

# Spectral learning for structured partially observable environments

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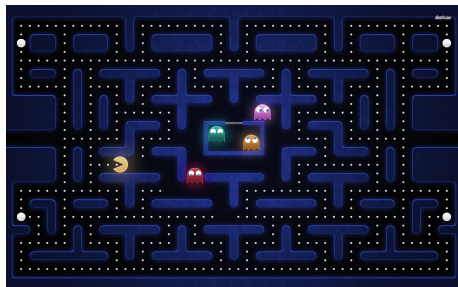
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# Structured Partially Observable Environments

## Structured Environments

- ① Goal: Predictions
- ② Plan: Exploit structure
- ③ Example: Pacman



# PSR: The Timing Case

- Model environment with a Predictive state representation
- For timing we have one observation symbol:  $\sigma$

Notation:  $\sigma$ : one time unit,  $\sigma^k$ : k time units

- PSR defined by:  $\langle \alpha_0, \{A_\sigma\}, \alpha_\infty \rangle$

$\alpha_0$ : Initial weighting on states  $1 \times n$

$A_\sigma$ : Transition matrix  $n \times n$

$\alpha_\infty$ : Normalizer  $n \times 1$

- PSRs compute probabilities of observations

$$f(\sigma^k) = \alpha_0 * A_\sigma^k * \alpha_\infty$$

- Example of a PSR: HMM

# Spectral Learning of PSRs

Step 1: Represent data as a matrix

Step 2: Singular value decomposition

Step 3: Pick number of states for PSR

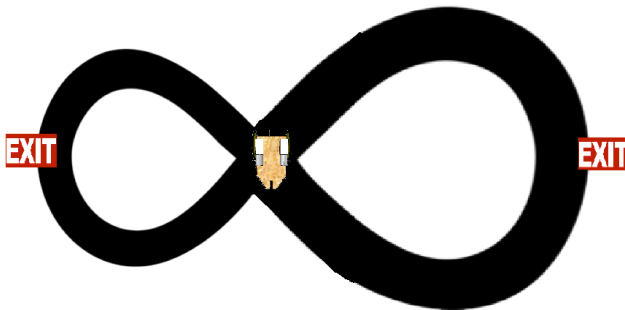
Step 4: Learn  $\langle \alpha_0, \{A_\sigma\}, \alpha_\infty \rangle$  with matrix computations

# The Base System

- Idea: Add  $\{A_\sigma, A_{\sigma^2}, A_{\sigma^4}, A_{\sigma^8}, \dots A_{\sigma^N}\}$  as extra transition operators
- Timing queries:  $f(\sigma^{11}) = \alpha_0 * A_{\sigma^8} * A_{\sigma^2} * A_{\sigma^1} * \alpha_\infty$
- Motivation:
  - 1) Express transitions directly
  - 2) Faster queries

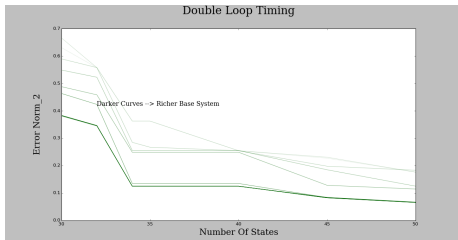
# Timing with the Base

Agent goes through loops until leaving through an exit state. Loop lengths are 64 and 16 time units (not to scale).

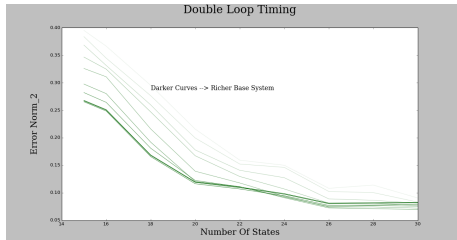


# Base System Performance for Loops

No noise in durations



Noise in durations

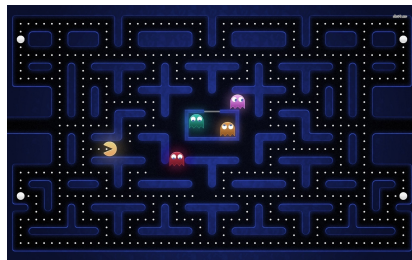
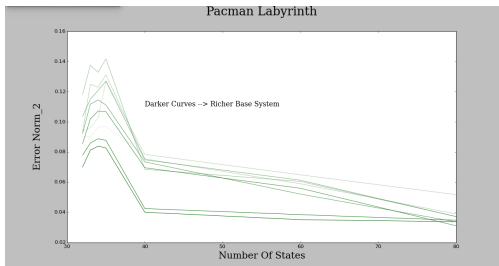


- Noise allows for smaller models

- $$\|f - \hat{f}\| = \sqrt{\sum_{x \in \text{observations}} (f(x) - \hat{f}(x))^2}$$

# Pacman Labyrinth

## Timing Predictions

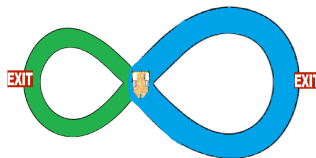
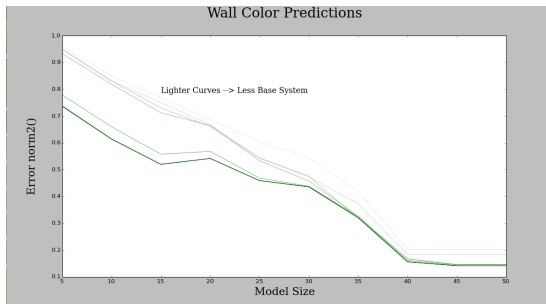


- $$\|f - \hat{f}\| = \sqrt{\sum_{x \in \text{observations}} (f(x) - \hat{f}(x))^2}$$



# Wall Color Predictions

We paint the first loop green and the second loop blue



- $$\|f - \hat{f}\| = \sqrt{\sum_{x \in \text{observations}} (f(x) - \hat{f}(x))^2}$$

# Picking the Base System

- Picking operators to exploit structure

Observations:  $\{ "a^{30}":10, "a^{60}":5, "b^{18}":15 \}$

Desired Base System:  $A_{a^{30}}, A_{b^{18}}, A_a, A_b$

- Substring properties: **long, frequent, diverse**
- Solution: iterative greedy heuristic

# Computing with the Base System

- Using the Base System well involves requires good **string partitions**

Query string: "abcacb", Base System =  $\{A_{ab}, A_{bca}, A_{cb}, A_a, A_b\}$

Desired partition: "a—bca—cb""

- Goal: minimize matrices used
- Solution: dynamic programming

# Conclusion and Future Work

- What's left for the Base System?
  - 1) Theoretical analysis
  - 2) Test heuristics on labyrinths
  - 3) Further optimize heuristics

# Questions? Comments?