

Optimal Lockdown in a Commuting Network

by Pablo Fajgelbaum, Amit Khandelwal, Wookun Kim, Cristiano Mantovani and Edouard Schaal

Discussion by: Fabrizio Perri
Minneapolis Fed



Federal Reserve Bank of Atlanta, December 2020

Lockdowns in a network

- Networks of human interactions (like commuting) generate, at the same time, economic value (+) and infections (-) (Fogli and Veldkamp, 2019)

Lockdowns in a network

- Networks of human interactions (like commuting) generate, at the same time, economic value (+) and infections (-) (Fogli and Veldkamp, 2019)
- In a pandemic (COVID) efficient to reduce interactions

Lockdowns in a network

- Networks of human interactions (like commuting) generate, at the same time, economic value (+) and infections (-) (Fogli and Veldkamp, 2019)
- In a pandemic (COVID) efficient to reduce interactions
- Different edges of network have different economic/infection potential
- Uniform reduction however typically not efficient

Lockdowns in a network

- Networks of human interactions (like commuting) generate, at the same time, economic value (+) and infections (-) (Fogli and Veldkamp, 2019)
- In a pandemic (COVID) efficient to reduce interactions
- Different edges of network have different economic/infection potential
- Uniform reduction however typically not efficient
- Want to restrict edges with highest infection transmission, lowest economic impact

Lockdowns in a network

- Networks of human interactions (like commuting) generate, at the same time, economic value (+) and infections (-) (Fogli and Veldkamp, 2019)
- In a pandemic (COVID) efficient to reduce interactions
- Different edges of network have different economic/infection potential
- Uniform reduction however typically not efficient
- Want to restrict edges with highest infection transmission, lowest economic impact
- Key challenge: shutting down an edge has repercussions across the network and across time

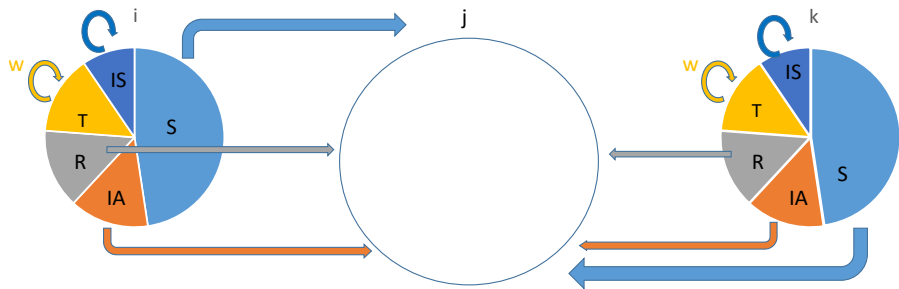
This paper

- Write down a commuting/shopping network of a city, add temporary pandemic (COVID)
- Solve for Pareto frontier of flows across city locations, and over time
- Calibrate to three cities and compare observed flows during COVID to frontier
- Main finding: observed flows are suboptimal, we can do better, **save lives and \$!**

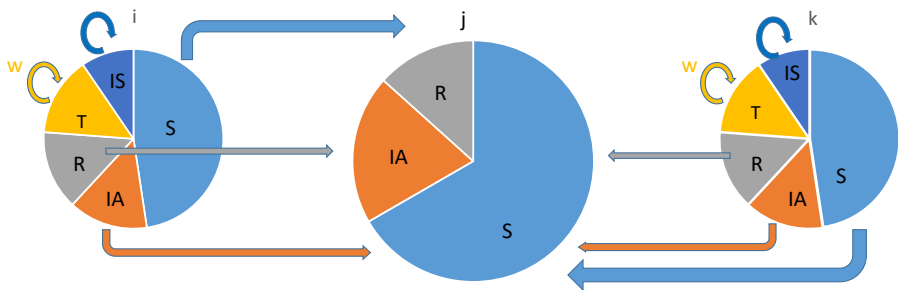
My discussion

1. Quick graphical summary
2. Praise and comments
3. Can we do even better?

The production/infection block



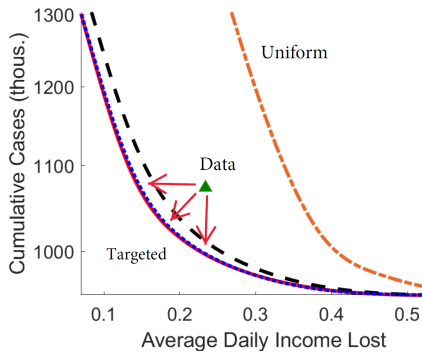
The production/infection block



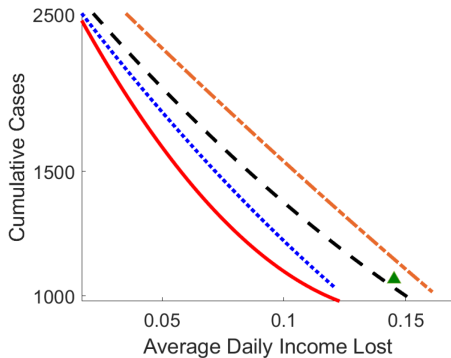
- Output: $[(S_i + IA_i + R_i)\chi_{ijt} + (S_k + IA_k + R_k)\chi_{kjt}]w_j + (T_i + T_k)w_j$
- Infections: $(S_i + S_k)(IA_i + IA_k)\chi_{ijt}\chi_{kjt}dens_j$
- Flows can't depend on status (no testing)
- In targeted lockdown χ_{ijt} can depend on characteristic of the locations ($w_j, dens_j, IS_k, ..$)

Key results

(c) NYC Metro



(b) Seoul



Praise

- Short and sweet!

Praise

- Short and sweet!
- Hard problem to solve as χ_{ijt} impacts output today, but has dynamic impact on future infections

Praise

- Short and sweet!
- Hard problem to solve as χ_{ijt} impacts output today, but has dynamic impact on future infections
- Love the planner approach and comparison with data. Is the current allocation is efficient? if not, what is the direction of the inefficiency, need more or less lockdowns?
 - ▶ Crucial in current policy debate across US states
 - ▶ No need to directly tackle behavioral response!

Implementation of optimal lockdowns, 1

- Model: planner can directly control flows of people from location i to j by '*lockdown measures, providing incentives, broadcasting information*'
- Practice: assessing the impact of these policies on flows require understanding of behavioral responses

Implementation of optimal lockdowns, 1

- Model: planner can directly control flows of people from location i to j by '*lockdown measures, providing incentives, broadcasting information*'
- Practice: assessing the impact of these policies on flows require understanding of behavioral responses
- Example: if authority cuts subway rides but people packs existing rides. Overcrowded PT, no reduction in flows and higher infection!

Implementation of optimal lockdowns, 2

- Model: planner can directly control flows of people and goods from location i to j
- Practice: people and goods can move from i to j in many ways (through $k, m, ..$), so controlling flows from i to j might be hard

Implementation of optimal lockdowns, 2

- Model: planner can directly control flows of people and goods from location i to j
- Practice: people and goods can move from i to j in many ways (through k, m, \dots), so controlling flows from i to j might be hard
- Example: if authority restricts moving from Brooklyn to Manhattan, people can go through Queens!

Estimation detail

- Model: fraction of tele-commuters δ is constant
reduction in output \simeq reduction in commuting flows
- Practice: large increase in tele-commuters (for economists 0 to 100%),
reduction in output \ll reduction in commuting flows
- Easy fix: estimate time variation in δ directly from data

The importance of the shopping block? GE?

- Model: χ_{kjt} also affects (in a iceberg way) the cost of shopping from k to j . Local output = local spending
- Practice: most shopping is either done locally (within k) or online (SOE). Large city level CA imbalances
- Suggestion: just solve the model relaxing GE (does not seem appropriate for cities) and without shopping block

The importance of the shopping block? GE?

- Model: χ_{kjt} also affects (in a iceberg way) the cost of shopping from k to j . Local output = local spending
- Practice: most shopping is either done locally (within k) or online (SOE). Large city level CA imbalances
- Suggestion: just solve the model relaxing GE (does not seem appropriate for cities) and without shopping block
- My prior: similar results but sharper message!

Is space the key dimension? Can we do even better?

- Paper stresses heterogeneity (in infection/output) across **space** for targeting lockdowns

Is space the key dimension? Can we do even better?

- Paper stresses heterogeneity (in infection/output) across **space** for targeting lockdowns
- Recent policy experience (second wave lockdowns) stresses heterogeneity across **business types** (i.e. bars, restaurants, gyms)
- My own research (Azzimonti et al. 2020) finds highest infection potential in businesses with **random long links** (concerts, rallies)

Is space the key dimension? Can we do even better?

- Paper stresses heterogeneity (in infection/output) across **space** for targeting lockdowns
- Recent policy experience (second wave lockdowns) stresses heterogeneity across **business types** (i.e. bars, restaurants, gyms)
- My own research (Azzimonti et al. 2020) finds highest infection potential in businesses with **random long links** (concerts, rallies)
- Maybe instead of shut down Manhattan and leave Queens open, shut down all bars in Manhattan and Queens? or close theme parks in Fla? ski resorts in the alps?

Is space the key dimension? Can we do even better?

- Paper stresses heterogeneity (in infection/output) across **space** for targeting lockdowns
- Recent policy experience (second wave lockdowns) stresses heterogeneity across **business types** (i.e. bars, restaurants, gyms)
- My own research (Azzimonti et al. 2020) finds highest infection potential in businesses with **random long links** (concerts, rallies)
- Maybe instead of shut down Manhattan and leave Queens open, shut down all bars in Manhattan and Queens? or close theme parks in Fla? ski resorts in the alps?
- Targeting business types might be more efficient and easier to implement