Workplaces time
Amsterdam	0	0.11	0.11	-0.01	0.09	-0.03
Athens	0.11	-0.12	-0.03	0.18	-0.09	-0.1
Belgrade	0.06	-0.03	0.01	-0.05	0.01	-0.04
Berlin	-0.22	0.1	0.25	0.01	0.27	0.23
Bratislava	-0.22	0.31	0.14	0	0.29	0.19
Brussels	0.26	-0.18	0.01	-0.28	-0.24	-0.12
Bucharest	-0.03	0.08	0.09	0.02	0.05	0.06
Budapest	-0.2	0.28	0.27	0.12	0.2	0.17
Copenhagen	-0.36	0.26	0.19	0.09	0.34	0.35
Helsinki	0.03	0.04	0.05	0.28	0	-0.14
Lisbon	0.14	-0.04	-0.04	-0.02	-0.09	-0.14
Ljubljana	-0.03	0.18	0.26	0.08	0.05	0.03
Luxembourg	0.23	-0.03	0.18	-0.1	-0.12	-0.17
Madrid	-0.4	0.46	0.51	0.41	0.47	0.41
Dubai	-0.15	0.19	0.05	0.35	0.15	0.03
Paris	0.22	0.05	0.09	-0.02	-0.1	-0.18
Prague	0.21	-0.23	-0.13	-0.26	-0.24	-0.24
Rome	-0.11	0.22	0.26	0.19	0.19	0.14
Sofia	0.18	-0.15	-0.07	-0.11	-0.14	-0.18
Stockholm	-0.06	0.18	0.16	0.44	0.16	-0.13
Valletta	0.16	-0.1	-0.15	-0.02	-0.14	-0.16
Vienna	0.06	-0.01	-0.02	0.05	-0.03	-0.03
Warsaw	0.17	0.06	0.11	-0.01	-0.04	-0.15
Zagreb	-0.07	0.25	0.3	0.2	0.18	0.14
Zurich	0.38	-0.25	0.05	0.11	-0.33	-0.13

EO pollutants that increased with more time at home? **PM<sub>2.5</sub>**, **Total PM**, **UV**

## PM 2.5

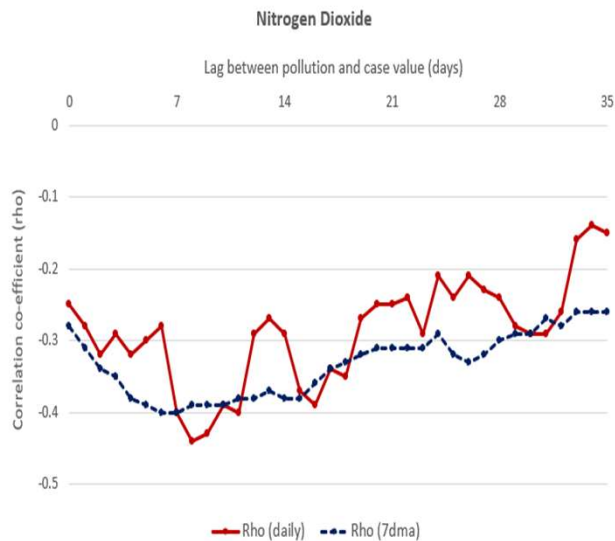
Region	Residential time	Retail and recreation time	Grocery and pharmacy time	Parks time	Transit time	Workplaces time
Amsterdam	0.53	-0.51	-0.49	0.15	-0.54	-0.49
Athens	0.32	-0.31	-0.24	-0.24	-0.33	-0.3
Belgrade	0.35	-0.34	-0.33	-0.28	-0.36	-0.31
Berlin	0.55	-0.55	-0.38	0.04	-0.59	-0.57
Bratislava	0.43	-0.35	-0.26	0.01	-0.4	-0.39
Brussels	0.52	-0.55	-0.54	-0.24	-0.57	-0.52
Bucharest	0.25	-0.24	-0.24	-0.21	-0.26	-0.24
Budapest	0.48	-0.41	-0.39	-0.41	-0.47	-0.47
Copenhagen	0.41	-0.42	-0.33	0.35	-0.51	-0.45
Helsinki	0.15	-0.36	-0.31	0.11	-0.29	-0.15
Lisbon	0.22	-0.21	-0.26	-0.27	-0.24	-0.26
Ljubljana	0.42	-0.42	-0.38	-0.31	-0.4	-0.37
Luxembourg	0.57	-0.58	-0.55	-0.42	-0.58	-0.54
Madrid	0.39	-0.48	-0.45	-0.52	-0.49	-0.41
Dubai	0.54	-0.51	-0.02	0.32	-0.55	-0.59
Paris	0.51	-0.57	-0.58	-0.55	-0.56	-0.5
Prague	0.37	-0.46	-0.39	-0.41	-0.44	-0.41
Rome	0.4	-0.42	-0.42	-0.47	-0.45	-0.41
Sofia	0.33	-0.29	-0.34	-0.3	-0.35	-0.36
Stockholm	0.41	-0.41	-0.36	0.16	-0.44	-0.42
Valletta	0.29	-0.25	-0.25	-0.31	-0.27	-0.22
Vienna	0.46	-0.41	-0.42	-0.17	-0.46	-0.47
Warsaw	0.24	-0.1	-0.11	0.05	-0.23	-0.21
Zagreb	0.51	-0.54	-0.56	-0.27	-0.58	-0.54
Zurich	0.45	-0.51	-0.21	0.11	-0.5	-0.45

# Can Earth Observation data be used to forecast the number of C-19 cases in an area?

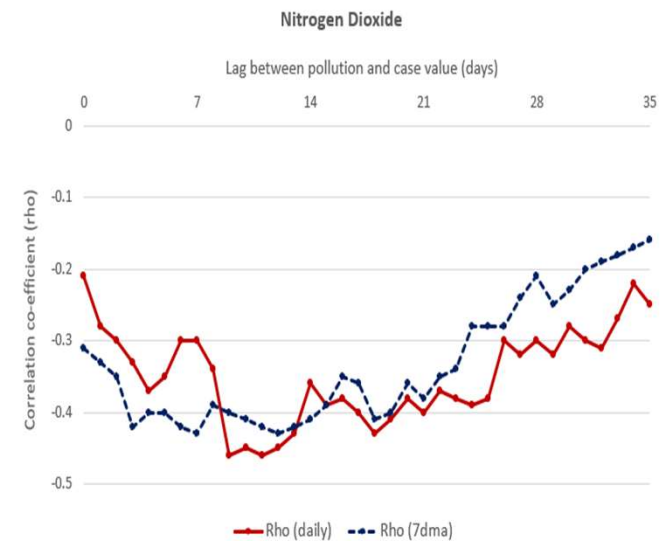


# Novid analysis: Surface air pollution vs C-19 cases with differential lag times

London, UK  
Nitrogen dioxide



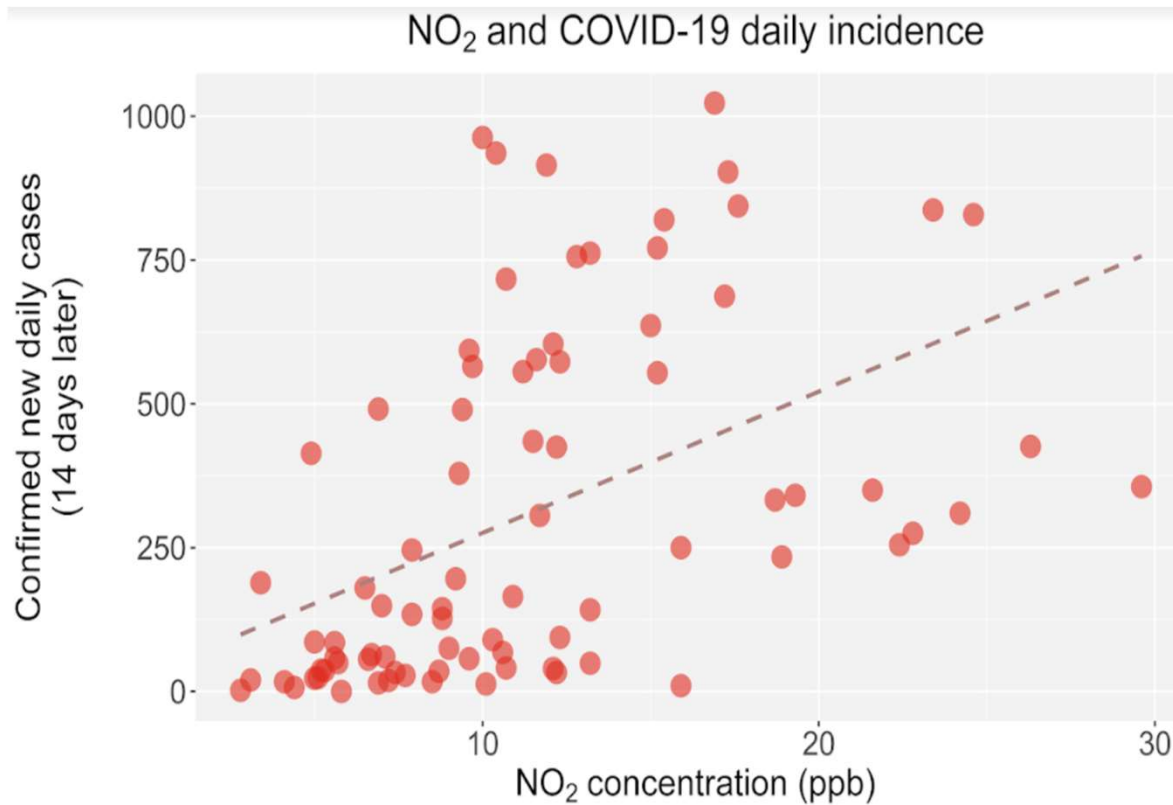
Budapest, HU  
Nitrogen dioxide



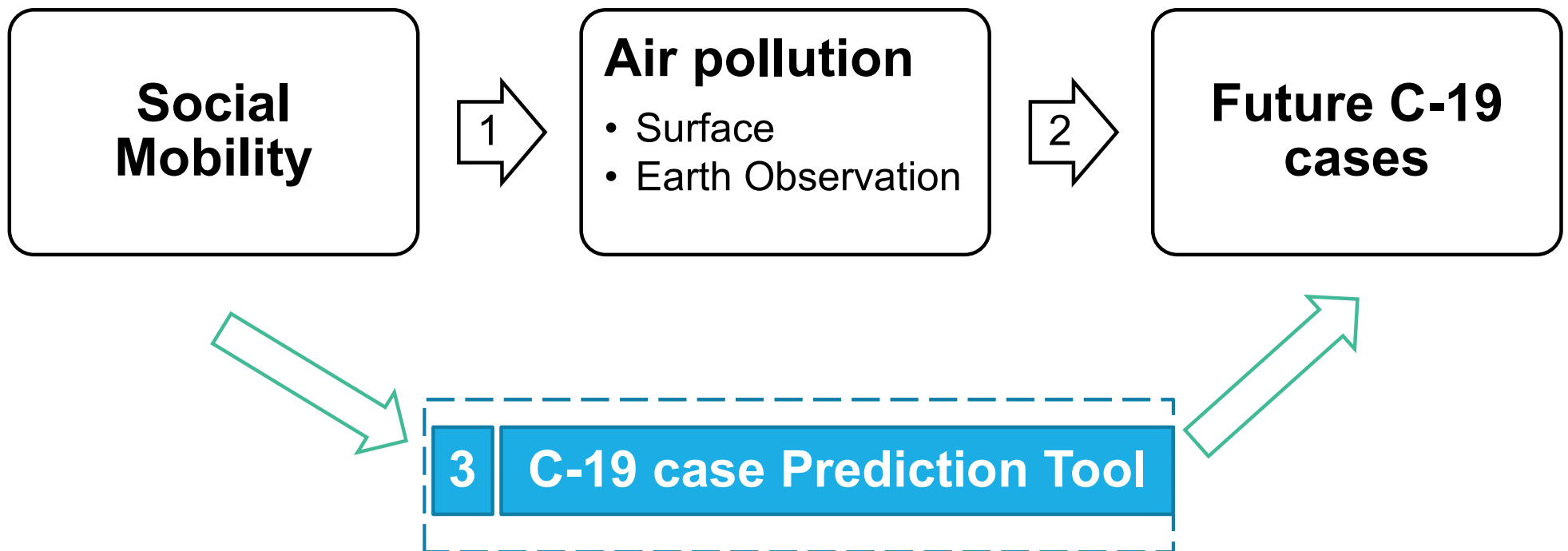
What is the best lag time (C-19 incubation time)? 14 days lag time (rho=0.59)



# Novid analysis: Air pollution vs C-19 cases with differential lag times

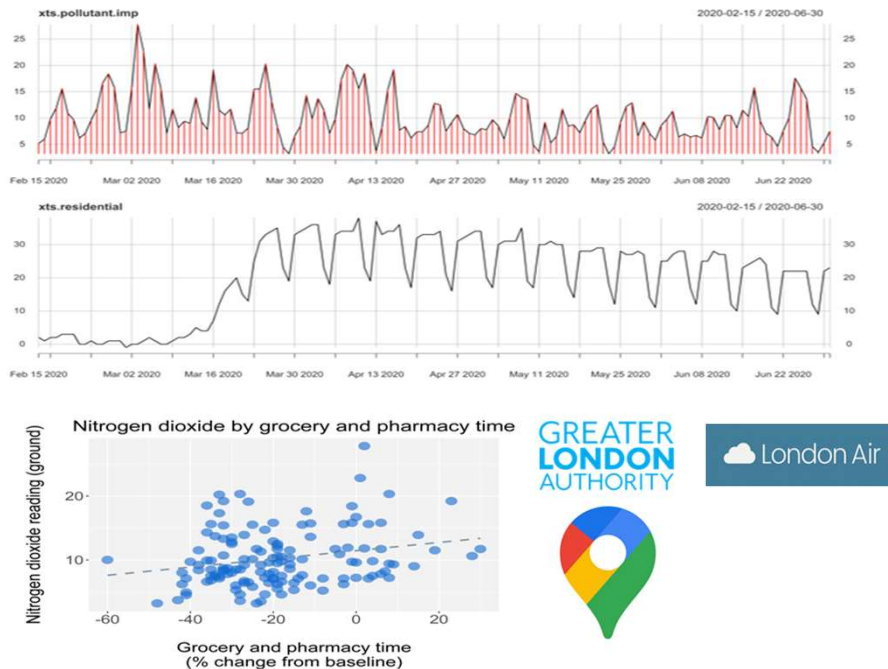


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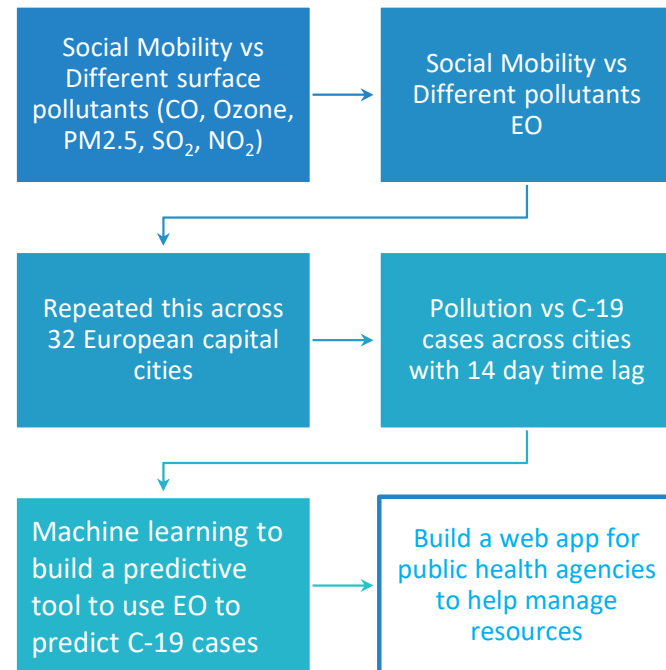


# Novid analysis: Social mobility vs Air pollution (surface)

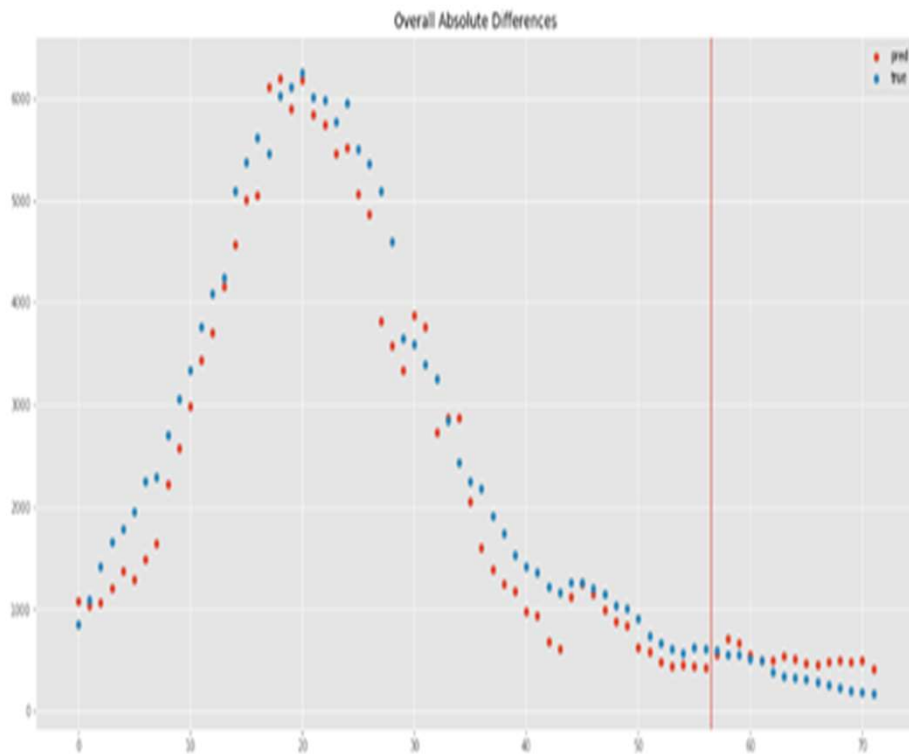
## Primary case study – Surface NO2 in London



## Additional analysis to build a 'forecasting tool'



# EO to predict C-19 – developing a ‘C-19 forecasting tool’



- Put all the information together in a ML Model
- 14 day lag time
- Trained the model with the data that we gathered
- **Model predicts Air pollution (NO2) T+0 and C-19 cases T+14days**
- **Limitations – vaccination drive, new variants incubation time etc.**

# Can Earth Observation data be used to forecast the number of C-19 cases in an area?

