

# Barely Random Algorithms and Collective Metrical Task Systems

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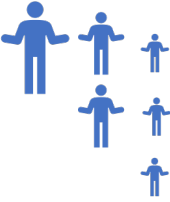
The logo for Inria, featuring the word "Inria" in a red, cursive script font.

**Keywords:** decision-making, online algorithm,  
randomness, collective algorithms

*5 mins*

# Warmup: Zero-sum Games

$M$  be a **cost** matrix in 2-player game



Cost	Rock	Paper	Scis.
Rock	0	1	-1
Paper	-1	0	1
Scis.	1	-1	0

	« Pure »	« Barely Random » ( $k=2$ )	« Mixed »
Strategy $x =$	$(1, 0, 0)$	$(\frac{1}{2}, \frac{1}{2}, 0)$	$(\frac{1}{3}, \frac{1}{3}, \frac{1}{3})$
Cost	1	0.5	0

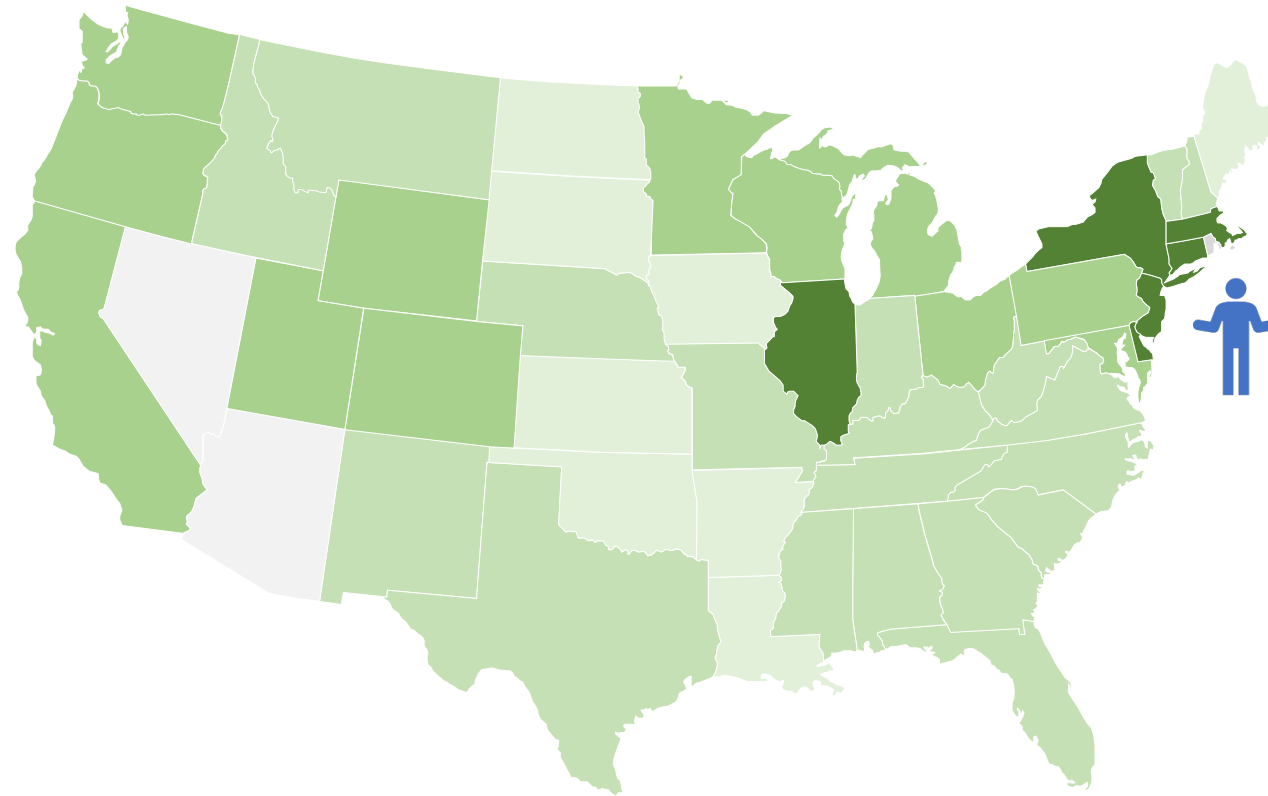
**Definition.** A  $k \in \mathbb{N}$  –barely random strategy  $x \in \mathcal{P}_k$  is one which

- ( $k$ -barely fractional)  $\longleftrightarrow$  • Takes value in  $\{0, \frac{1}{k}, \frac{2}{k}, \dots, 1\}$
- ( $k$ -barely random)  $\longleftrightarrow$  • Can be sampled with a random seed  $S \sim \mathcal{U}(\{1, \dots, k\})$
- ( $k$ -collective)  $\longleftrightarrow$  • Represents the distribution of  $k$  players playing collectively

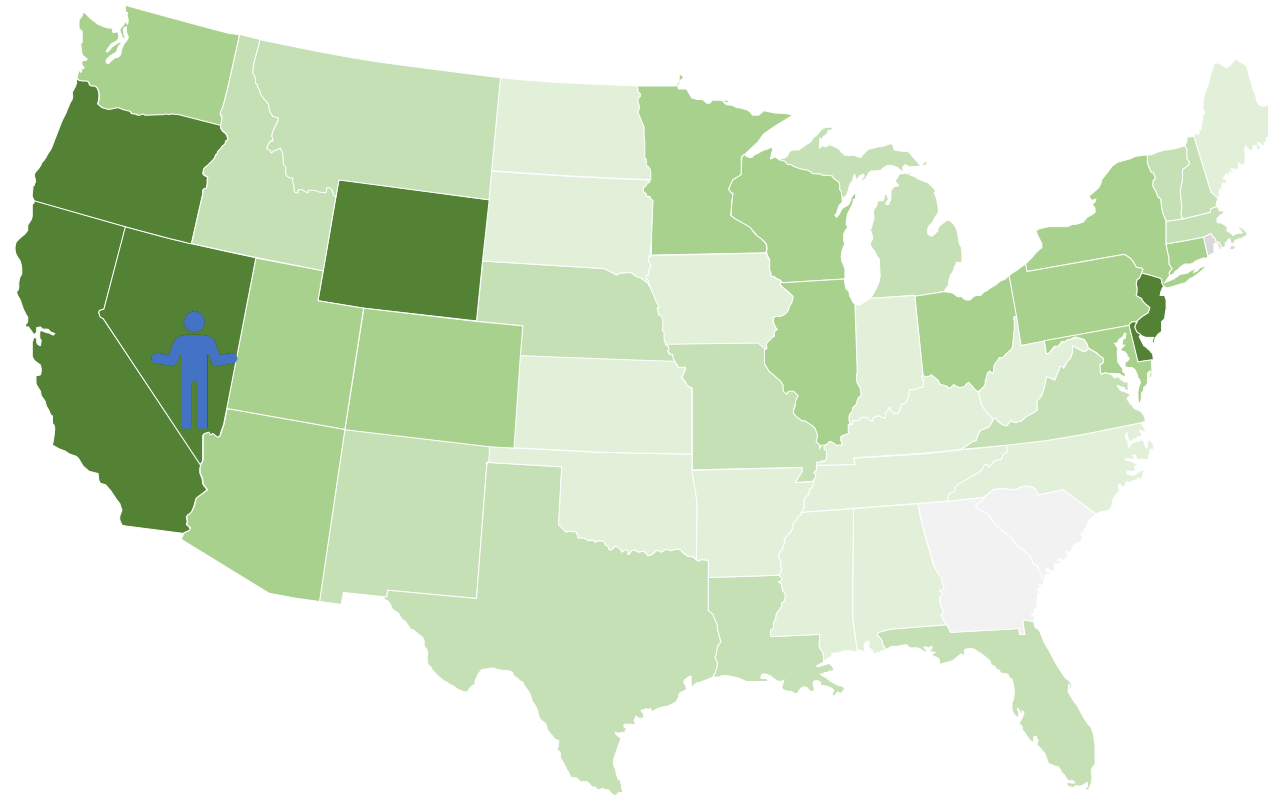
How does the payoff/cost scales with « bareyness »  $k \in \mathbb{N}$  ?

In the context of real-time games (online algorithms)

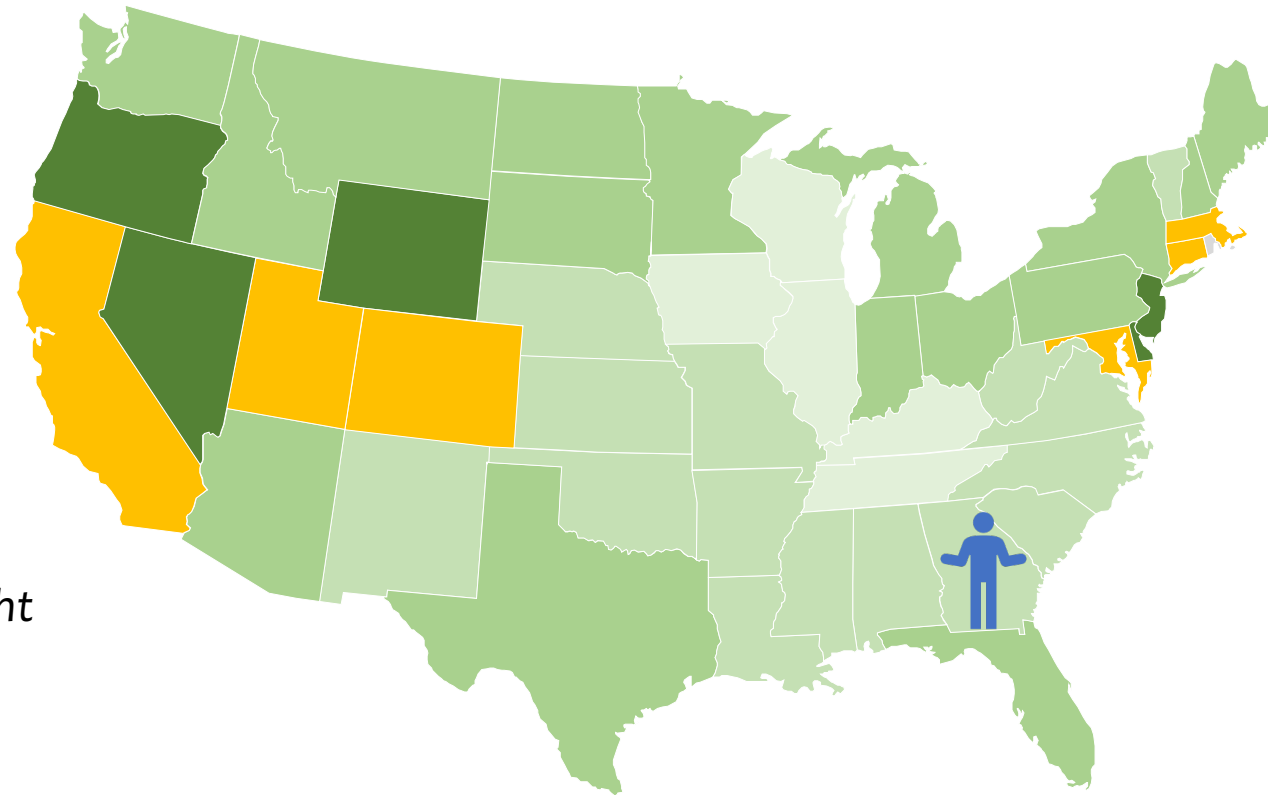
# Price of Rent in 1950



# Price of Rent in 1980



# Price of Rent in 2000



$$\text{Comp. Ratio} = \frac{\text{Cost}}{\text{OPT}}$$

= cost due to lack of hindsight

# Metrical Task Systems

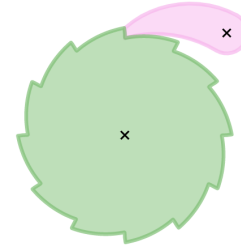
When should I move? (between jobs, cities, power saving modes ...)

- Metric space  $(\mathcal{X}, d)$ , with  $|\mathcal{X}| = n$  number of positions.
- Input  $\mathbf{c}(\cdot)$  : Cost vector  $\mathbf{c}(1), \mathbf{c}(2), \dots \in \mathbb{R}_+^{\mathcal{X}}$
- Output  $x(\cdot)$  : Agent's position  $x(1), x(2), \dots \in \mathcal{X}$  or  $k \in \mathbb{N}$  (barely random) positions  $\in \mathcal{P}_k(\mathcal{X})$ 
  - With past data  $\mathbf{c}(\leq t)$
- **Cost** = Cost vector + Movement, i.e.,  $\text{Cost}(x(\cdot), \mathbf{c}(\cdot)) = \sum_t c_{x(t)}(t) + d(x(t), x(t+1))$
- **OPT** = Best with hindsight, i.e.  $\text{OPT} = \inf_{x(\cdot)} \text{Cost}(x(\cdot), \mathbf{c}(\cdot))$

## Main result

	Deterministic : $k = 1$	« Barely Random » $k \in \mathbb{N}$	Randomized: $k = \infty$
<b>Comp. Ratio</b>	$2n - 1$ [BLS]	$\Theta(\log^2 n)$ if $k \geq n^2$ [This work]	$\Theta(\log^2 n)$ [Bubeck et al. 2019-22]

# Techniques:



physical device with hysteresis:  
ratchet (green)+ pawl (purple)

- **Idea:**

- **Transform a** fully fractional strategy  $\mathbf{y}(t) \in \mathcal{P}(X)$
- **In a**  $k$  – barely fractional strategy  $\mathbf{x}(t) \in \mathcal{P}_k(X)$

- **Naïve transform :**

$$\mathbf{x}(t) = \arg \min_{\mathbf{x} \in \mathcal{P}_k(X)} \text{OT}(\mathbf{x}, \mathbf{y}(t))$$

- **Good transform :** (*Hysteresis!*) or (« *L1 Wasserstein Prox* »)

$$\mathbf{x}(t) = \arg \min_{\mathbf{x} \in \mathcal{P}_k(X)} \text{OT}(\mathbf{x}, \mathbf{y}(t)) + \text{OT}(\mathbf{x}(t-1), \mathbf{x}).$$

The issue (e.g. for  $k = 2$ )

$\mathbf{y}(t)$	$\mathbf{x}(t)$
(0.76, 0.24)	(1, 0)
(0.74, 0.25)	(0.5, 0.5)

# Thank you!

And think of Hysteresis to build Robust systems in face of Uncertainty

