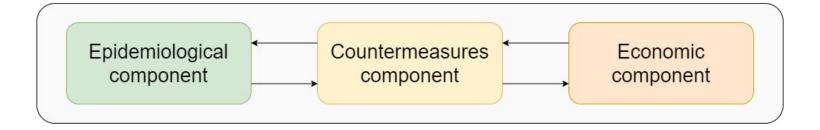
# Measuring COVID-19 economic and epidemiological countermeasures impact

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#### **Simulation - Architecture**



## **Epidemiological component - Environment**



Bipartite graph



CBGs

**POIs** 

## **Epidemiological component - Data**

Core POI

Weekly Patterns



Home Summary

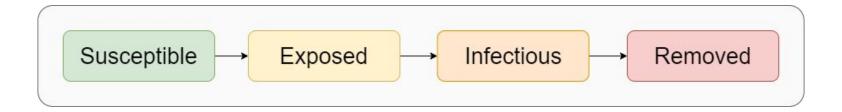
POI Area

American Community Survey - 2016

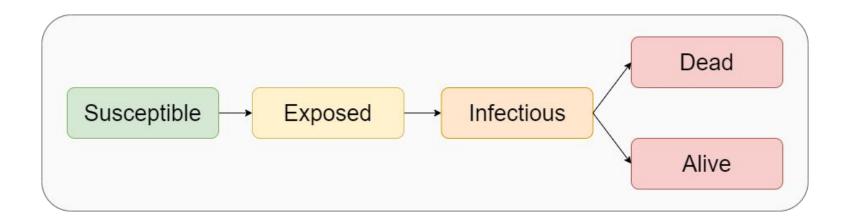
(US Census Bureau)

Social Distancing

#### **Epidemiological component - Model**



## **Epidemiological component - Model**



#### **Epidemiological component - Model**

$$N_{c_{i}} = S_{c_{i}}^{(t)} + E_{c_{i}}^{(t)} + I_{c_{i}}^{(t)} + R_{c_{i}}^{(t)}$$

$$\Delta S_{c_{i}}^{(t)} = -N_{S_{c_{i}} \to E_{c_{i}}}^{(t)}$$

$$\Delta E_{c_{i}}^{(t)} = N_{S_{c_{i}} \to E_{c_{i}}}^{(t)} - N_{E_{c_{i}} \to I_{c_{i}}}^{(t)}$$

$$\Delta I_{c_{i}}^{(t)} = N_{E_{c_{i}} \to I_{c_{i}}}^{(t)} - N_{I_{c_{i}} \to R_{c_{i}}}^{(t)}$$

$$\Delta R_{c_{i}}^{(t)} = N_{I_{c_{i}} \to R_{c_{i}}}^{(t)}$$

#### **Epidemiological component - Initialization**

$$S_{c_i}^{(0)} = N_{c_i} - E_{c_i}^{(0)}$$

$$E_{c_i}^{(0)} = Binom(N_{c_i}, p_0)$$

$$I_{c_i}^{(0)} = 0$$

$$R_{c_i}^{(0)} = 0$$

#### **Epidemiological component - Simulation**

$$\begin{split} N_{S_{c_{i}} \to E_{c_{i}}}^{(t)} &= Binom(w_{ij}^{(t)} \cdot \frac{S_{c_{i}}^{(t)}}{N_{c_{i}}}, \lambda_{p_{j}}^{(t)}) + Binom(S_{c_{i}}^{(t)}, \lambda_{c_{i}}^{(t)}) \\ N_{E_{c_{i}} \to I_{c_{i}}}^{(t)} &= Binom(E_{c_{i}}^{(t)}, \frac{1}{\delta_{E}}) \\ N_{I_{c_{i}} \to R_{c_{i}}}^{(t)} &= Binom(I_{c_{i}}^{(t)}, \frac{1}{\delta_{I}}) \\ \lambda_{c_{i}}^{(t)} &= \beta_{base} \frac{I_{c_{i}}^{(t)}}{N_{c_{i}}} \end{split}$$

#### **Epidemiological component - Grid Search**

 $\beta_{\text{base}}$ 

Transmission constant for CBGs

10 values in the range 0.0012 - 0.024

Ψ

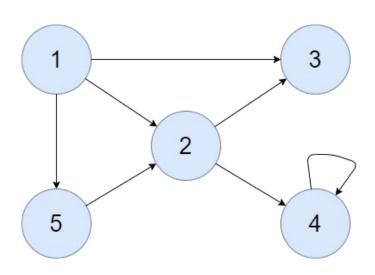
Transmission constant for POIs

9 values in the range 0.001 - 50  $\mathbf{O}_0$ 

Probability of being exposed at time t<sub>o</sub>

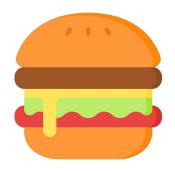
These 10 values 1.10<sup>-2</sup> 5.10<sup>-3</sup> 2.10<sup>-3</sup> 1.10<sup>-3</sup> 5.10<sup>-4</sup>, 2.10<sup>-4</sup> 1.10<sup>-4</sup> 5.10<sup>-5</sup> 2.10<sup>-5</sup> 1.10<sup>-5</sup>

## **Economic component - Input Output Theory**



Economic Activities	1	2	3	 Z	Final Demand	Exports
1						
2						
3						
Z						
Imports						

### **Countermeasure component**



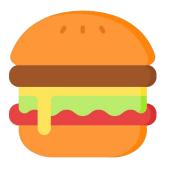
Closure of food activities



Closure of cinemas and theaters



Closure of religious organizations



Closure of food activities after 18

#### **Implementation**



O PyTorch

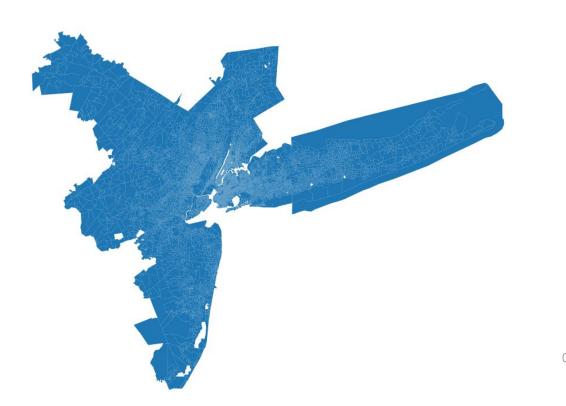
AMD 3900X - 64 GB -Nvidia RTX 3090 (using the sparse matrix capabilities)

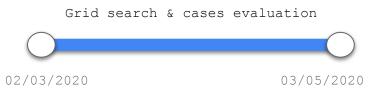
Grid Search ~7 days

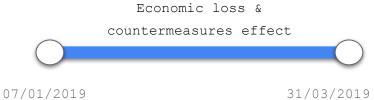
Single simulation on SSD ~7 minutes

Single simulation on HDD ~1 hour

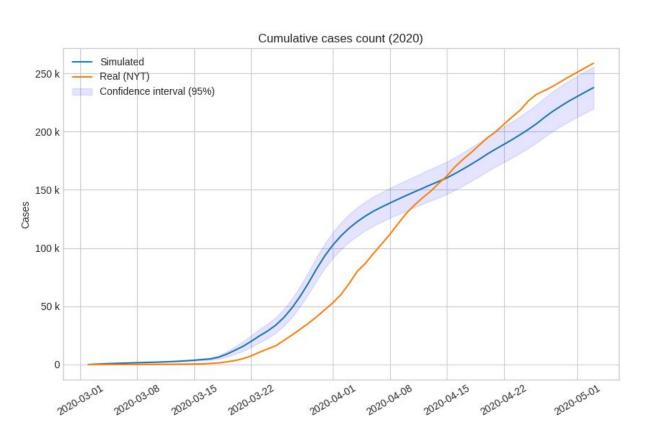
#### **Area and Period**



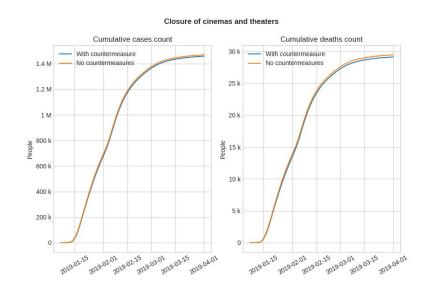


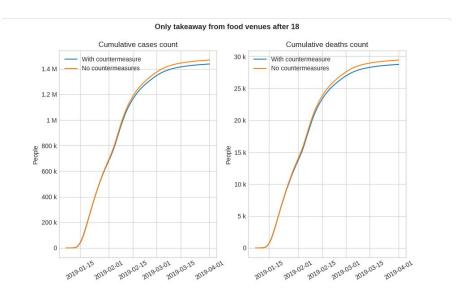


#### Results - Cumulative real / simulated cases

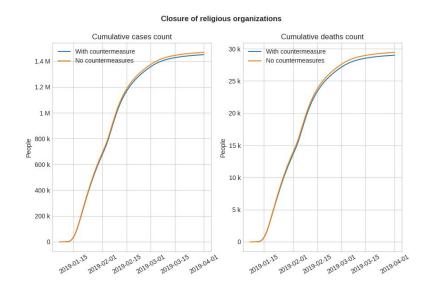


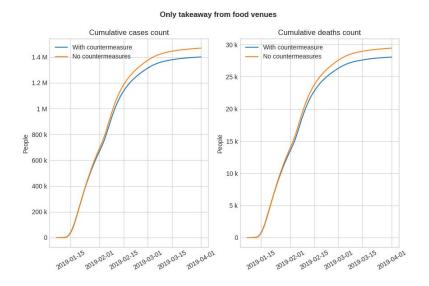
## **Results - Countermeasure effects (1)**



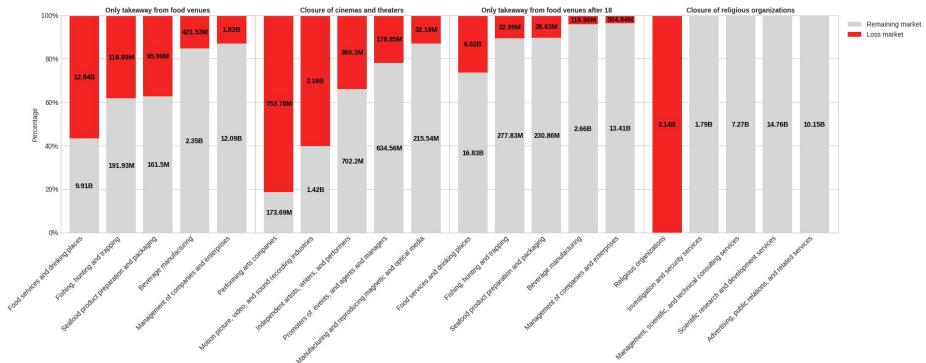


# Results - Countermeasure effects (2)

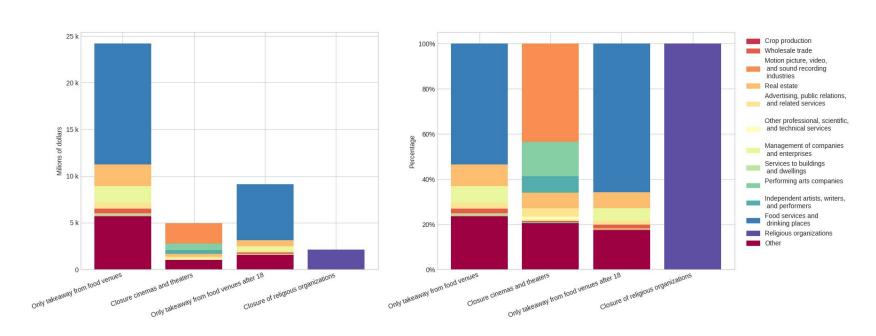




# Results - Countermeasure economic impact (sector viewpoint)



# Results - Countermeasure economic effects (economy viewpoint)



#### **Results - Conclusion**

Counter-measure	Economic losses (in B\$)	People saved	Cases avoided	Ratio (in M\$ x people saved)
Closure of cinemas and theaters	4.95	336	10365	14.7
Closure of religious organizations	2.14	443	17393	4.8
Only takeaway from food venues	24.19	1389	66837	17.4
Only takeaway from food venues after 18	9.14	682	30629	13.4

#### **Conclusion**

- New simulation paradigm, combines Al-like workloads (characterized by the use of large quantities of data) with the classic simulation world
- Fills the gap between agent-oriented simulation and high-level statistical techniques
- Can be improved by correlating the disease spreading mechanisms with people social behaviour inside different POI categories.
- From the economic viewpoint, the main problem is the unavailability of data with finer granularity

#### **Discussion**

