

# Markets for Efficient Public Good Allocation with Social Distancing

Devansh Jalota, Qi Qi, Marco Pavone, Yinyu Ye



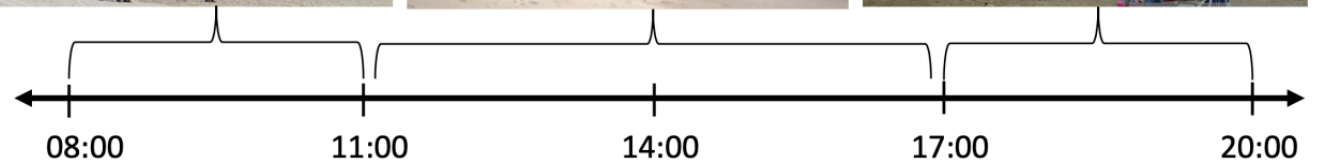
Stanford University



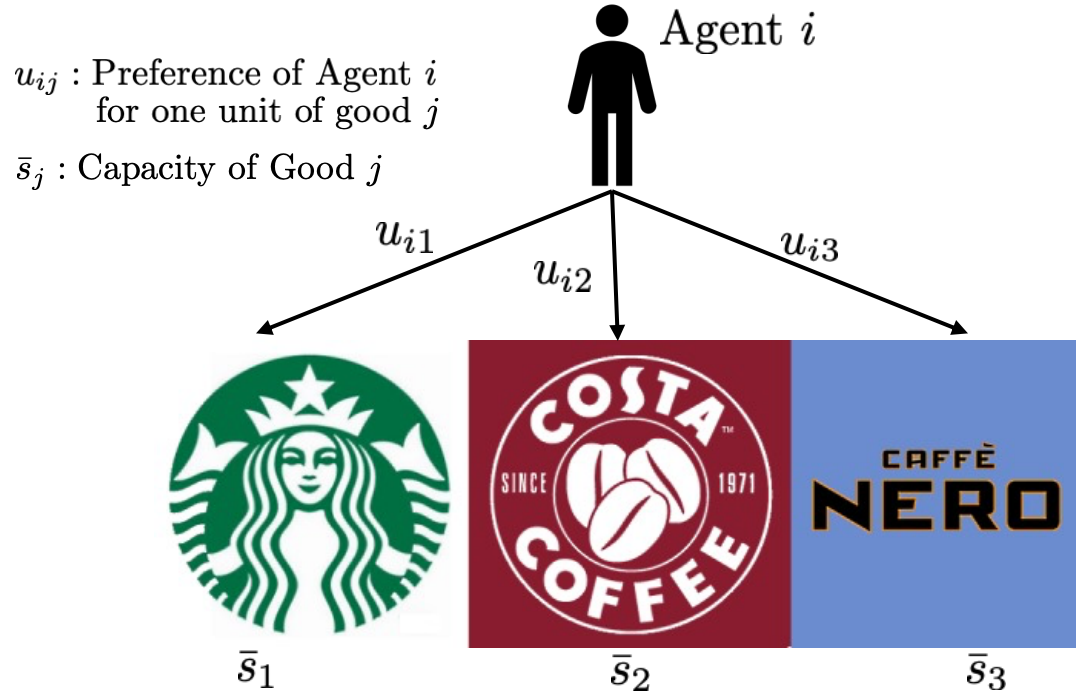
**Motivation:** Public goods are either over-consumed, or remain unused, as in the Covid-19 pandemic



**Our Approach:** Design market mechanism using *Fisher Markets* and price “times of use” to achieve an intermediate outcome and in doing so overcome a fundamental limitation of *Fisher Markets* that only consider budget and capacity constraints



# Classical Fisher Markets are not expressive enough to capture additional physical constraints



**Social Optimization Problem (SOP):**

$$\max_{\mathbf{x}_i, \forall i \in [N]} \sum_i w_i \log \left( \sum_j u_{ij} x_{ij} \right)$$

$$\text{s.t.} \quad \sum_i x_{ij} = \bar{s}_j, \forall j \in [M]$$

$$x_{ij} \geq 0, \forall i, j \quad \text{Capacity Constraint}$$

$p_j$  : Price of Good  $j$  = Dual Variable of Constraint  $j$

**Individual Optimization Problem:**

$$\max_{\mathbf{x}_i} \sum_j u_{ij} x_{ij}$$

$$\text{s.t.} \quad \mathbf{p}^T \mathbf{x}_i \leq w_i$$

$$\mathbf{x}_i \geq \mathbf{0}$$

Budget Constraint



On a given day, I would like to go to one park, one library and one coffee shop



With physical constraints, budget perturbations reweight utilities and provide more constrained people a fair chance to use budgets

### IOP with Additional Constraints

$$\begin{aligned}
 & \text{IOP} \\
 & \max_{\mathbf{x}_i} \sum_j u_{ij} x_{ij} \\
 & \text{s.t. } \mathbf{p}^T \mathbf{x}_i \leq w_i \\
 & \quad A_t^{(i)} \mathbf{x}_i \leq 1, \forall t \in T_i \\
 & \quad \mathbf{x}_i \geq \mathbf{0}
 \end{aligned}$$

Physical Constraints

**Theorem 0:** Under mild technical assumptions, the market clearing equilibrium price vector exists.

### SOP with Additional Constraints

$$\begin{aligned}
 & \text{SOP1} \\
 & \max_{\mathbf{x}_i, \forall i \in [N]} \sum_i w_i \log \left( \sum_j u_{ij} x_{ij} \right) \\
 & \text{s.t. } \sum_i x_{ij} \leq \bar{s}_j, \forall j \in [M] \\
 & \quad A_t^{(i)} \mathbf{x}_i \leq 1, \forall t \in T_i, \forall i \in [N] \\
 & \quad x_{ij} \geq 0, \forall i, j
 \end{aligned}$$

**Theorem 1:** The price vector corresponding to the optimal dual variables of the capacity constraint of **SOP1** is not an equilibrium price, i.e., the market clearing first order KKT conditions of **IOP** and **SOP1** are not equivalent.

### Budget Perturbed SOP



$$\begin{aligned}
 & \text{BP-SOP} \quad \text{Budget Perturbation} \\
 & \max_{\mathbf{x}_i} \sum_i (w_i + \lambda_i) \log \left( \sum_j u_{ij} x_{ij} \right) \\
 & \text{s.t. } \sum_i x_{ij} \leq \bar{s}_j, \forall j \in [M] \\
 & \quad A_t^{(i)} \mathbf{x}_i \leq 1, \forall t \in T_i, \forall i \in [N] \\
 & \quad x_{ij} \geq 0, \forall i, j
 \end{aligned}$$

$r_{it}$ : Dual Variable of Physical Constraint

**Theorem 2:** There is a one-to-one correspondence between the equilibrium price vector and a *fixed-point* solution of **BP-SOP**.

Fixed Point of **BP-SOP**:  $\lambda_i = \sum_t r_{it}$

# We use a fixed-point scheme to determine perturbation constants and establish convergence through experiments

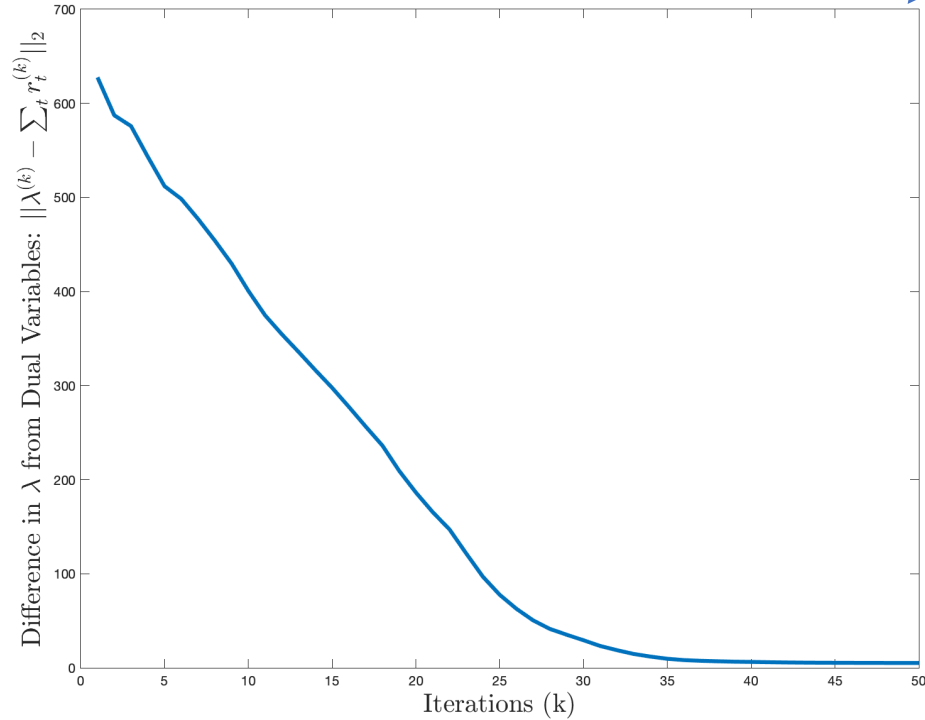
---

**Algorithm 1: Fixed Point Scheme**

---

**Input** : Tolerance  $\epsilon$ , Function  $G(\cdot)$  to calculate dual variables  
**Output**: Budget Perturbation Parameters  $\lambda$   
 $\lambda \leftarrow \mathbf{0}$  ;  
 $\mathbf{R} \leftarrow G(\lambda)$  ;  
 $q_i \leftarrow \sum_{t=1}^i r_{it}, \forall i$  ;  
**while**  $\|\lambda - \mathbf{q}\|_2 > \epsilon$  **do**  
     $\lambda_i \leftarrow \sum_{t=1}^i r_{it}, \forall i$  ;  
     $\mathbf{R} \leftarrow G(\lambda)$  ;  
     $q_i \leftarrow \sum_{t=1}^i r_{it}, \forall i$  ;  
**end**

---



**Solution Efficiency:** For fixed perturbations, **BP-SOP** is a convex optimization problem and so can be solved efficiently

**Convergence:** Numerical Experiments indicate convergence to a fixed point

## Future Research Directions

- Theoretically establish convergence guarantees for fixed-point scheme
- Determine whether fixed point scheme is PPAD-Complete to classify computational complexity of problem
- Extend results to discrete allocations and quantify loss in social objective
- Generalize framework to online setting with sequential customer arrival