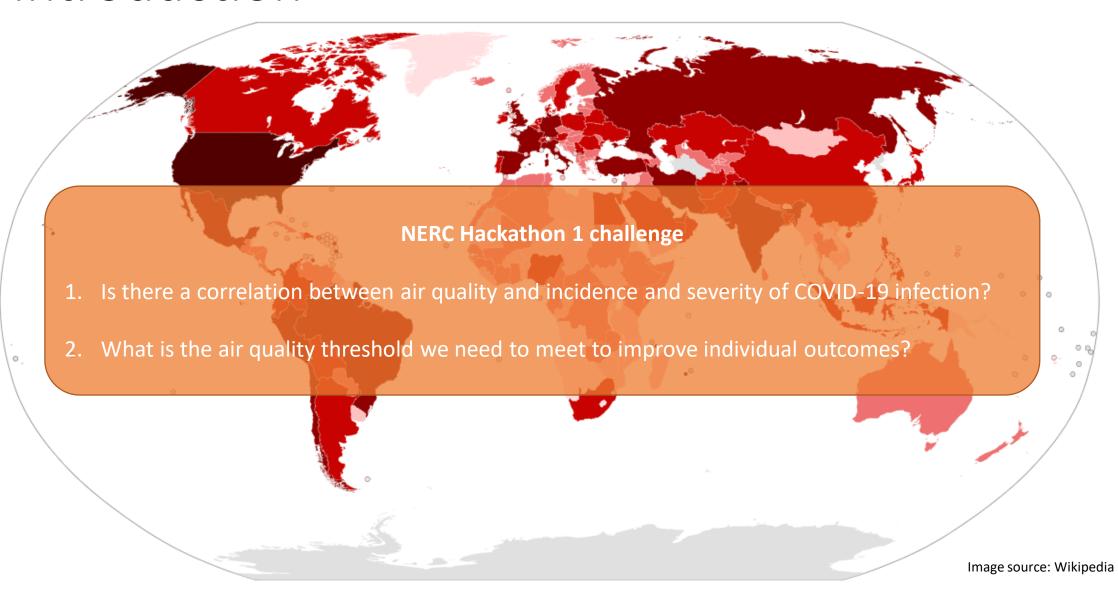


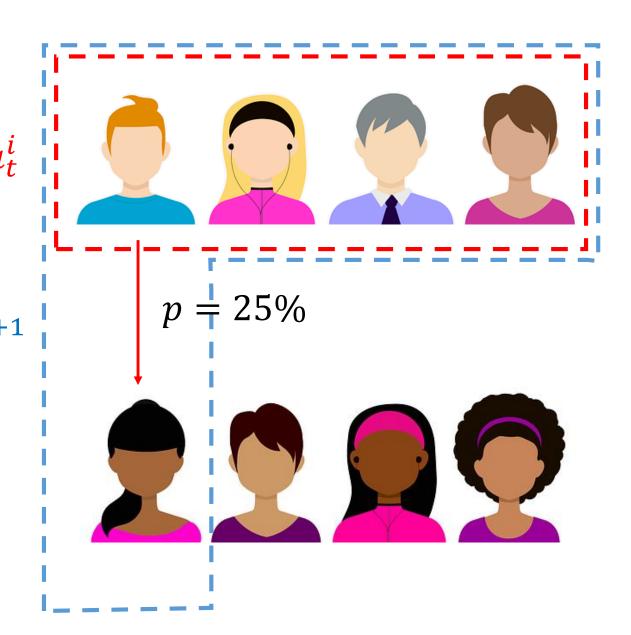
#### Introduction



#### Infection model

$$a_{t+1}^i - a_t^i = \text{Poisson}(a_t^i p_t^i)$$

$$\ln(p_t^i) = \theta_0 + \theta X_t^i + u^i + v_t + w_t^i$$
  $a_t^i$ 
Linear coefficients Parameters Random effects



# Bayesian framework in R



Pollutant \_data\_preprocessing

pollution\_data.csv



COVID data, county data

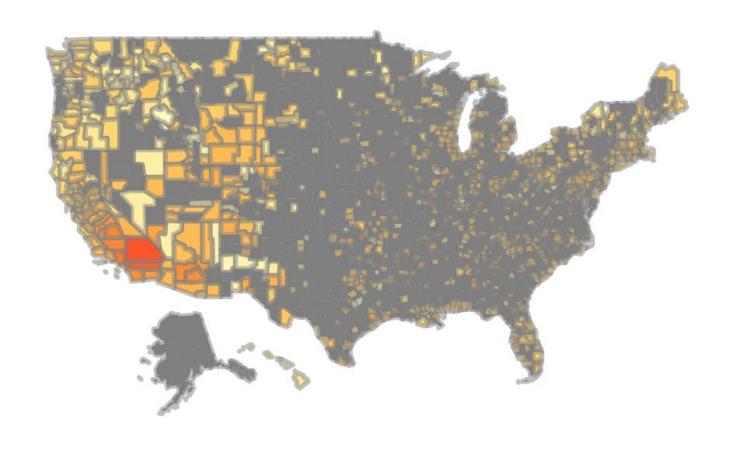
INLA\_data\_preprocessing

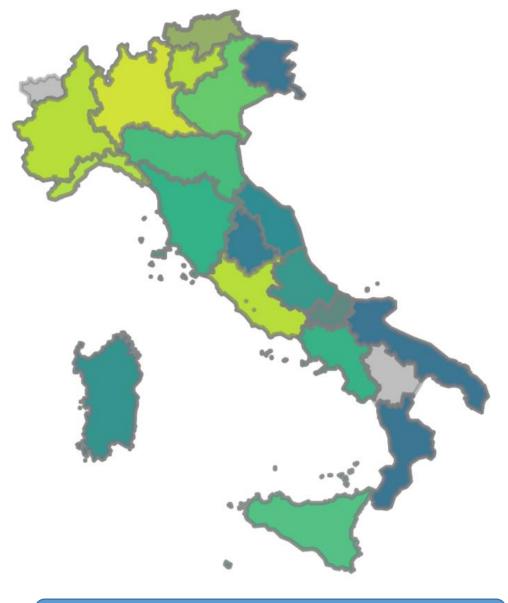
county\_data.csv weekly\_data.csv daily\_data.csv Estimates of effect of each factor on infection rate





INLA\_Bayesian\_framework





>50 parameters for USA

6 parameters for Italy

#### How to create model?

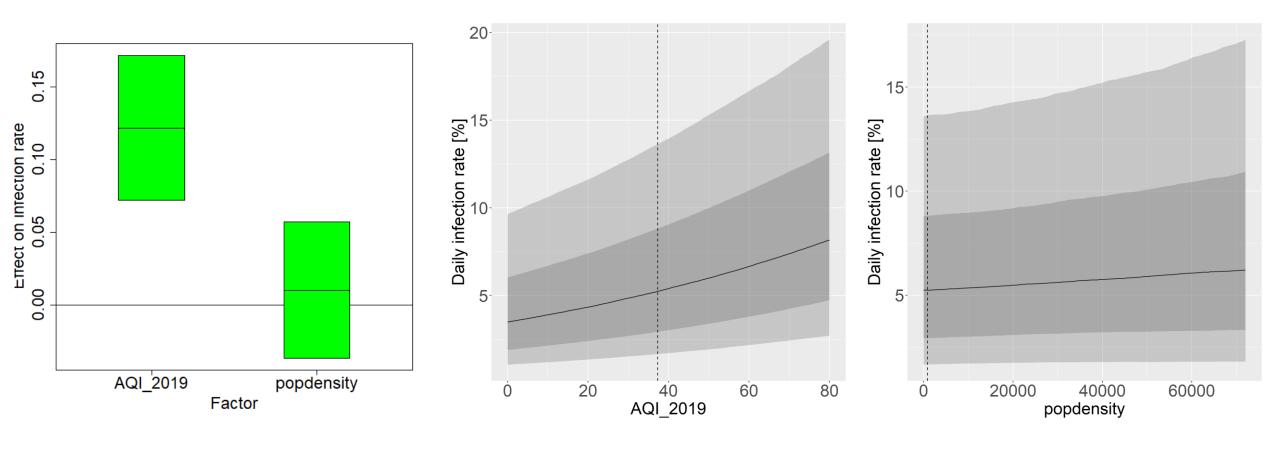
```
use_deaths <- TRUE \ln(p_t^i) = \theta_0 + \theta_1 AQI_{2019} + \theta_2 D + u^i + v_t + w_t^i use_weekly <- TRUE
```

```
covariates <- c("AQI_2weeks", "popdensity")

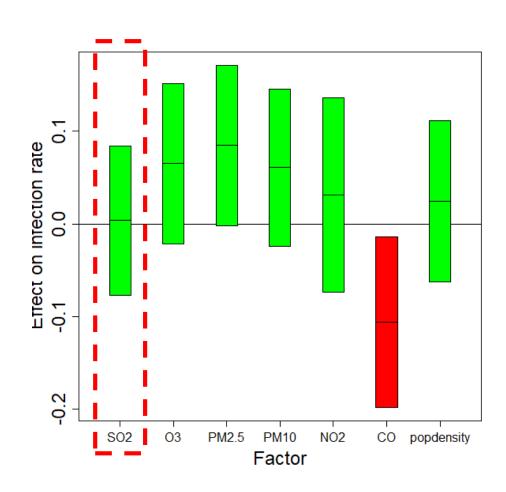
formula <- new_cases ~ AQI_2019 + popdensity +

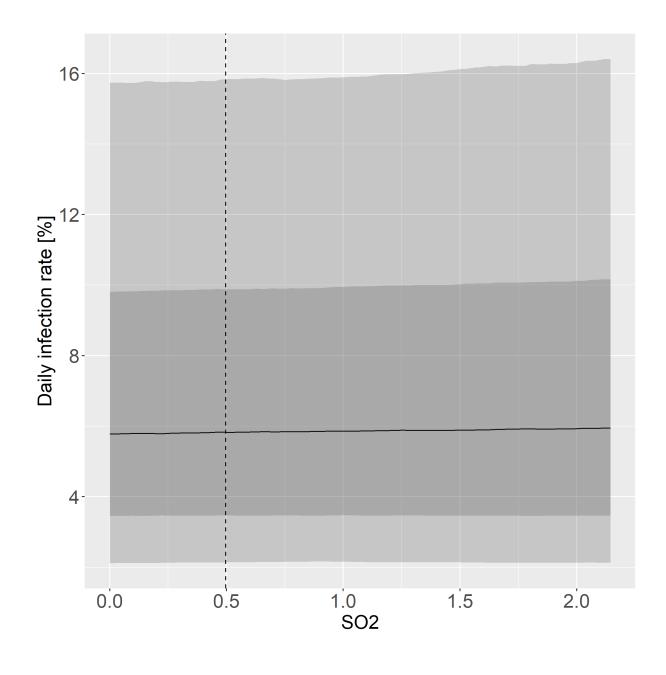
f(id, model = "iid") + f(day, model = "ar1") + f(rowId, model = "iid")
```

# Model output

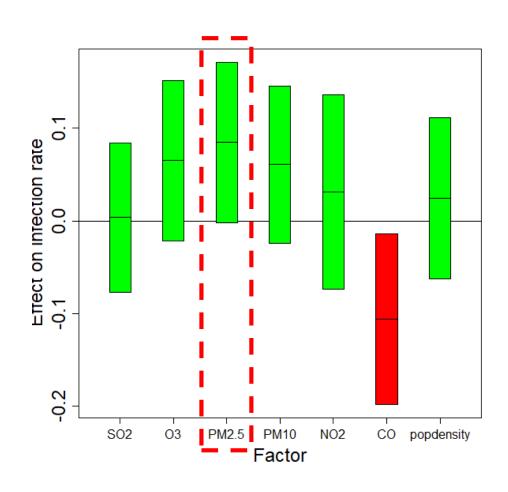


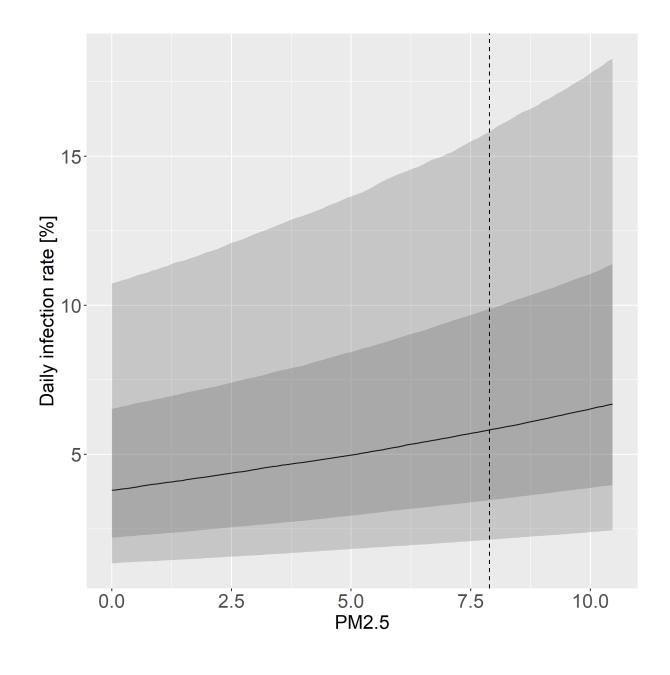
#### Another result



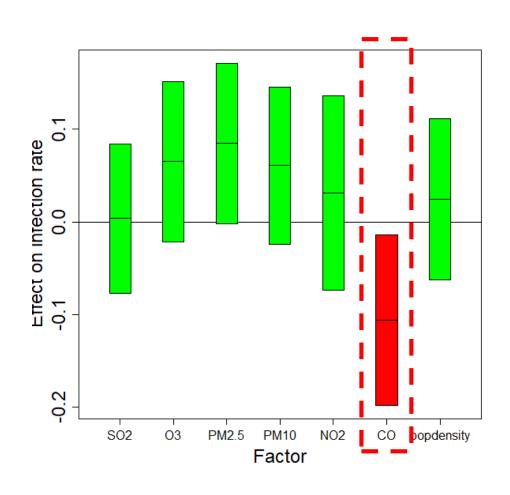


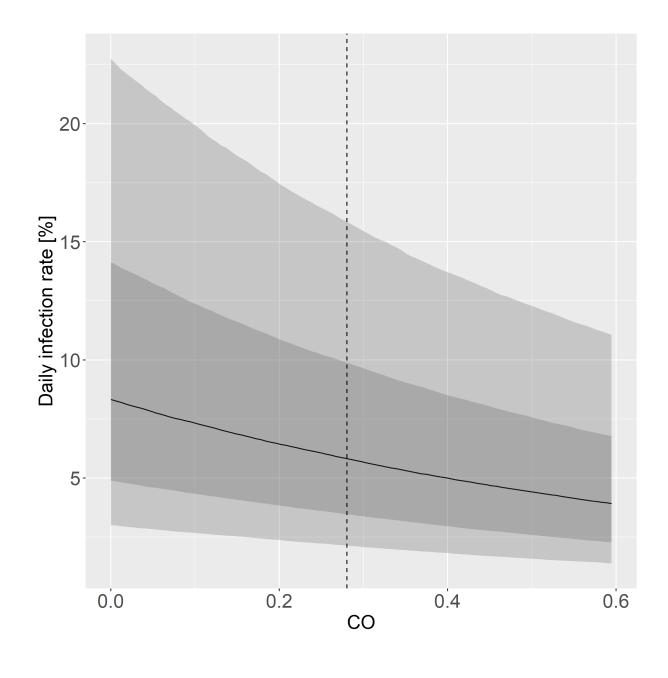
#### Another result





### Another result





# Summary

- Flexible, easy to use Bayesian framework for COVID-19 research
- Preliminary results allowed to investigate effect of air pollution in different time scales and effect of different pollutants.

# Summary

- Flexible, easy to use Bayesian framework for COVID-19 research
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#### What's next?

- Extending air pollution data set with a transportation model assimilated with both satellites and monitoring stations data.
- Studying the standard epidemic models (e.g. SIR) with infection rate dependent on pollutant concentration.