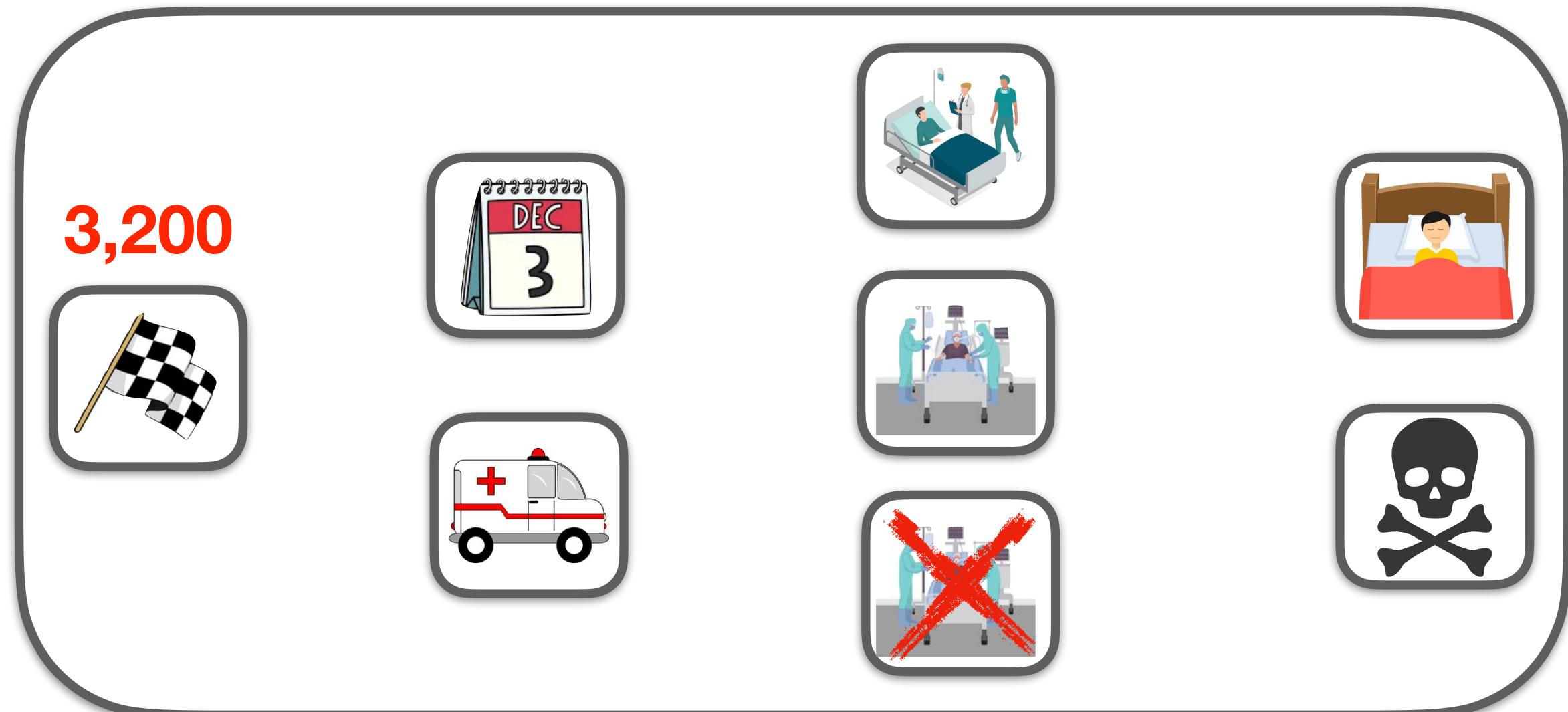


Weakly Coupled Counting DPs

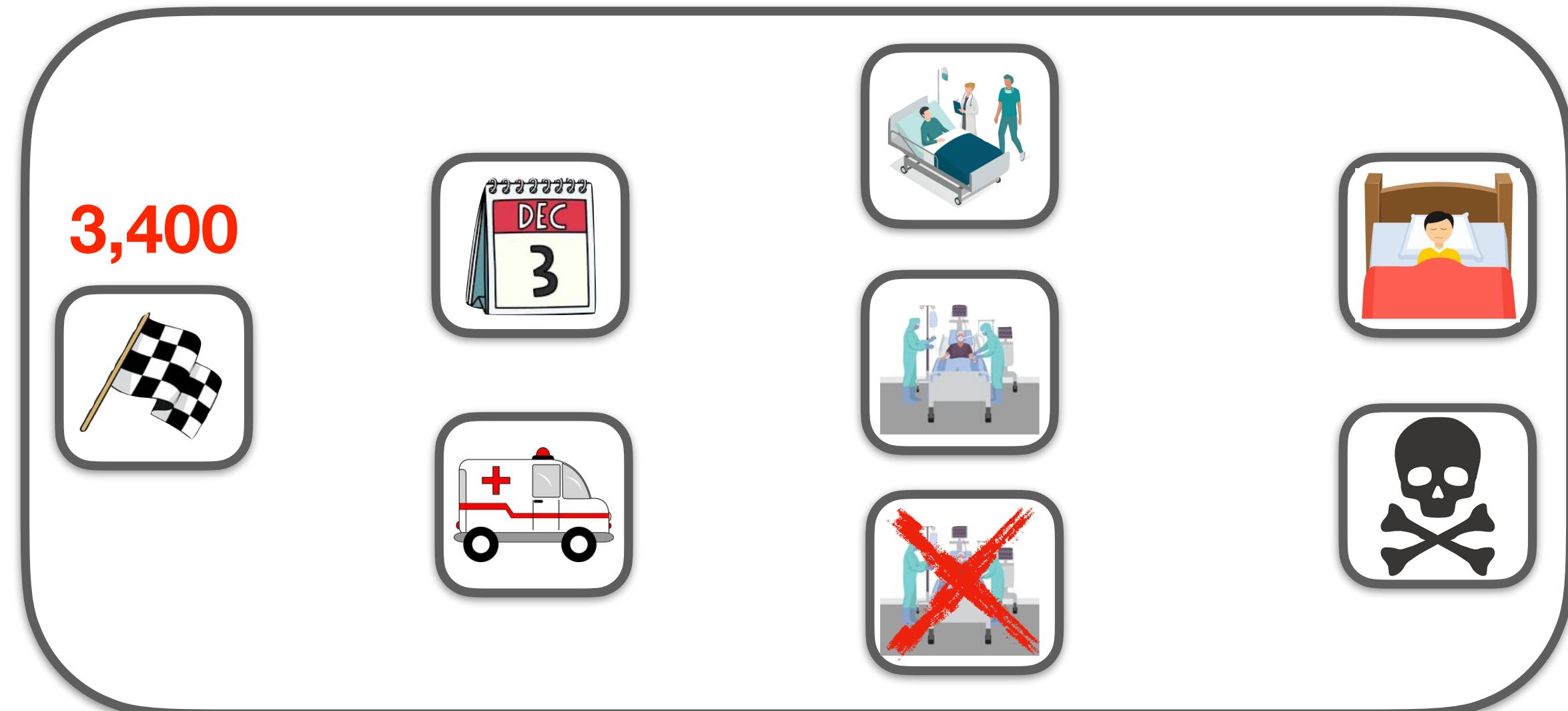
Time:



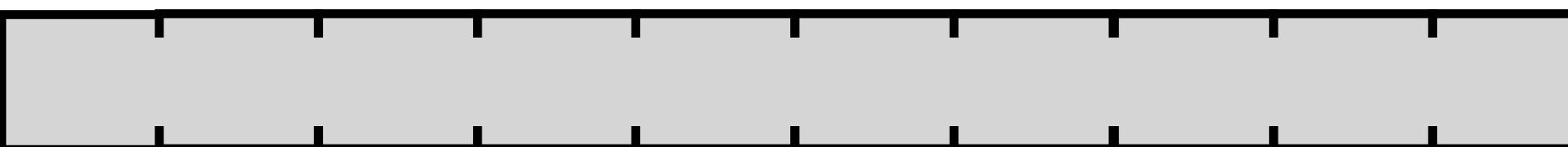
Patient group 1



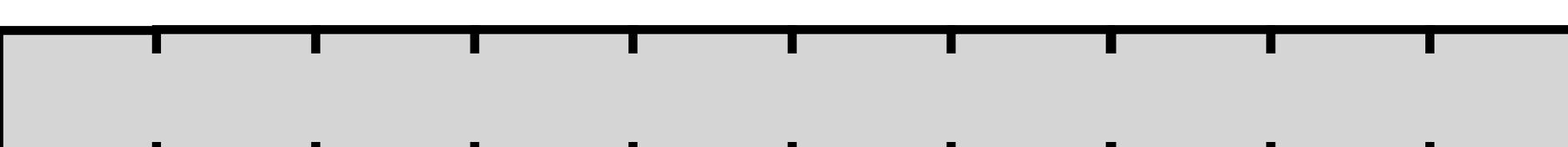
Patient group 2



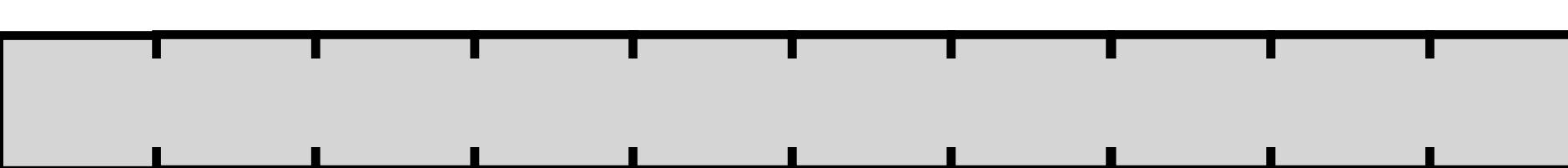
Senior doctors:



Junior doctors:



Nurses:



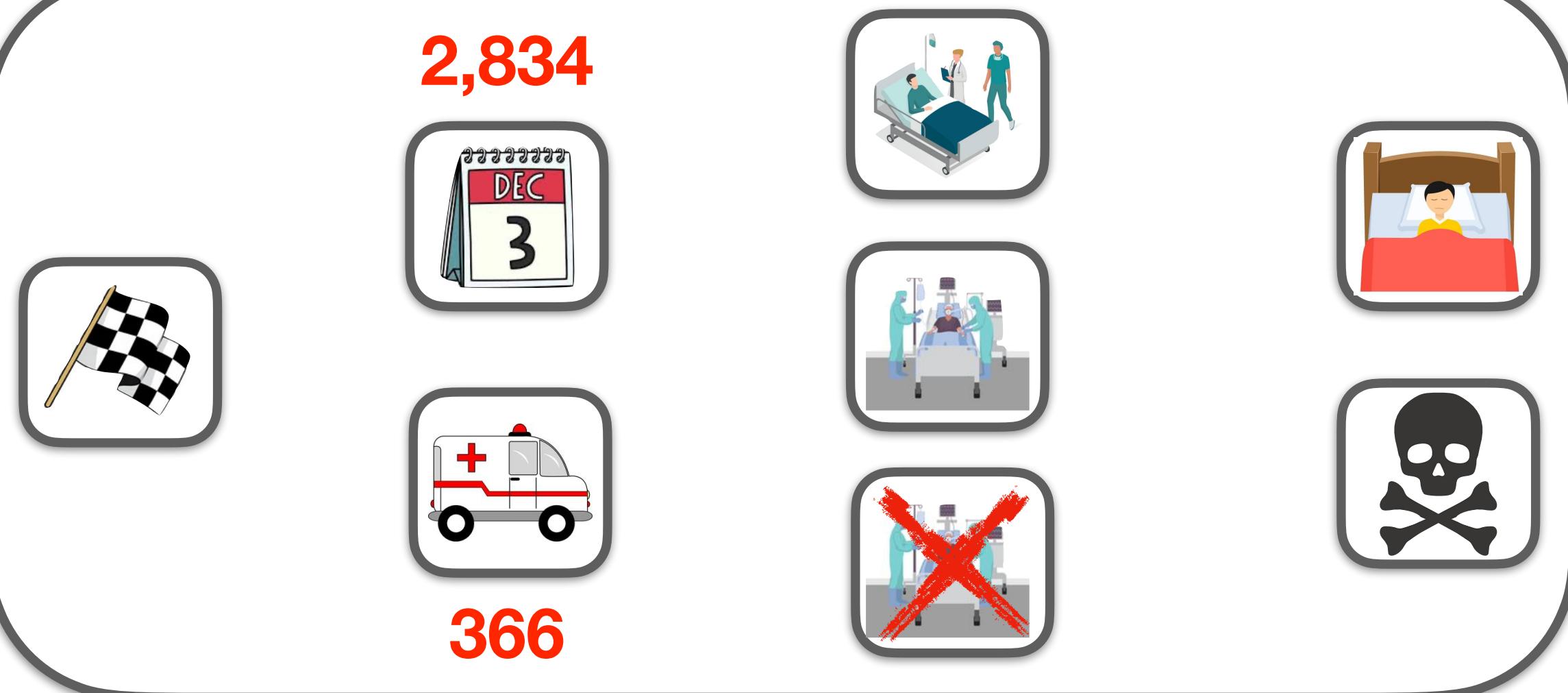
Weakly Coupled Counting DPs

Time:



Patient group 1

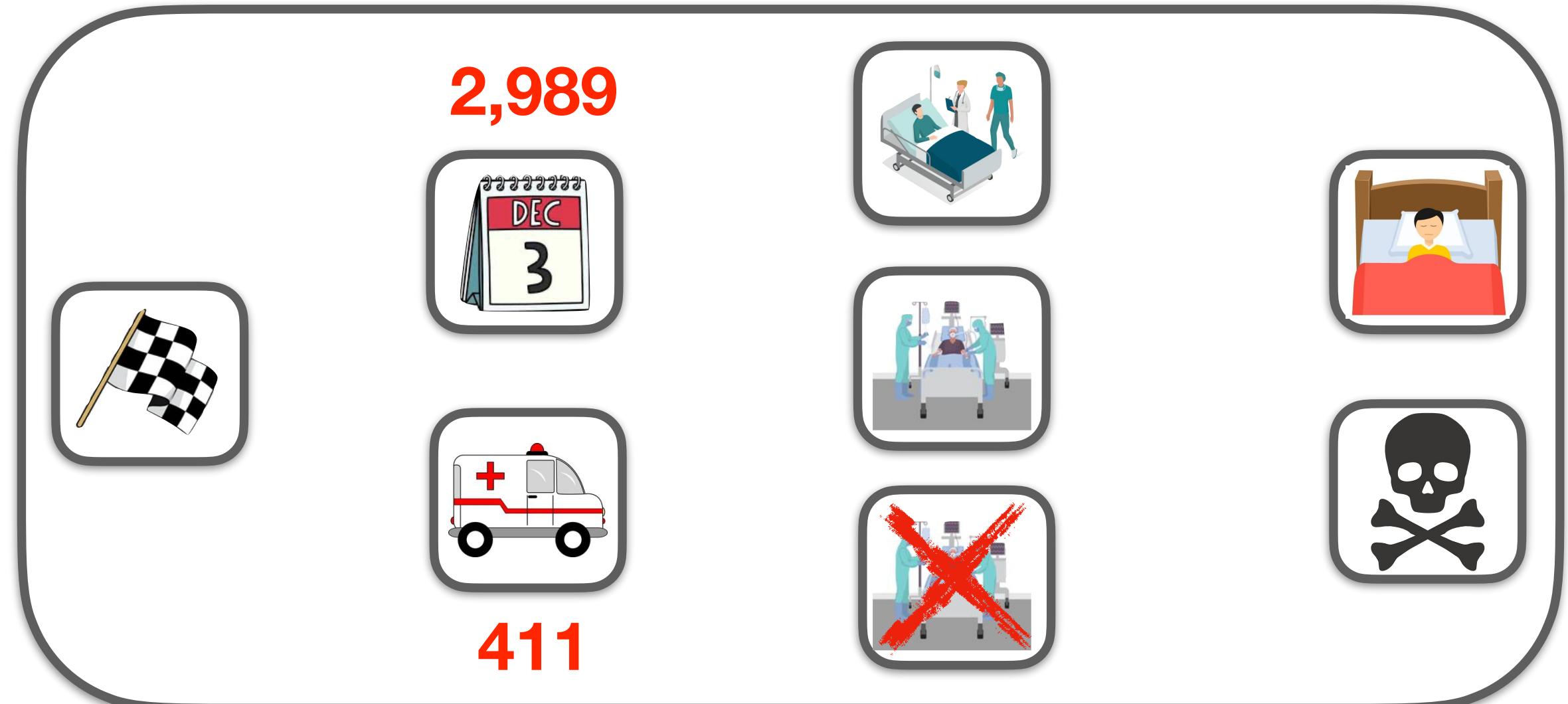
2,834



366

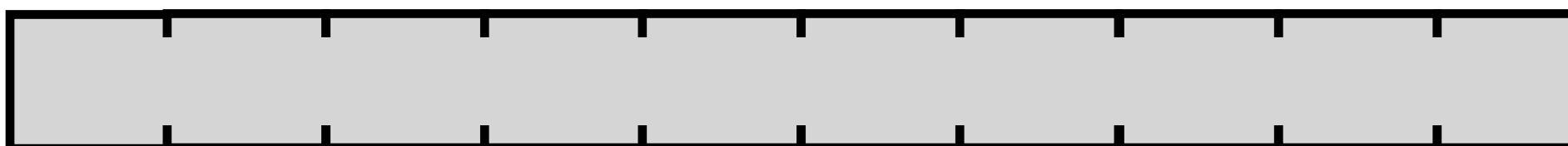
Patient group 2

2,989

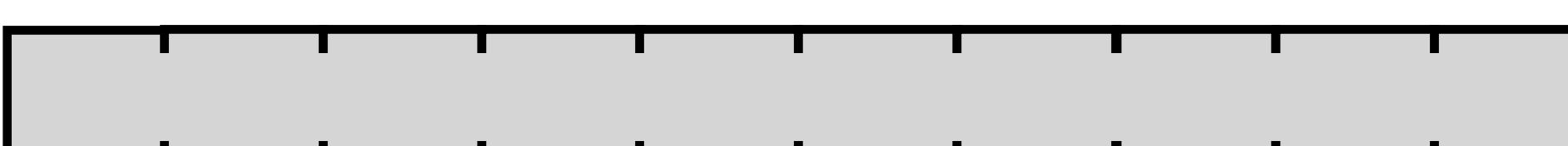


411

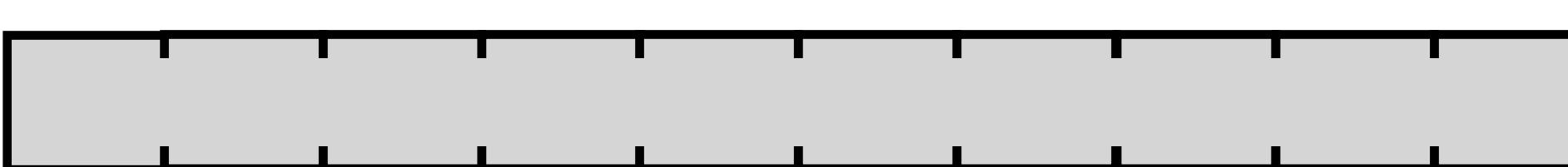
Senior doctors:



Junior doctors:



Nurses:

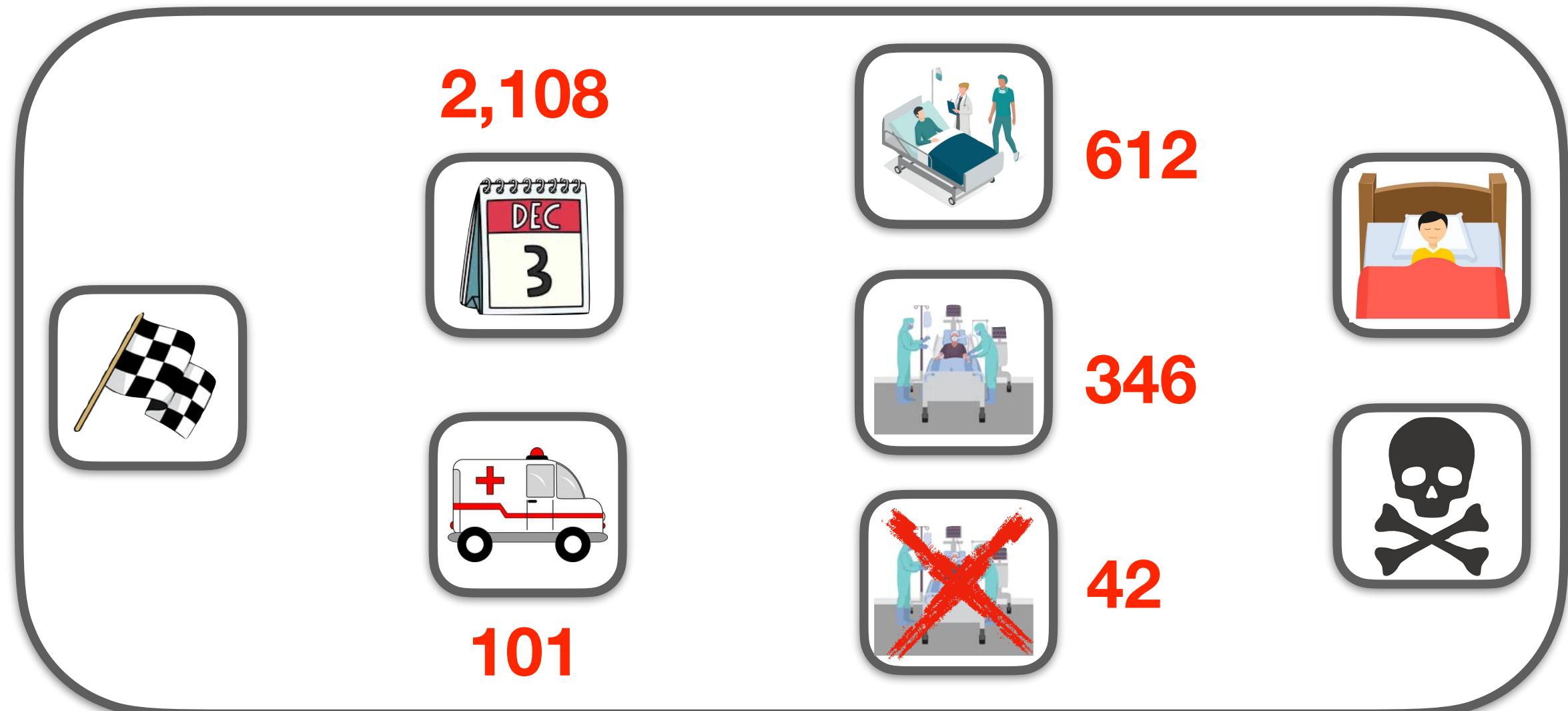


Weakly Coupled Counting DPs

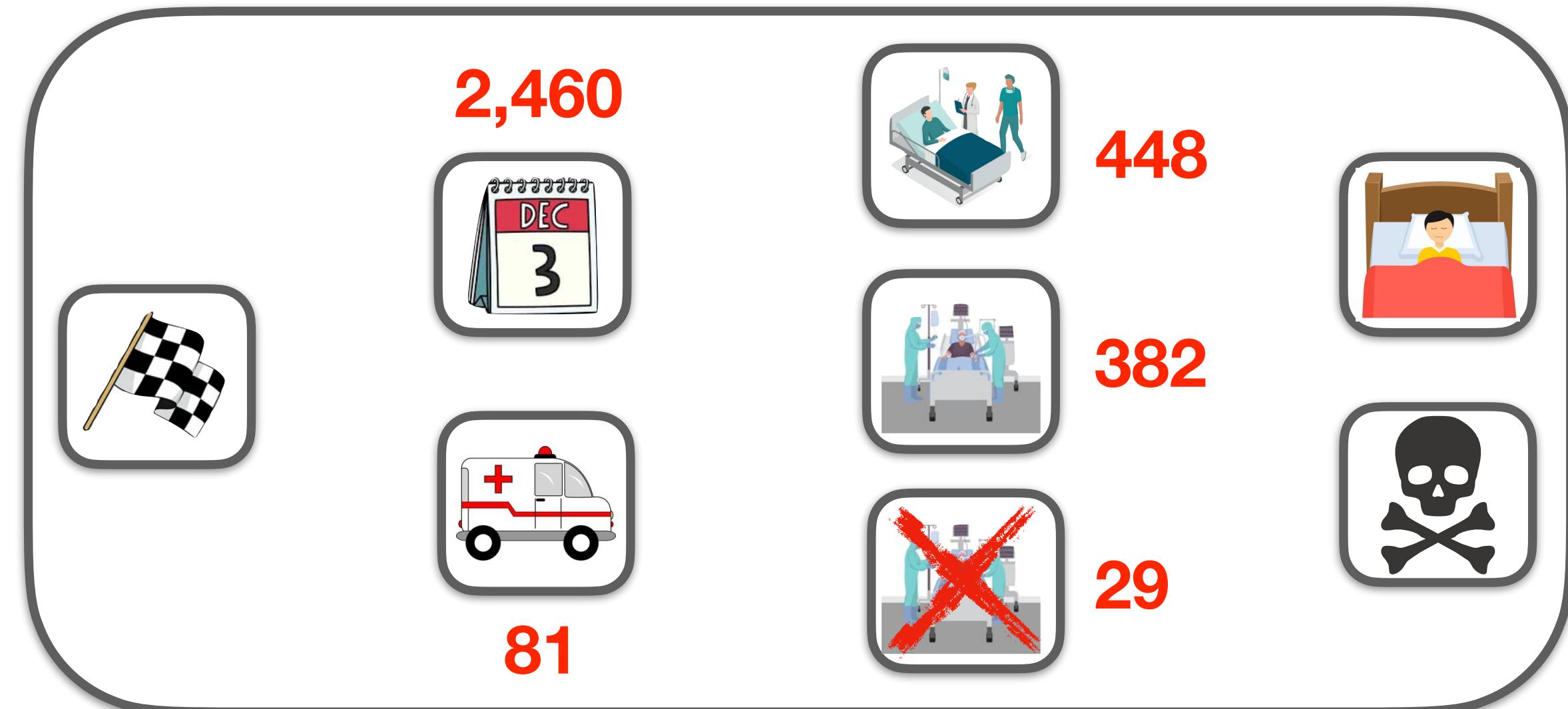
Time:



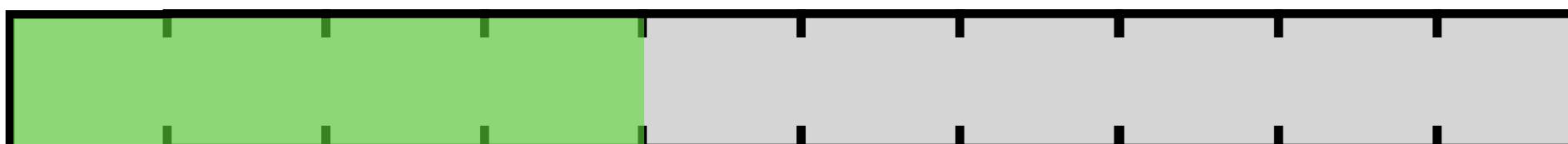
Patient group 1



Patient group 2



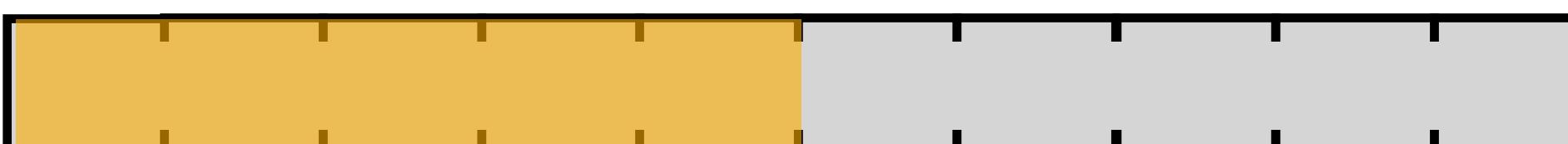
Senior doctors:



Junior doctors:



Nurses:

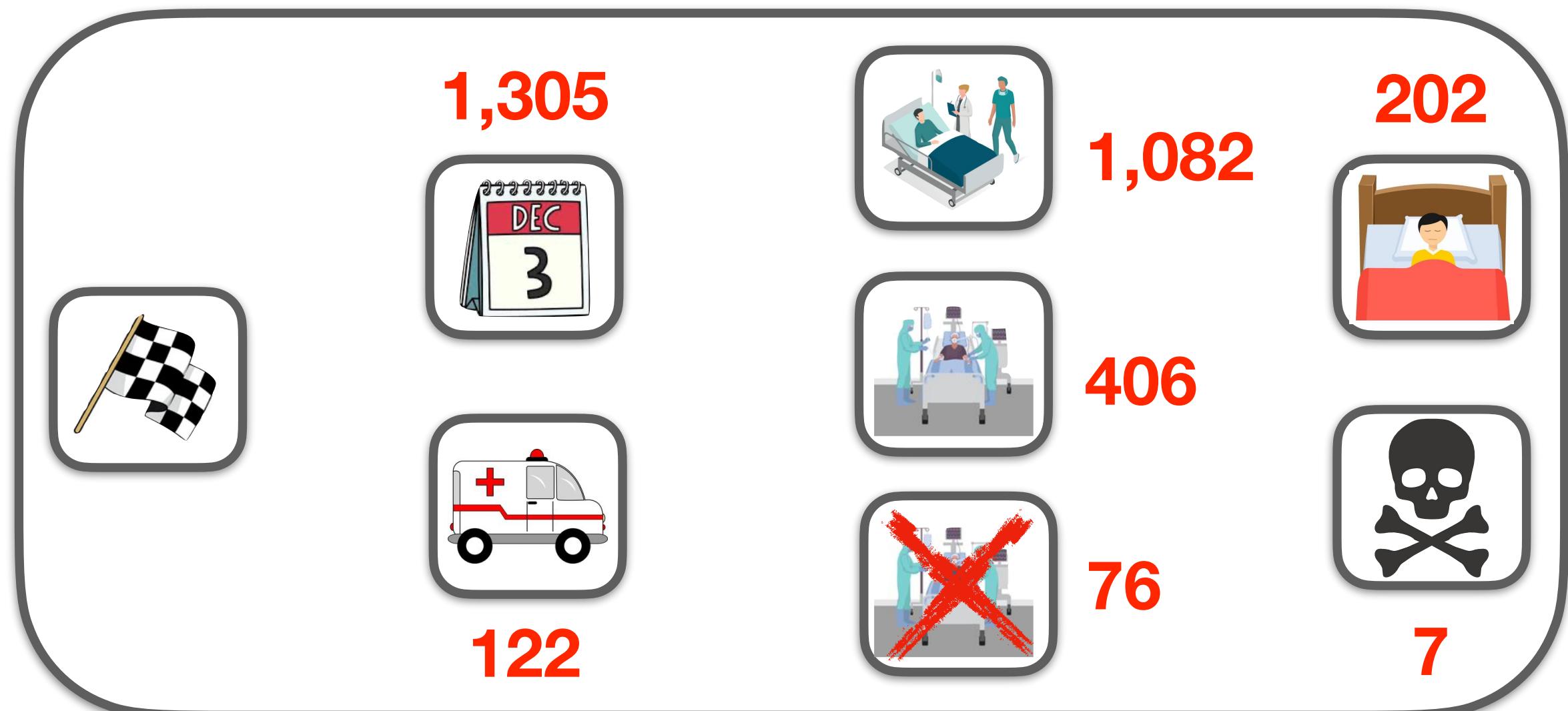


Weakly Coupled Counting DPs

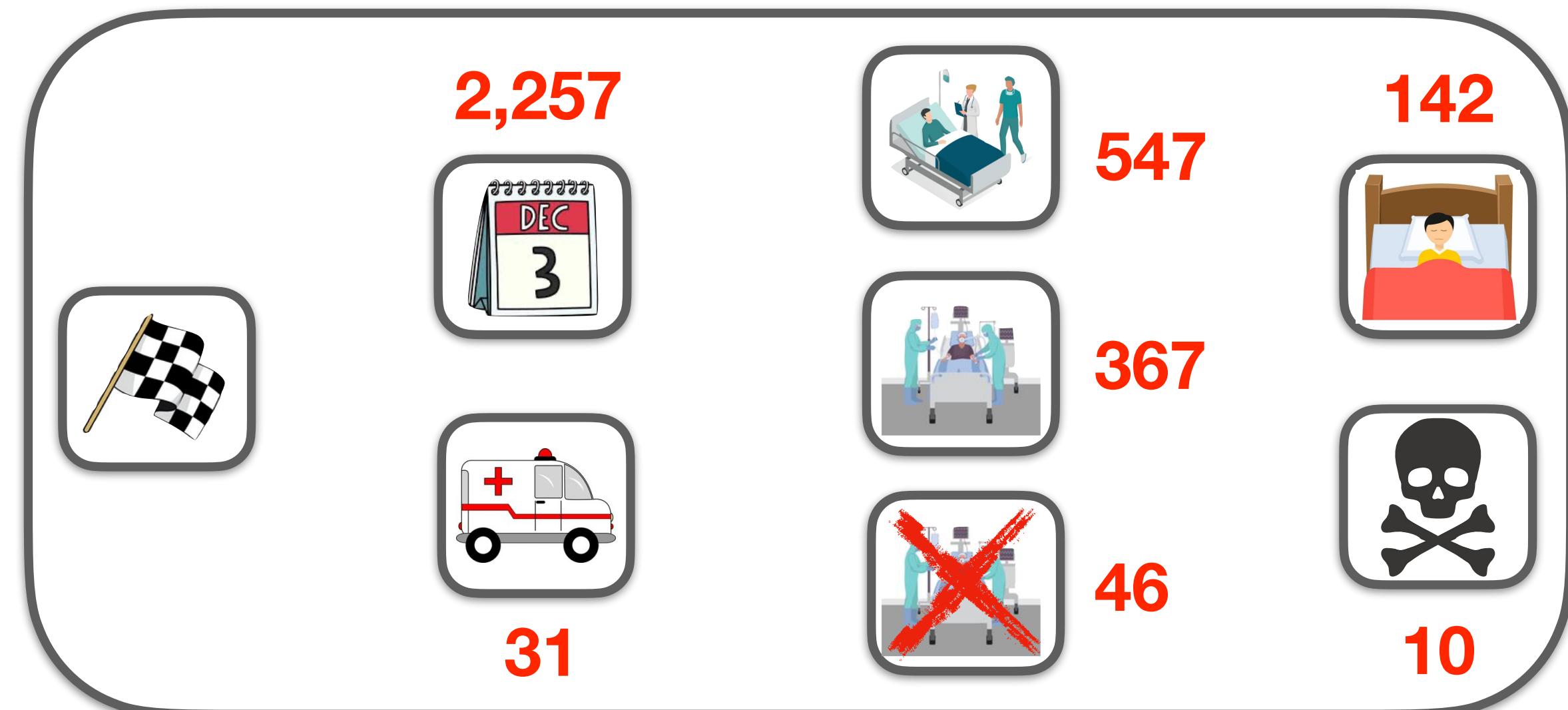
Time:



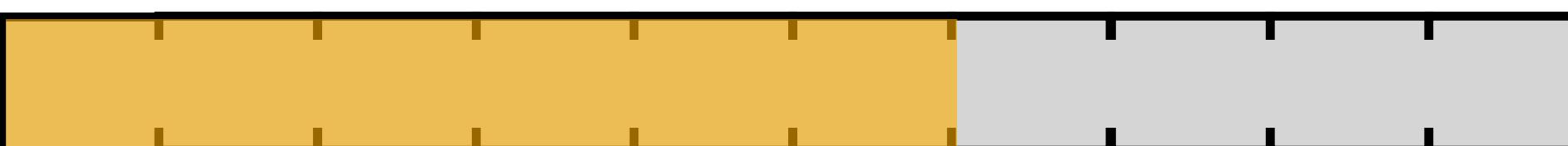
Patient group 1



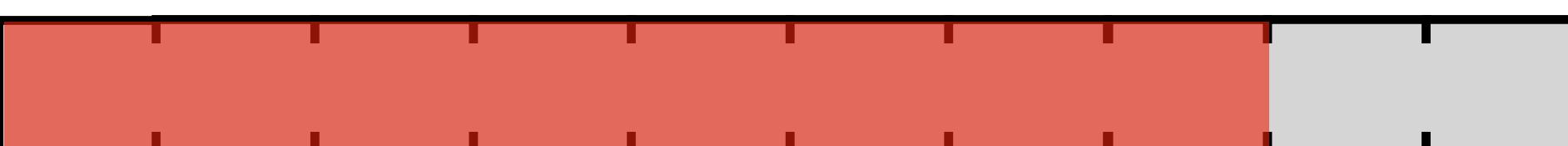
Patient group 2



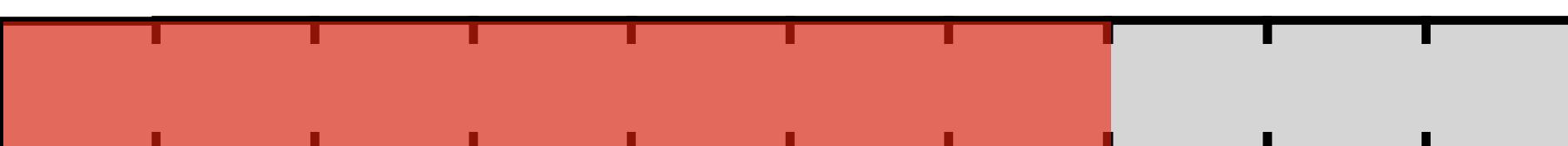
Senior doctors:



Junior doctors:



Nurses:



- 1 ~~Weakly Coupled Counting Dynamic Programs~~
- 2 The Fluid Approximation
- 3 Performance Guarantees
- 4 Case Study

An optimal policy for the fluid DP can be found via the fluid LP:

Fluid LP

maximize
 $\sigma, \pi \geq 0$

$$\sum_{t \in \mathcal{T}} \sum_{j \in \mathcal{J}} \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} r_{jt}(s, a) \cdot \pi_{tj}(s, a)$$

subject to

$$\sigma_{lj}(s) = n_j \cdot q_j(s) \quad \forall j, \forall s \in \mathcal{S}_j$$

$$\sigma_{t+1,j}(s') = \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} p_{jt}(s' | s, a) \cdot \pi_{tj}(s, a) \quad \forall j, \forall s' \in \mathcal{S}_j, \forall t$$

$$\sum_{j \in \mathcal{J}} \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} c_{tlj}(s, a) \cdot \pi_{tj}(s, a) \leq b_{tl} \quad \forall l, \forall t$$

$$\sum_{a \in \mathcal{A}_j} \pi_{tj}(s, a) = \sigma_{tj}(s) \quad \forall j, \forall s \in \mathcal{S}_j, \forall t$$

Fluid Approximation: Fluid LP

An optimal policy for the fluid DP can be found via the fluid LP:

Fluid LP

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subject to

$$\sum_{t \in \mathcal{T}} \sum_{j \in \mathcal{J}} \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} r_{jt}(s, a) \cdot \pi_{tj}(s, a)$$

$\sigma_{tj}(s)$: # of patients in patient group j
that are in state s in week t

$$\sigma_{t+1,j}(s') = \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} p_{jt}(s' | s, a) \cdot \pi_{tj}(s, a) \quad \forall j, \forall s' \in \mathcal{S}_j, \forall t$$

$$\sum_{j \in \mathcal{J}} \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} c_{tlj}(s, a) \cdot \pi_{tj}(s, a) \leq b_{tl} \quad \forall l, \forall t$$

$$\sum_{a \in \mathcal{A}_j} \pi_{tj}(s, a) = \sigma_{tj}(s) \quad \forall j, \forall s \in \mathcal{S}_j, \forall t$$

Fluid Approximation: Fluid LP

An optimal policy for the fluid DP can be found via the fluid LP:

Fluid LP

maximize
 $\sigma, \pi \geq 0$

subject to

$$\sum_{t \in \mathcal{T}} \sum_{i \in \mathcal{I}} \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} r_{jt}(s, a) \cdot \pi_{tj}(s, a)$$

$\sigma_{tj}(s)$: # of patients in patient group j
that are in state s in week t

$\pi_{tj}(s, a)$: # of patients from $\sigma_{tj}(s)$
that receive action a

$$\sum_{j \in \mathcal{J}} \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} \pi_{tj}(s, a) = \sigma_{tj}(s)$$

$$\sum_{a \in \mathcal{A}_j} \pi_{tj}(s, a) = \sigma_{tj}(s)$$

$$j, \forall s \in \mathcal{S}_j$$

$$j, \forall s' \in \mathcal{S}_j, \forall t$$

$$l, \forall t$$

$$\forall j, \forall s \in \mathcal{S}_j, \forall t$$

Fluid Approximation: Fluid LP

An optimal policy for the fluid DP can be found via the fluid LP:

Fluid LP

maximize
 $\sigma, \pi \geq 0$

$$\sum_{t \in \mathcal{T}} \sum_{j \in \mathcal{J}} \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} r_{jt}(s, a) \cdot \pi_{tj}(s, a)$$

objective function:
maximize rewards

subject to $\sigma_{lj}(s) = n_j \cdot q_j(s) \quad \forall j, \forall s \in \mathcal{S}_j$

$$\sigma_{t+1,j}(s') = \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} p_{jt}(s' | s, a) \cdot \pi_{tj}(s, a) \quad \forall j, \forall s' \in \mathcal{S}_j, \forall t$$

$$\sum_{j \in \mathcal{J}} \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} c_{tlj}(s, a) \cdot \pi_{tj}(s, a) \leq b_{tl} \quad \forall l, \forall t$$

$$\sum_{a \in \mathcal{A}_j} \pi_{tj}(s, a) = \sigma_{tj}(s) \quad \forall j, \forall s \in \mathcal{S}_j, \forall t$$

An optimal policy for the fluid DP can be found via the fluid LP:

Fluid LP

maximize
 $\sigma, \pi \geq 0$

$$\sum_{t \in \mathcal{T}} \sum_{j \in \mathcal{J}} \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} r_{jt}(s, a) \cdot \pi_{tj}(s, a)$$

initial states:
 must follow q

subject to

$$\sigma_{1j}(s) = n_j \cdot q_j(s)$$

$$\forall j, \forall s \in \mathcal{S}_j$$

$$\sigma_{t+1,j}(s') = \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} p_{jt}(s' | s, a) \cdot \pi_{tj}(s, a) \quad \forall j, \forall s' \in \mathcal{S}_j, \forall t$$

$$\sum_{j \in \mathcal{J}} \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} c_{tlj}(s, a) \cdot \pi_{tj}(s, a) \leq b_{tl} \quad \forall l, \forall t$$

$$\sum_{a \in \mathcal{A}_j} \pi_{tj}(s, a) = \sigma_{tj}(s) \quad \forall j, \forall s \in \mathcal{S}_j, \forall t$$

Fluid Approximation: Fluid LP

An optimal policy for the fluid DP can be found via the fluid LP:

Fluid LP

maximize
 $\sigma, \pi \geq 0$

$$\sum_{t \in \mathcal{T}} \sum_{j \in \mathcal{J}} \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} r_{jt}(s, a) \cdot \pi_{tj}(s, a)$$

transitions:
must follow p

subject to

$$\sigma_{lj}(s) = n_j \cdot q_j(s)$$

$$\forall j, \forall s \in \mathcal{S}_j$$

$$\sigma_{t+1,j}(s') = \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} p_{jt}(s' | s, a) \cdot \pi_{tj}(s, a)$$

$$\forall j, \forall s' \in \mathcal{S}_j, \forall t$$

$$\sum_{j \in \mathcal{J}} \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} c_{tlj}(s, a) \cdot \pi_{tj}(s, a) \leq b_{tl}$$

$$\forall l, \forall t$$

$$\sum_{a \in \mathcal{A}_j} \pi_{tj}(s, a) = \sigma_{tj}(s)$$

$$\forall j, \forall s \in \mathcal{S}_j, \forall t$$

Fluid Approximation: Fluid LP

An optimal policy for the fluid DP can be found via the fluid LP:

Fluid LP

maximize
 $\sigma, \pi \geq 0$

$$\sum_{t \in \mathcal{T}} \sum_{j \in \mathcal{J}} \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} r_{jt}(s, a) \cdot \pi_{tj}(s, a)$$

resources:
budgets must be kept

subject to

$$\sigma_{lj}(s) = n_j \cdot q_j(s) \quad \forall j, \forall s \in \mathcal{S}_j$$

$$\sigma_{t+1,j}(s') = \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} p_t(s' | s, a) \cdot \pi_{tj}(s, a) \quad \forall j, \forall s' \in \mathcal{S}_j, \forall t$$

$$\sum_{j \in \mathcal{J}} \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} c_{tlj}(s, a) \cdot \pi_{tj}(s, a) \leq b_{tl} \quad \forall l, \forall t$$

$$\sum_{a \in \mathcal{A}_j} \pi_{tj}(s, a) = \sigma_{tj}(s) \quad \forall j, \forall s \in \mathcal{S}_j, \forall t$$

An optimal policy for the fluid DP can be found via the fluid LP:

Fluid LP

maximize
 $\sigma, \pi \geq 0$

$$\sum_{t \in \mathcal{T}} \sum_{j \in \mathcal{J}} \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} r_{jt}(s, a) \cdot \pi_{tj}(s, a)$$

“flow preservation”:
 we cannot “drop” patients

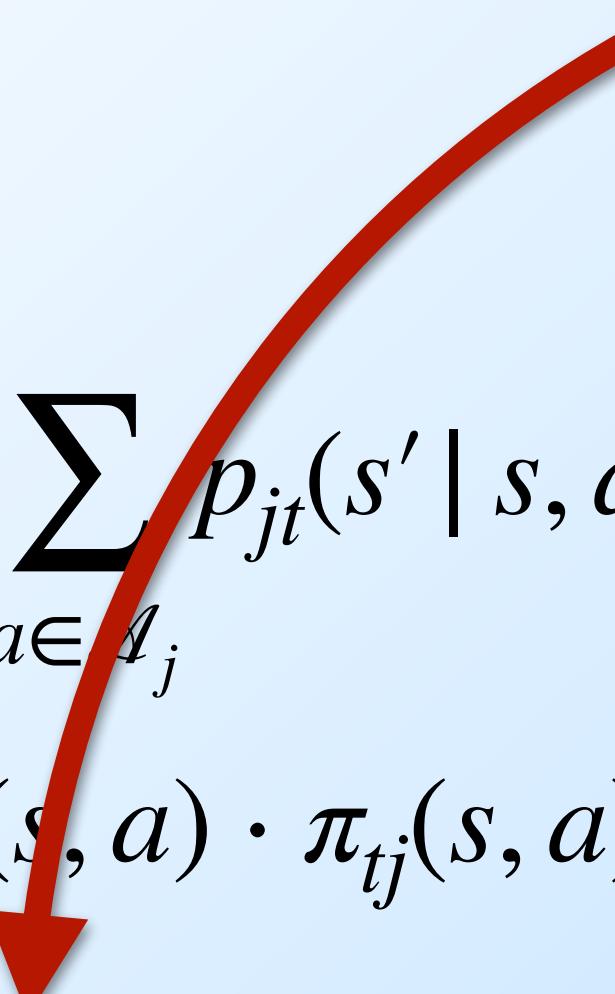
subject to

$$\sigma_{lj}(s) = n_j \cdot q_j(s) \quad \forall j, \forall s \in \mathcal{S}_j$$

$$\sigma_{t+1,j}(s') = \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} p_{jt}(s' | s, a) \cdot \pi_{tj}(s, a) \quad \forall j, \forall s' \in \mathcal{S}_j, \forall t$$

$$\sum_{j \in \mathcal{J}} \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} c_{tlj}(s, a) \cdot \pi_{tj}(s, a) \leq b_{tl} \quad \forall l, \forall t$$

$$\sum_{a \in \mathcal{A}_j} \pi_{tj}(s, a) = \sigma_{tj}(s) \quad \forall j, \forall s \in \mathcal{S}_j, \forall t$$



Fluid Approximation: Fluid LP

An optimal policy for the fluid DP can be found via the fluid LP:

Fluid LP

maximize
 $\sigma, \pi \geq 0$

$$\sum_{t \in \mathcal{T}} \sum_{j \in \mathcal{J}} \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} r_{jt}(s, a) \cdot \pi_{tj}(s, a)$$

subject to

Proposition

The fluid LP constitutes a **relaxation** of the
weakly coupled counting DP.

$$\sum_{j \in \mathcal{J}} \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} c_{tlj}(s, a) \cdot \pi_{tj}(s, a) \leq b_{tl} \quad \forall l, \forall t$$

$$\sum_{a \in \mathcal{A}_j} \pi_{tj}(s, a) = \sigma_{tj}(s) \quad \forall j, \forall s \in \mathcal{S}_j, \forall t$$

- 1 ~~Weakly Coupled Counting Dynamic Programs~~
- 2 ~~The Fluid Approximation~~
- 3 Performance Guarantees
- 4 Case Study

Recovery of Individual DP Policies: Randomized Approach

Algorithm

- 1 Solve Fluid LP to obtain the solution $(\sigma^\star, \pi^\star)$.
- 2 At each time stage $t \in \mathcal{T}$, implement for each patient of group $j \in \mathcal{J}$ the action $a \in \mathcal{A}_j$ with probability

$$\frac{\pi_{tj}^\star(s, a)}{\sigma_{tj}^\star(s)}$$

if the patient is in state $s \in \mathcal{S}_j$.

Recovery of Individual DP Policies: Randomized Approach

Theorem (Expected Case)

Let θ^* be **expected total reward** of our randomized policy and θ^{DP} the **expected total reward** of the optimal DP policy. Then

$$\theta^* \geq \theta^{\text{DP}}$$

as well as, **with high probability**,

$$\sum_{j \in \mathcal{J}} \sum_{i=1}^{n_j} c_{tlj}(\tilde{s}_{t,(j,i)}, \tilde{a}_{t,(j,i)}) \leq b_{tl} + \epsilon \cdot \sum_{j \in \mathcal{J}} n_j \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} c_{tlj}(s, a) \quad \forall t, \forall l.$$

Recovery of Individual DP Policies: Randomized Approach

Theorem (Worst Case)

Let $\tilde{\theta}^*$ be *random total reward* of our randomized policy and θ^{DP} the *expected total reward* of the optimal DP policy. Then, **with high probability**,

$$\tilde{\theta}^* \geq \theta^{\text{DP}} - \epsilon \cdot \sum_{t \in \mathcal{T}} \sum_{j \in \mathcal{J}} n_j \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} r_{jt}(s, a)$$

as well as

$$\sum_{j \in \mathcal{J}} \sum_{i=1}^{n_j} c_{tlj}(\tilde{s}_{t,(j,i)}, \tilde{a}_{t,(j,i)}) \leq b_{tl} + \epsilon \cdot \sum_{j \in \mathcal{J}} n_j \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} c_{tlj}(s, a) \quad \forall t, \forall l.$$

Recovery of Individual DP Policies: Deterministic Approach

Algorithm

- 1 Solve Fluid LP to obtain the solution $(\sigma^\star, \pi^\star)$.
- 2 At each time stage $t \in \mathcal{T}$, implement for group $j \in \mathcal{J}$ each action a in each state s

$$\left\{ \Pr \left[\frac{\pi_{tj}^\star(s, a)}{\sigma_{tj}^\star(s)} \cdot \sigma_{tj}(s) \right] \right\}_{s,a} \text{ many times}$$

if the group is in state $\sigma_j \in \mathfrak{S}_j$, where \Pr is a projection.

Recovery of Individual DP Policies: Deterministic Approach

Theorem (Expected Case)

Let θ^* be **expected total reward** of our deterministic policy and θ^{DP} the **expected total reward** of the optimal DP policy. Then

$$\theta^* \geq \theta^{\text{DP}} - \sum_{t \in \mathcal{T}} \sum_{j \in \mathcal{J}} \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} r_{jt}(s, a) \cdot \epsilon_1$$

as well as, **with high probability**, for all t and l ,

$$\sum_{j \in \mathcal{J}} \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} c_{tlj}(s, a) \cdot [\pi_t^*(\tilde{\sigma}_t^*)]_j(s, a) \leq b_{tl} + \sum_{j \in \mathcal{J}} (1 + \epsilon_2 n_j) \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} c_{tlj}(s, a) \cdot \epsilon_1.$$

Recovery of Individual DP Policies: Deterministic Approach

Theorem (Worst Case)

Let $\tilde{\theta}^*$ be *random total reward* of our deterministic policy and θ^{DP} the *expected total reward* of the optimal DP policy. Then, **with high probability**,

$$\tilde{\theta}^* \geq \theta^{\text{DP}} - \sum_{t \in \mathcal{T}} \sum_{j \in \mathcal{J}} \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} r_{jt}(s, a) \cdot (1 + \epsilon_2 n_j) \cdot \epsilon_1$$

as well as, for all t and l ,

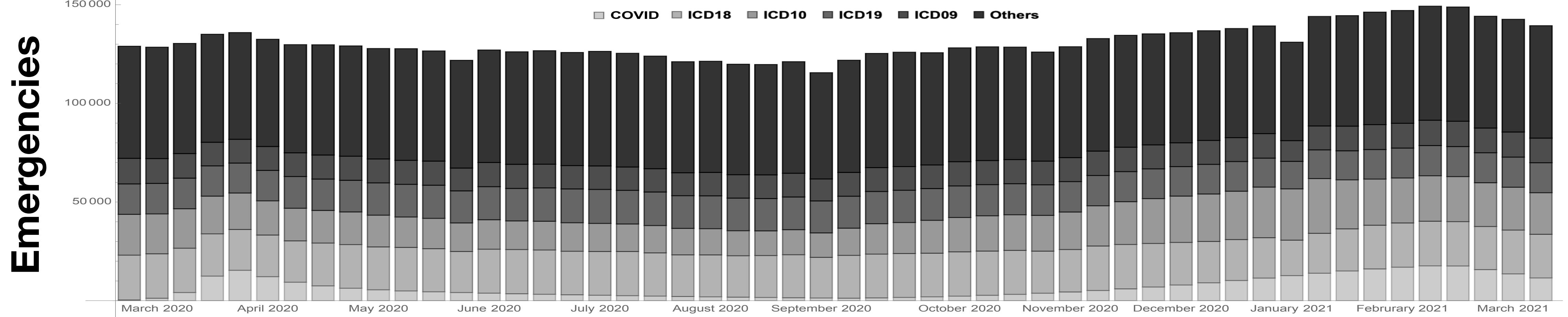
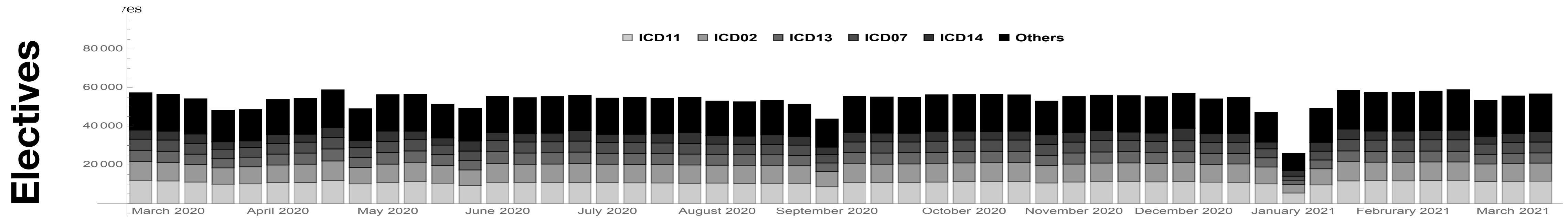
$$\sum_{j \in \mathcal{J}} \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} c_{tlj}(s, a) \cdot [\pi_t^*(\tilde{\theta}_t^*)]_j(s, a) \leq b_{tl} + \sum_{j \in \mathcal{J}} (1 + \epsilon_2 n_j) \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} c_{tlj}(s, a) \cdot \epsilon_1.$$

Agenda

- 1 ~~Weakly Coupled Counting Dynamic Programs~~
- 2 ~~The Fluid Approximation~~
- 3 ~~Performance Guarantees~~
- 4 Case Study

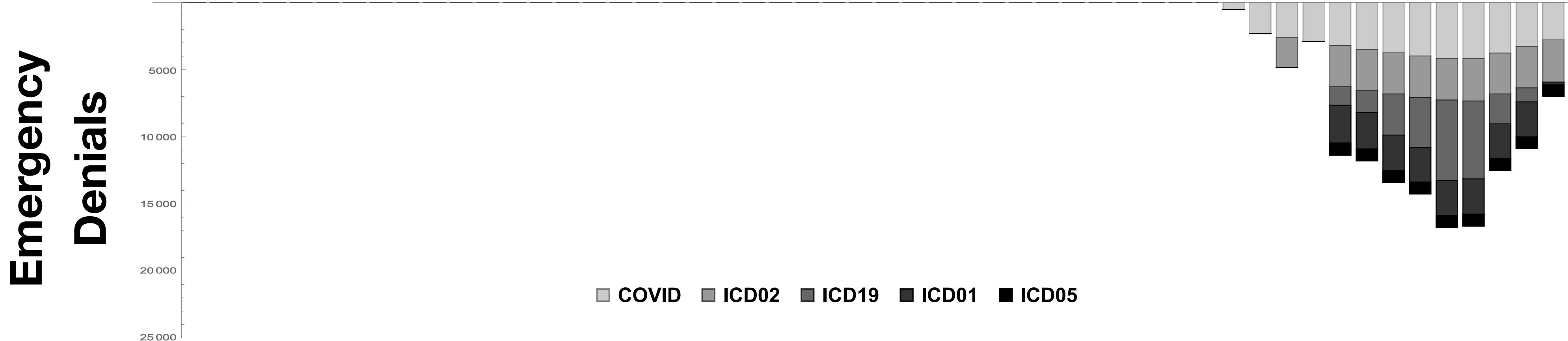
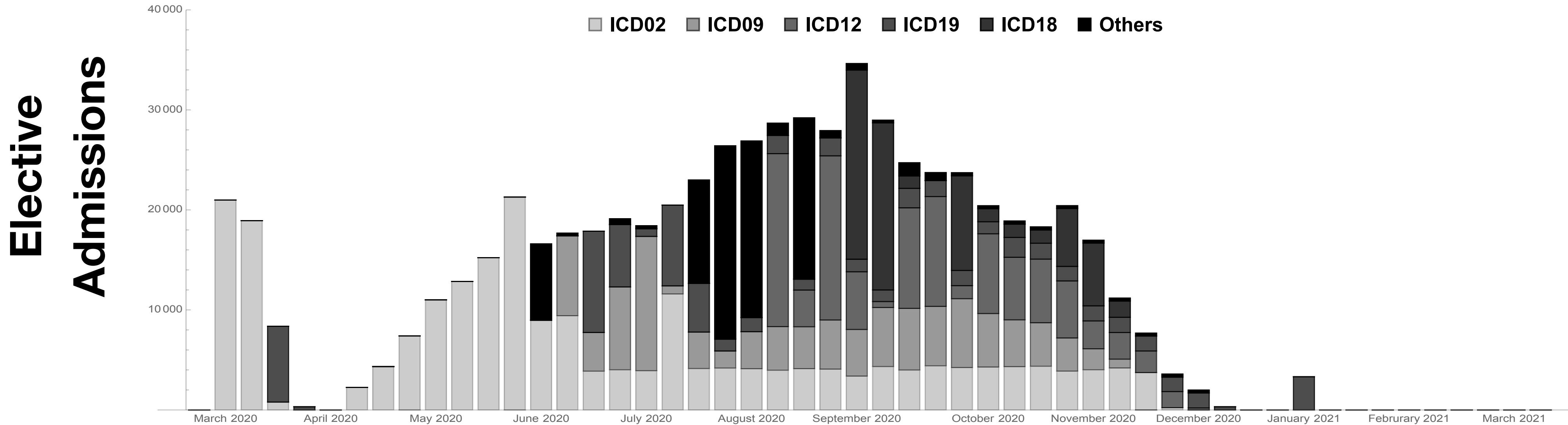
Input Data

	Capacity				Staff-to-bed Ratio		
	Beds	Senior Doctors	Junior Doctors	Nurses	Senior Doctors	Junior Doctors	Nurses
G&A	102,186	10,764	8,539	43,214	15	15	5
CC	4,122	1,013	963	18,856	15	8	1



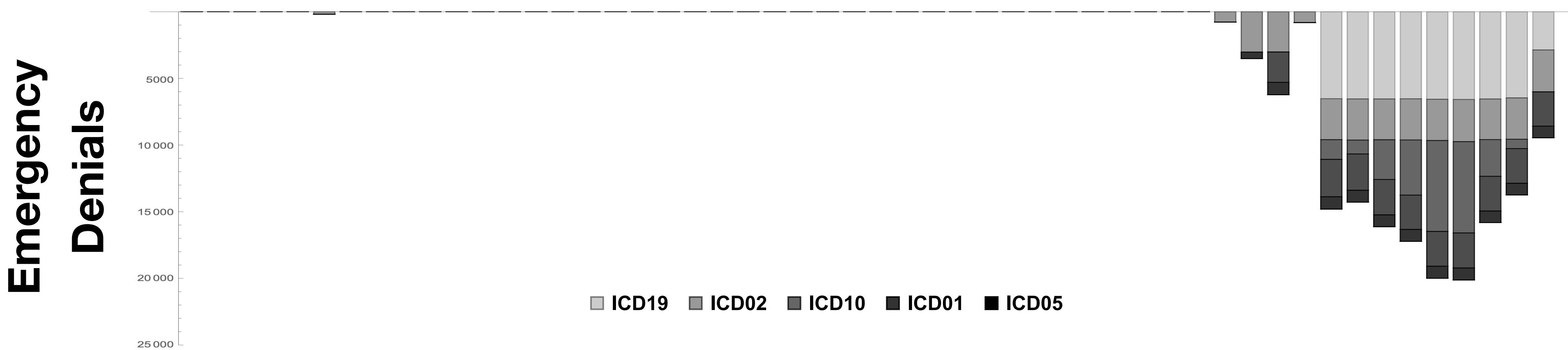
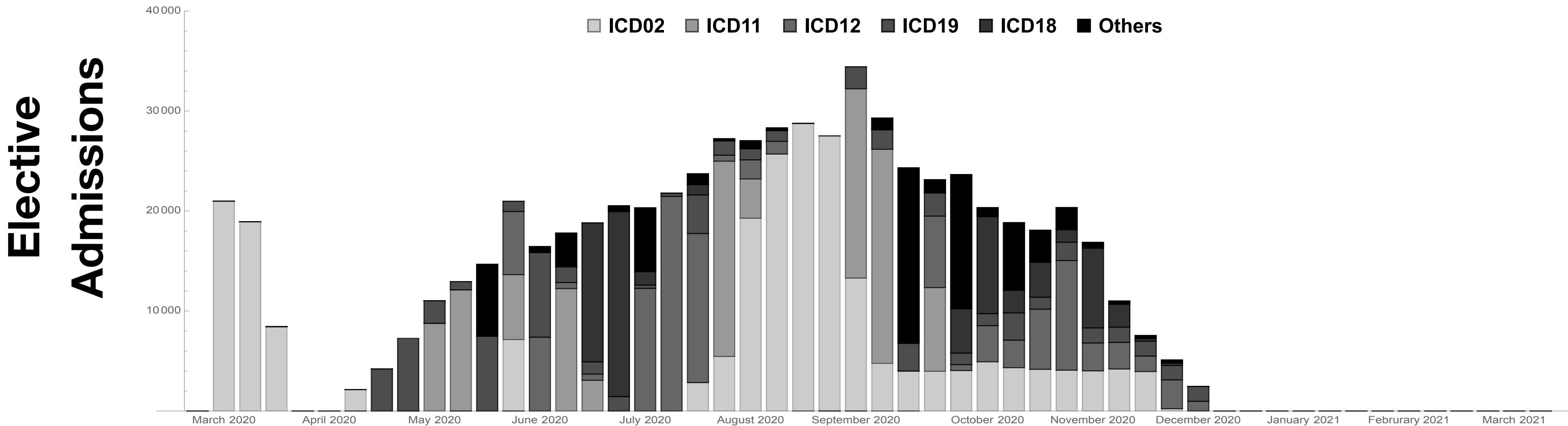
Admissions/Denials: Optimization vs Gov. Policy

Optimized Schedule



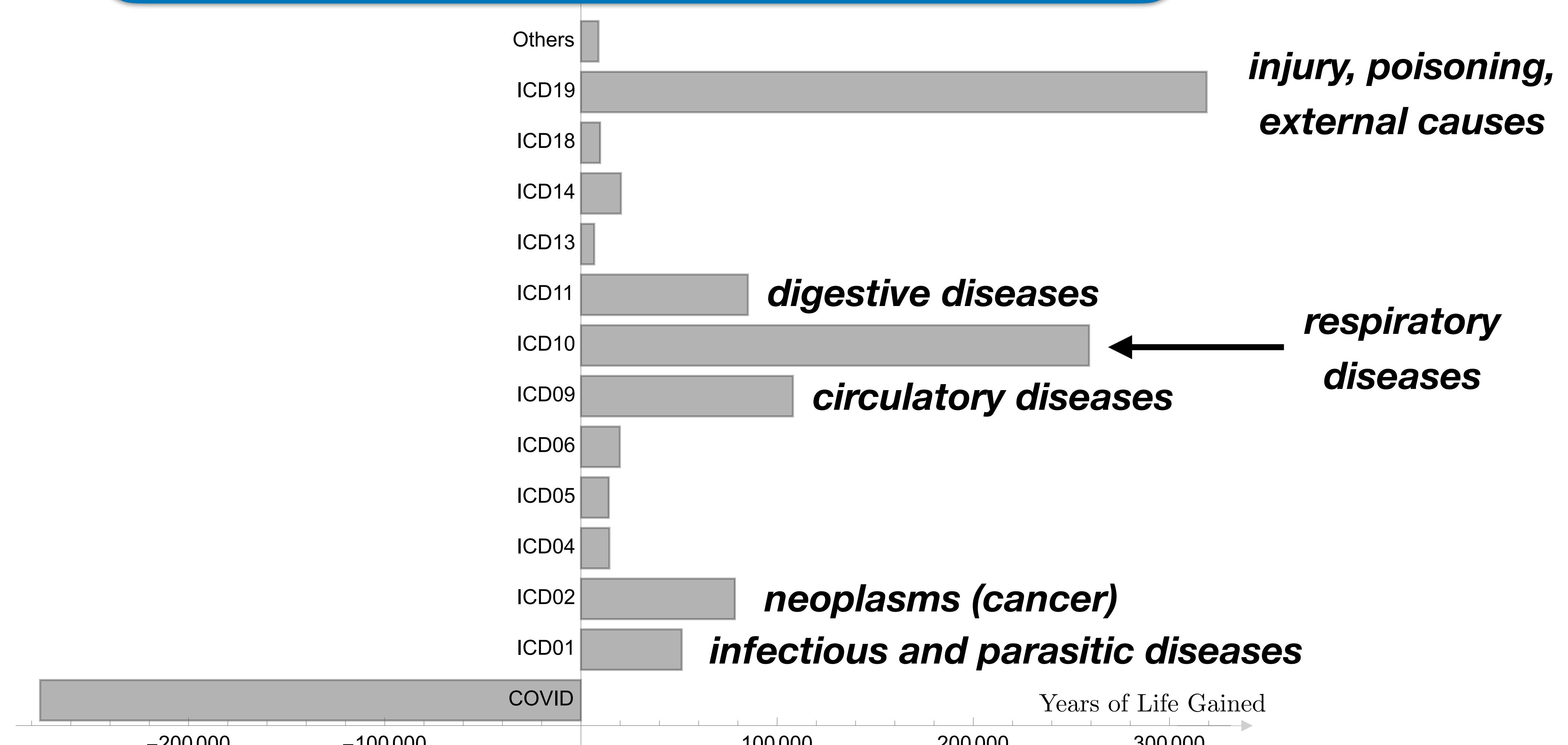
Admissions/Denials: Optimization vs Gov. Policy

Simulation of Government Policy



Comparison: Years of Life Gained

Years of Life Gained by Optimized Schedule



*injury, poisoning,
external causes*

digestive diseases

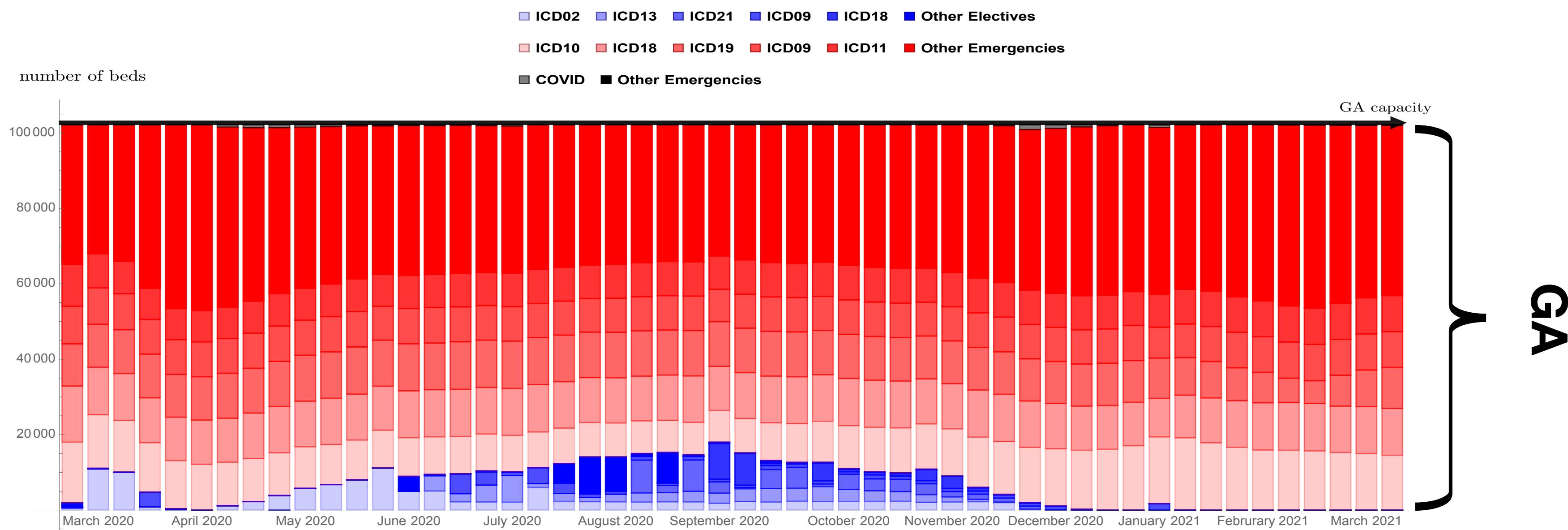
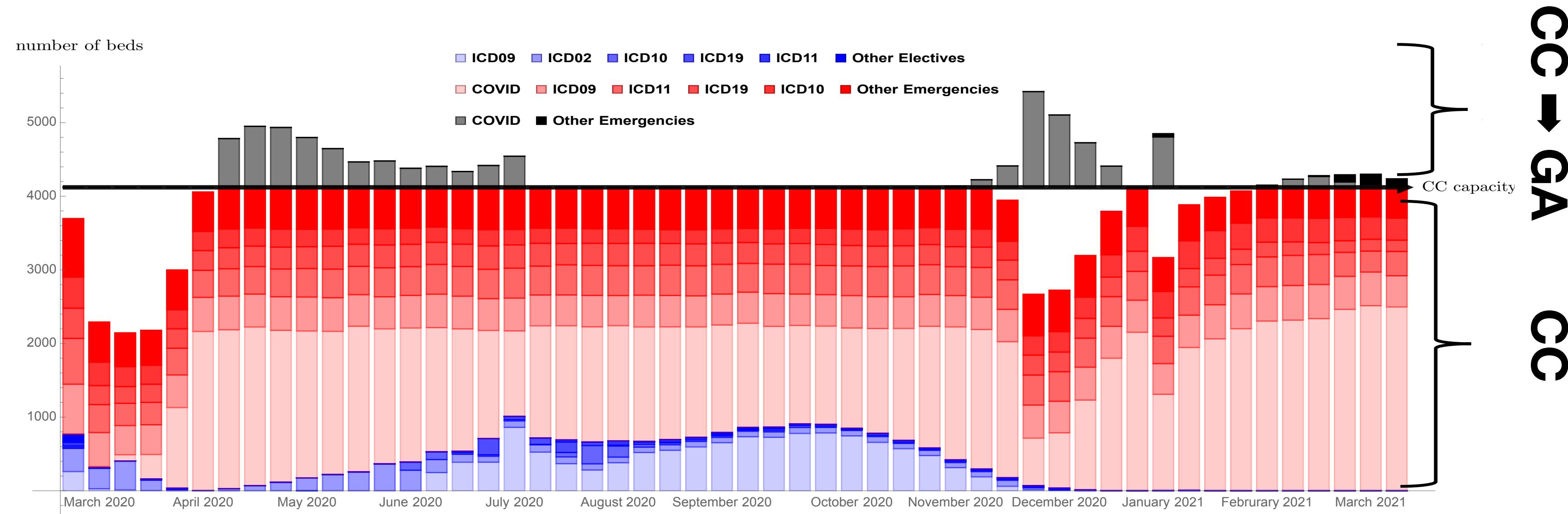
***respiratory
diseases***

circulatory diseases

neoplasms (cancer)

infectious and parasitic diseases

Optimized Schedule: Hospital Occupancy



Deterministic Rounding Approach:

- * YLL +0.02%, G&A +0.04%, CC +0.31%

Randomization Approach:

- * YLL -0.01%, G&A +0.05%, CC +1.56%

Hospital Occupancy:

		Mar	Apr	May	Jun	Jul	Aug	Sep	Oct	Nov	Dec	Jan	Feb	Mar
LP	G&A	100%	100%	100%	100%	100%	100%	100%	100%	100%	100%	100%	100%	100%
	CC	63%	93%	100%	100%	100%	100%	100%	100%	100%	76%	91%	99%	100%
DR	G&A	100%	101%	100%	100%	100%	100%	100%	100%	100%	100%	99%	101%	100%
	CC	63%	94%	99%	100%	100%	100%	100%	100%	98%	99%	78%	93%	101%
R	G&A	100%	100%	100%	100%	100%	100%	100%	100%	100%	100%	99%	100%	100%
	CC	66%	95%	107%	108%	101%	104%	97%	96%	100%	76%	90%	100%	101%

Bibliography

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D'Aeth et al. *Optimal Hospital Care Scheduling During the SARS-CoV-2 Pandemic.*
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Esma



Huikang



Shubhy



Stefano



Wolfram

BACKUP SLIDES

Recovery of Individual DP Policies: Randomized Approach

Proof (Expected Total Reward): $\theta^* \geq \theta^{\text{DP}}$

$$\theta^* = \mathbb{E} \left[\sum_{t \in \mathcal{T}} \sum_{j \in \mathcal{J}} \sum_{i=1}^{n_j} r_{jt}(\tilde{s}_{t,(j,i)}, \tilde{a}_{t,(j,i)}) \right]$$

random state and action of
DP i in group j at time t

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$$= \mathbb{E} \left[\sum_{t \in \mathcal{T}} \sum_{j \in \mathcal{J}} \sum_{i=1}^{n_j} \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} r_{jt}(s, a) \cdot \mathbf{1} [\tilde{s}_{t,(j,i)} = s \wedge \tilde{a}_{t,(j,i)} = a] \right]$$



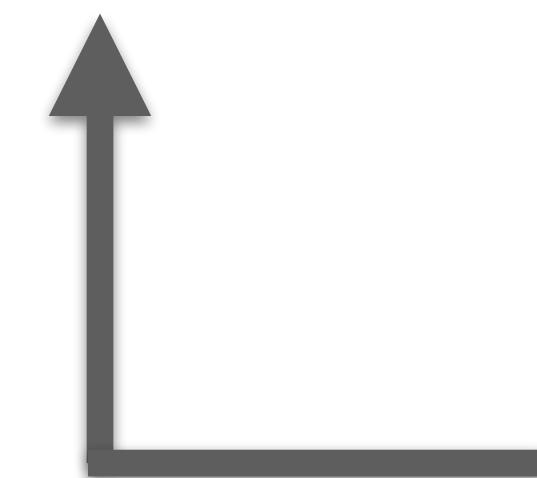
summation over all
possible state-action pairs

Recovery of Individual DP Policies: Randomized Approach

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$$= \sum_{t \in \mathcal{T}} \sum_{j \in \mathcal{J}} \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} r_{jt}(s, a) \cdot \mathbb{E} \left[\sum_{i=1}^{n_j} \mathbf{1} \left[\tilde{s}_{t,(j,i)} = s \wedge \tilde{a}_{t,(j,i)} = a \right] \right]$$



move expectation
inside

Recovery of Individual DP Policies: Randomized Approach

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$$= \sum_{t \in \mathcal{T}} \sum_{j \in \mathcal{J}} \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} r_{jt}(s, a) \cdot \pi_{tj}^{\text{LP}}(s, a)$$



use definition of
expectation

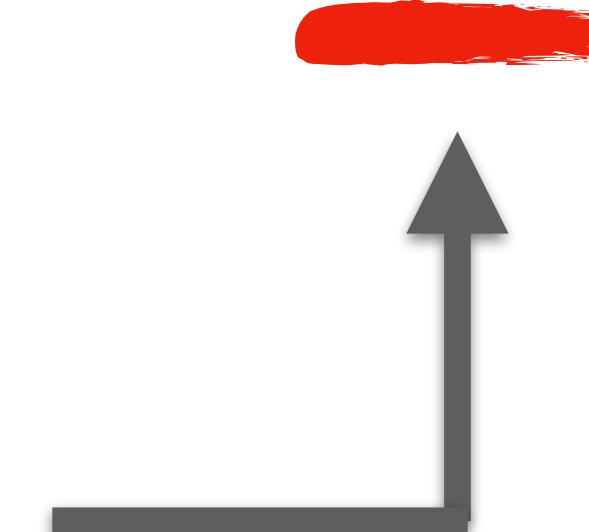
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$$= \sum_{t \in \mathcal{T}} \sum_{j \in \mathcal{J}} \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} r_{jt}(s, a) \cdot \pi_{tj}^{\text{LP}}(s, a) = \theta^{\text{LP}}$$

use definition of
objective function



Recovery of Individual DP Policies: Randomized Approach

Proof (Expected Total Reward): $\theta^* \geq \theta^{\text{DP}}$

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$$= \sum_{t \in \mathcal{T}} \sum_{j \in \mathcal{J}} \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} r_{jt}(s, a) \cdot \pi_{tj}^{\text{LP}}(s, a) = \theta^{\text{LP}} \geq \theta^{\text{DP}}$$

fluid LP is relaxation
of fluid DP



Recovery of Individual DP Policies: Randomized Approach

Proof (Resource Violation): $\sum_{j \in \mathcal{J}} \sum_{i=1}^{n_j} c_{tlj}(\tilde{s}_{t,(j,i)}, \tilde{a}_{t,(j,i)}) \leq b_{tl} + \epsilon \cdot \sum_{j \in \mathcal{J}} n_j \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} c_{tlj}(s, a)$

$$\sum_{j \in \mathcal{J}} \sum_{i=1}^{n_j} c_{tlj}(\tilde{s}_{t,(j,i)}, \tilde{a}_{t,(j,i)})$$

Recovery of Individual DP Policies: Randomized Approach

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$$\sum_{j \in \mathcal{J}} \sum_{i=1}^{n_j} c_{tlj}(\tilde{s}_{t,(j,i)}, \tilde{a}_{t,(j,i)})$$

$$= \sum_{j \in \mathcal{J}} \sum_{i=1}^{n_j} \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} c_{tlj}(s, a) \cdot \mathbf{1} [\tilde{s}_{t,(j,i)} = s \wedge \tilde{a}_{t,(j,i)} = a]$$



summation over all
possible state-action pairs

Recovery of Individual DP Policies: Randomized Approach

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$$= \sum_{j \in \mathcal{J}} \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} c_{tlj}(s, a) \cdot \sum_{i=1}^{n_j} \mathbf{1} [\tilde{s}_{t,(j,i)} = s \wedge \tilde{a}_{t,(j,i)} = a]$$



move sum
inside

Recovery of Individual DP Policies: Randomized Approach

Proof (Resource Violation): $\sum_{j \in \mathcal{J}} \sum_{i=1}^{n_j} c_{tlj}(\tilde{s}_{t,(j,i)}, \tilde{a}_{t,(j,i)}) \leq b_{tl} + \epsilon \cdot \sum_{j \in \mathcal{J}} n_j \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} c_{tlj}(s, a)$

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$$= \sum_{j \in \mathcal{J}} \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} c_{tlj}(s, a) \cdot \sum_{i=1}^{n_j} \mathbf{1} [\tilde{s}_{t,(j,i)} = s \wedge \tilde{a}_{t,(j,i)} = a]$$

$$\leq b_{tl} + \epsilon \sum_{j \in \mathcal{J}} n_j \cdot \sum_{s \in \mathcal{S}_j} \sum_{a \in \mathcal{A}_j} c_{tlj}(s, a)$$

Hoeffding's inequality
(fluid LP gives expected value)

