

POMDP Value Iteration	$ S  \sim 10-20$
SARSOP	$ S  \sim 10,000$

# POMDP Formulation Approximations

# POMDP Computational Complexity

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Sad facts ● ☹️

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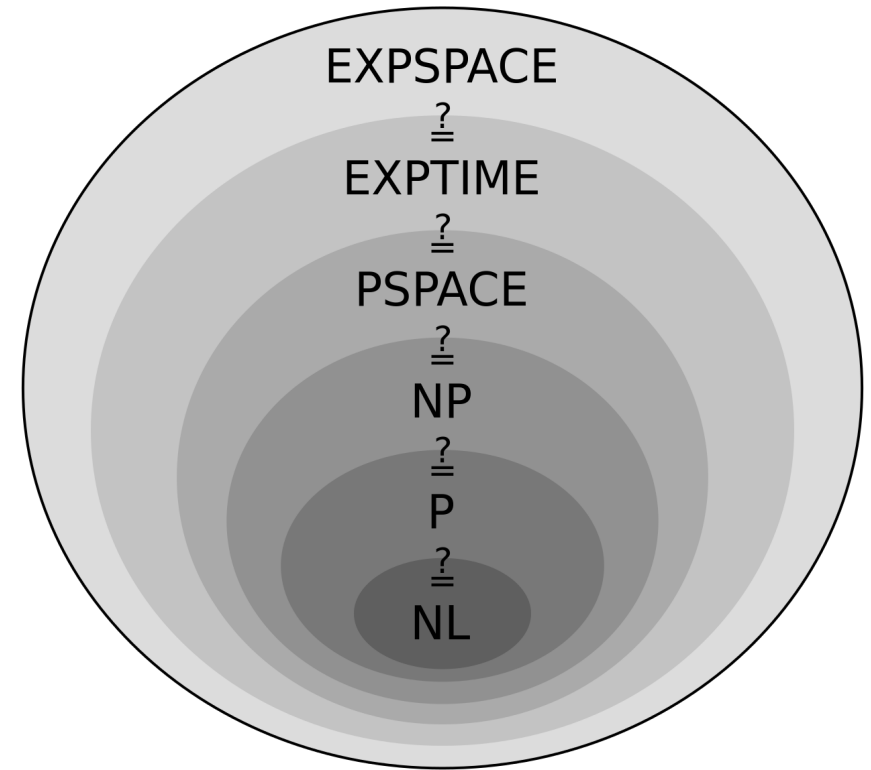
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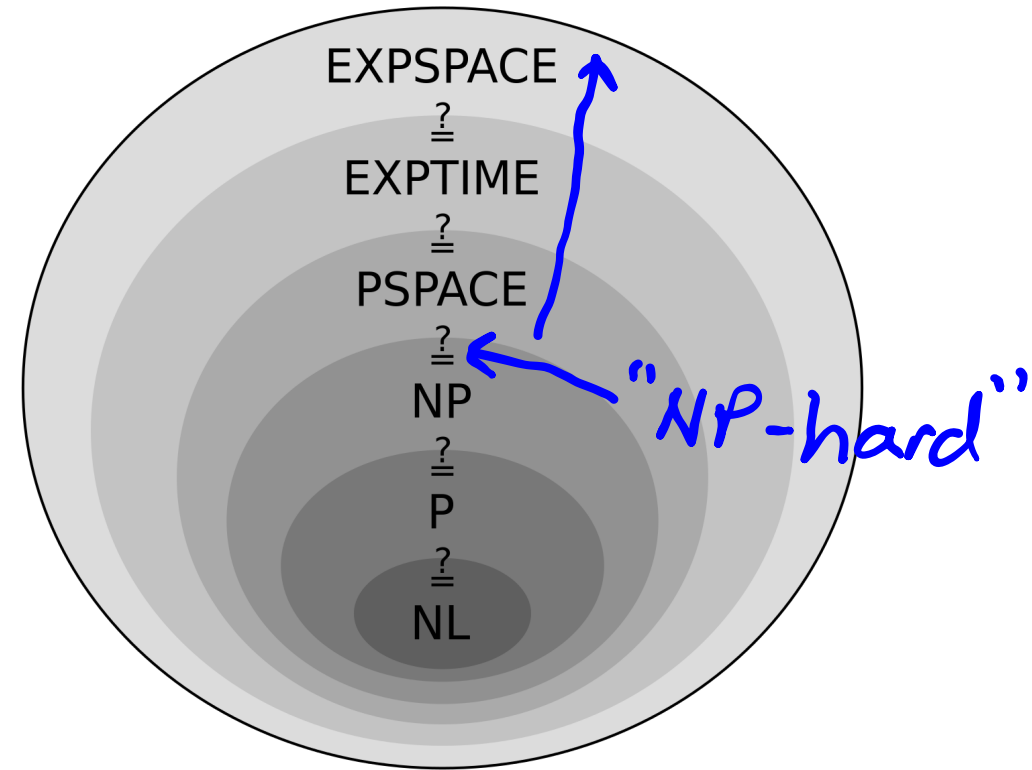
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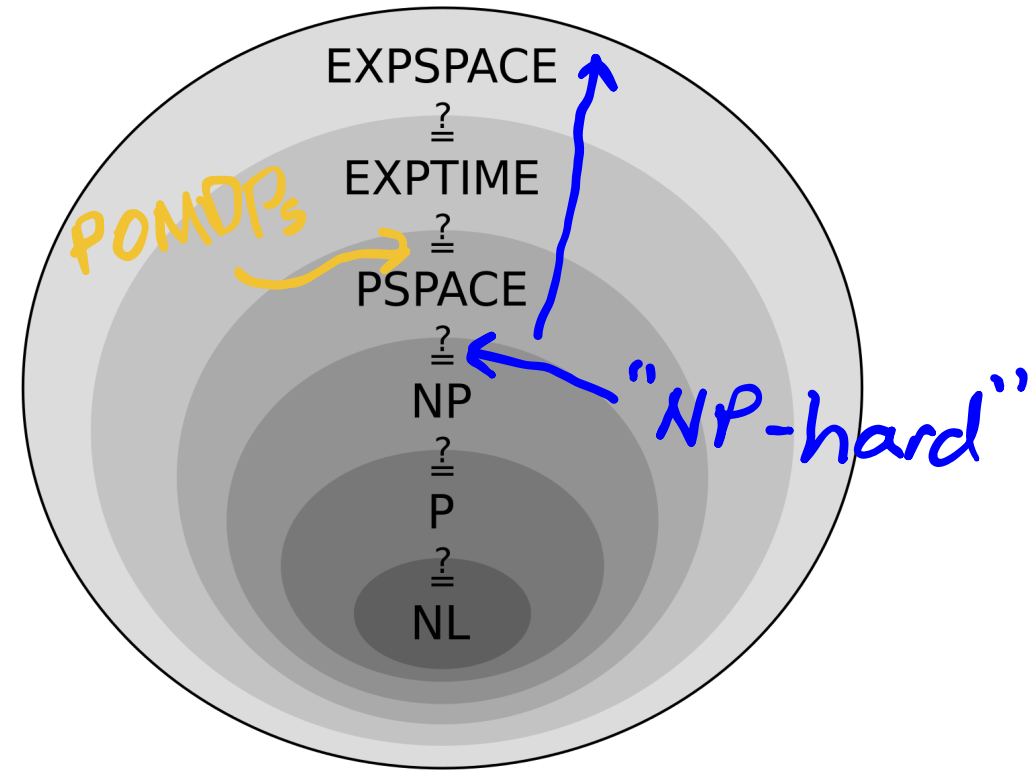
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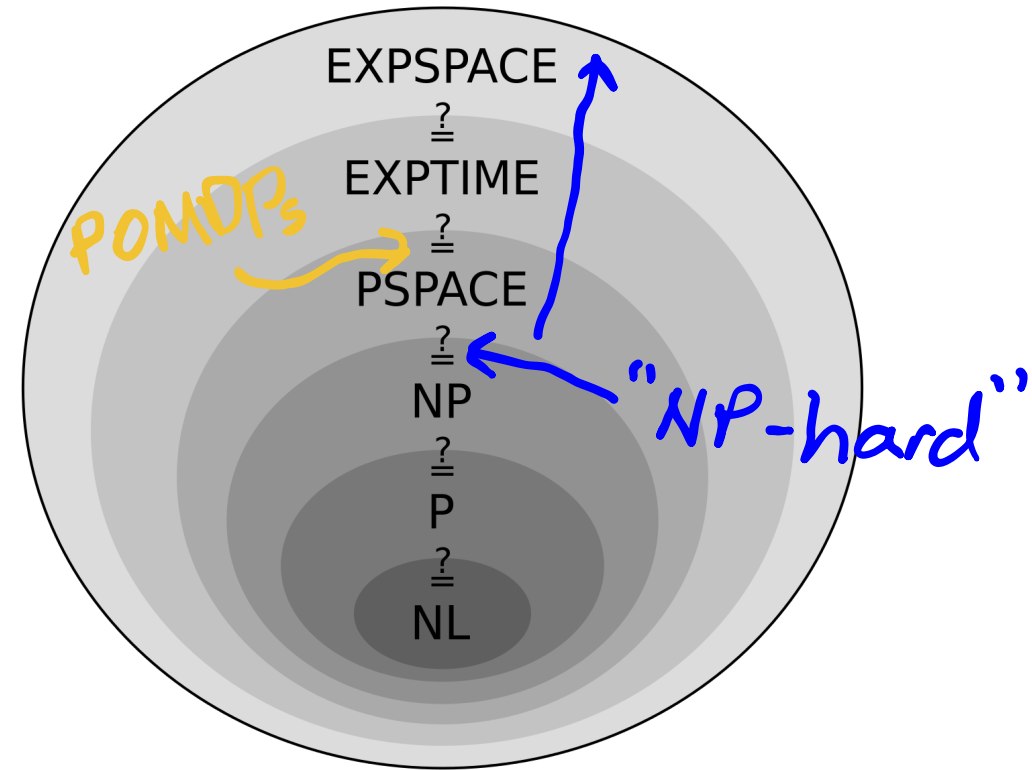




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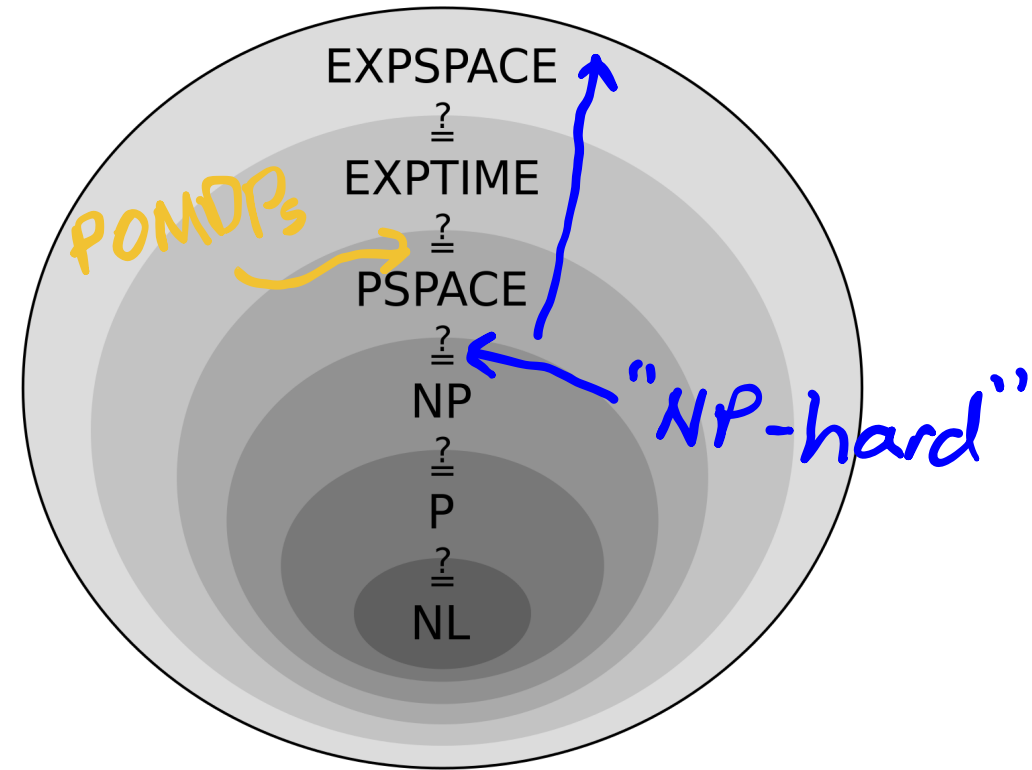
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# Approximate POMDP Solutions

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## Numerical Approximations

(approximately solve original problem)

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**Offline**

# Approximate POMDP Solutions

$$\bar{V} - \underline{V} = \varepsilon$$

## Numerical Approximations

(approximately solve original problem)



Offline

~~Last week~~

Today

scalable to  $|S| \sim 10,000$

# Approximate POMDP Solutions

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**Offline**

Last week



**Online**

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## Formulation Approximations

(solve a slightly different problem)

# Approximate POMDP Solutions

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## Formulation Approximations

(solve a slightly different problem)

Today!

# POMDP Objective

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$$\pi^* = \operatorname{argmax}_{\pi: B \rightarrow A} \mathbb{E} \left[ \sum_{t=0}^{\infty} \gamma^t R(s_t, \pi(b_t)) \right]$$

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$$\underline{\underline{b' = \tau(b, a, o)}}$$

# Certainty Equivalent

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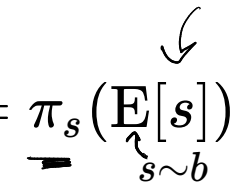
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$$\pi_{\text{CE}}(b) = \pi_{\underline{s}}(\mathbb{E}[s])$$



$$b' = \tau(b, a, o)$$



# Certainty Equivalent

MDP LQR

Optimal for LQG

LQG POMDP

$$T(\mathbf{s}' | \mathbf{s}, \mathbf{a}) = \mathcal{N}(\mathbf{s}' | \mathbf{T}_s \mathbf{s} + \mathbf{T}_a \mathbf{a}, \Sigma_s)$$

Linear Dynamics

$$O(\mathbf{o} | \mathbf{s}') = \mathcal{N}(\mathbf{o} | \mathbf{O}_s \mathbf{s}', \Sigma_o)$$

Gaussian Process Noise

Gaussian Observation Noise

$$b(\mathbf{s}) = \mathcal{N}(\mathbf{s} | \overset{\text{mean}}{\mu_b}, \overset{\text{covariance}}{\Sigma_b})$$

$$\mu_p \leftarrow \mathbf{T}_s \mu_b + \mathbf{T}_a \mathbf{a}$$

$$\Sigma_p \leftarrow \mathbf{T}_s \Sigma_b \mathbf{T}_s^\top + \Sigma_s$$

$$\mathbf{K} \leftarrow \Sigma_p \mathbf{O}_s^\top (\mathbf{O}_s \Sigma_p \mathbf{O}_s^\top + \Sigma_o)^{-1}$$

$$\mu_b \leftarrow \mu_p + \mathbf{K} (\mathbf{o} - \mathbf{O}_s \mu_p)$$

$$\Sigma_b \leftarrow (\mathbf{I} - \mathbf{K} \mathbf{O}_s) \Sigma_p$$

Bayesian Update  
Kalman Filter

$$\pi^*(b) = -\mathbf{K}_{LQR} \mu_b$$

# QMDP

## POMDP Objective

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$$\pi_{\text{QMDP}}(b) = \operatorname{argmax}_{a \in A} \mathbb{E}_{s \sim b} [\underbrace{Q_{\text{MDP}}(s, a)}]$$

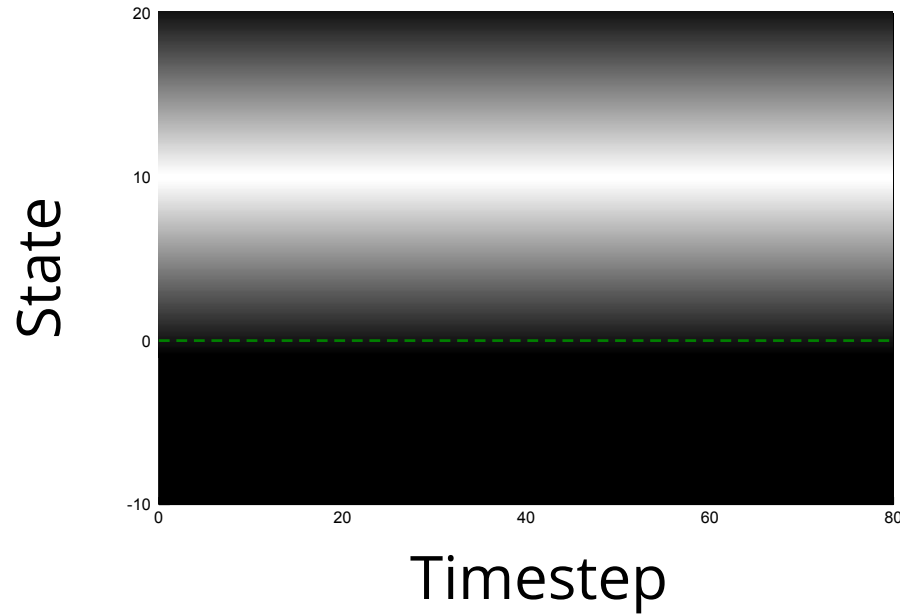
solve underlying  
MDP  
to get QMDP



$$b' = \tau(b, a, o)$$

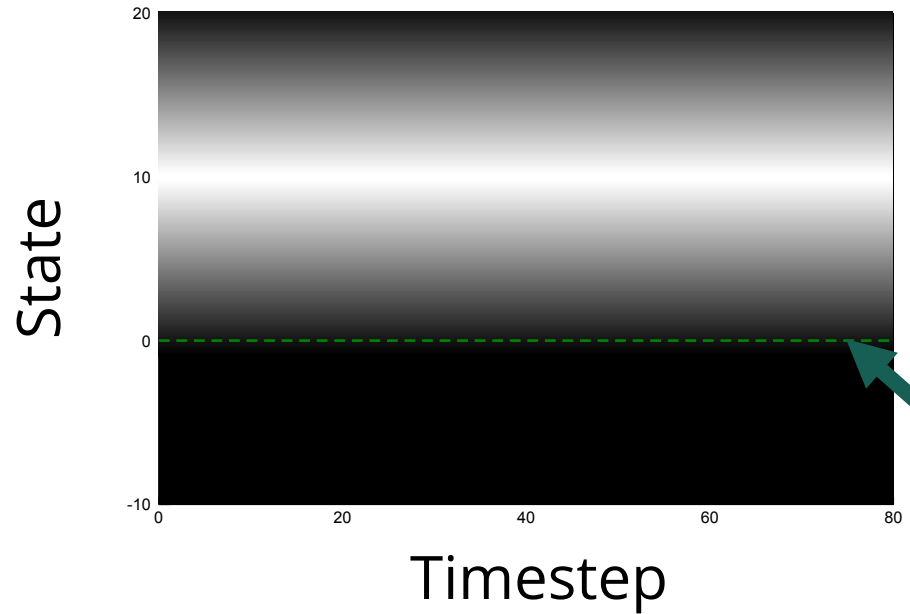
# Example: Tiger POMDP with Waiting

# POMDP Example: Light-Dark



$$\begin{aligned}\mathcal{S} &= \mathbb{Z} & \mathcal{O} &= \mathbb{R} \\ s' &= s + a & o &\sim \mathcal{N}(s, s - 10) \\ \mathcal{A} &= \{-10, -1, 0, 1, 10\} \\ R(s, a) &= \begin{cases} 100 & \text{if } a = 0, s = 0 \\ -100 & \text{if } a = 0, s \neq 0 \\ -1 & \text{otherwise} \end{cases}\end{aligned}$$

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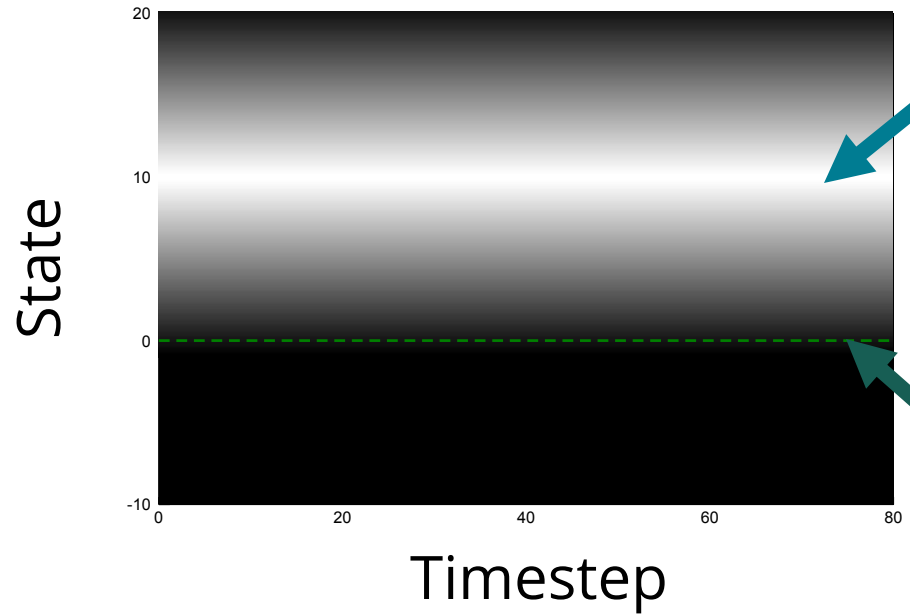


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Accurate Observations



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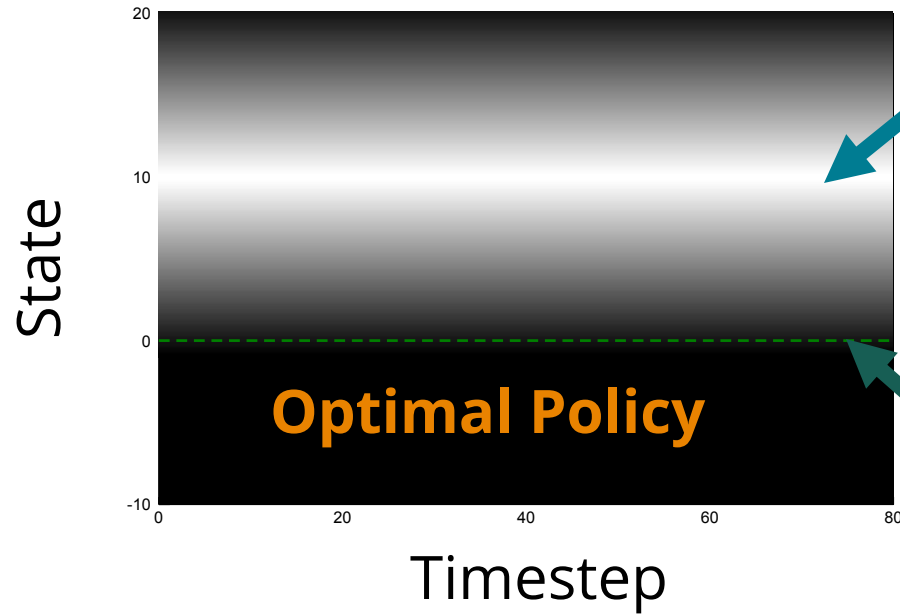
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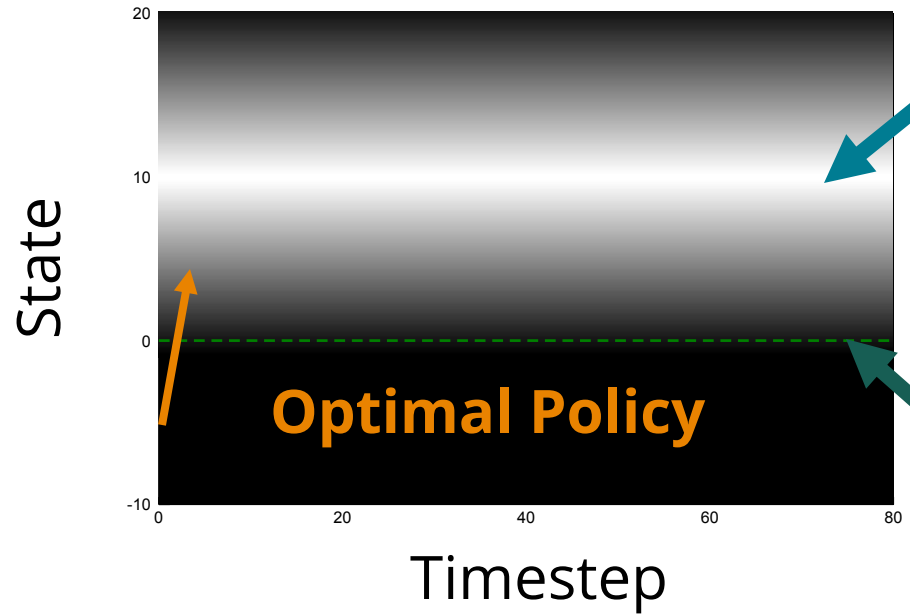
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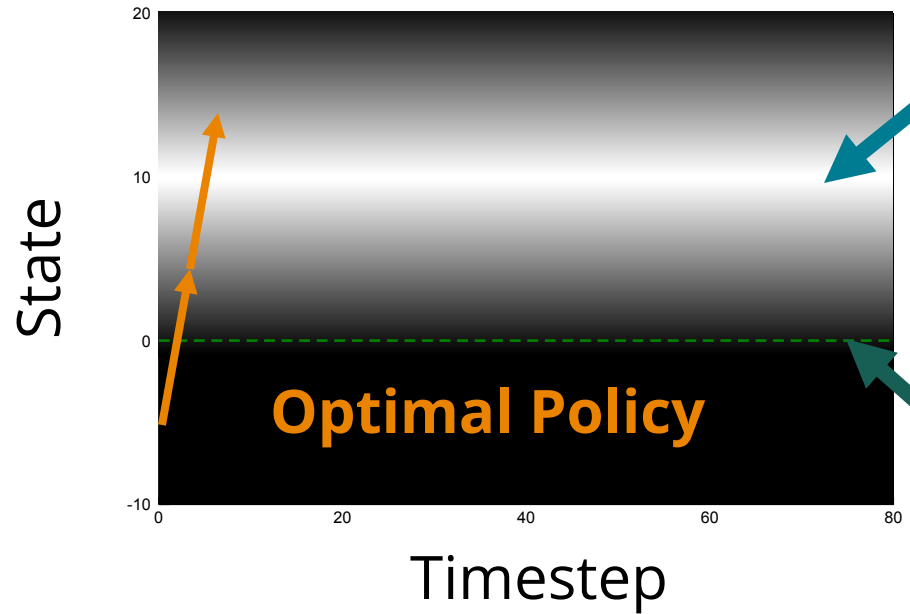


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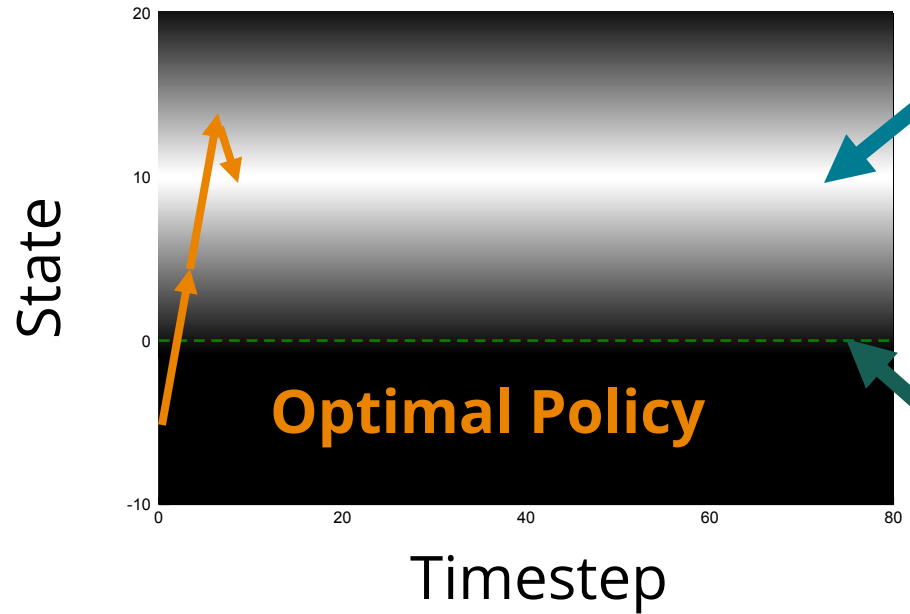


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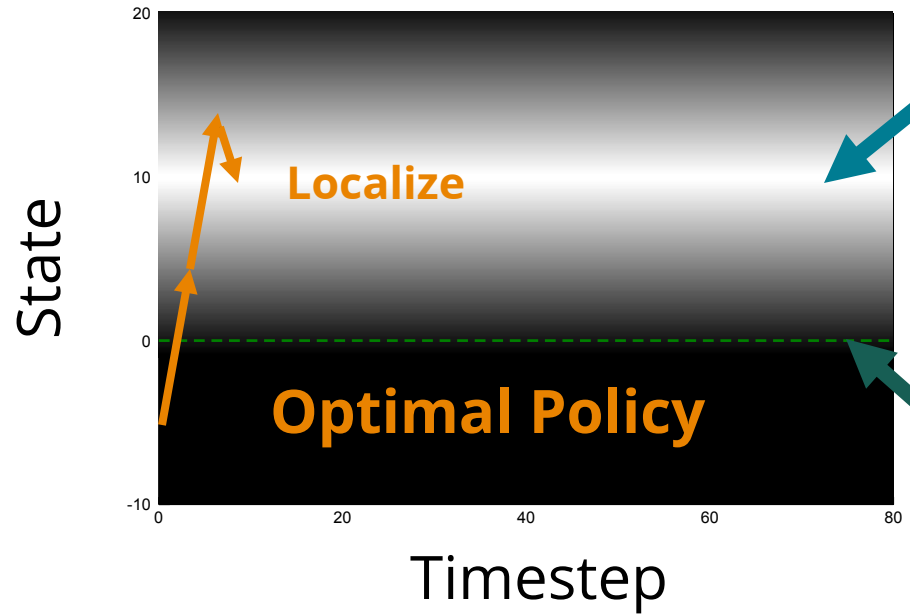


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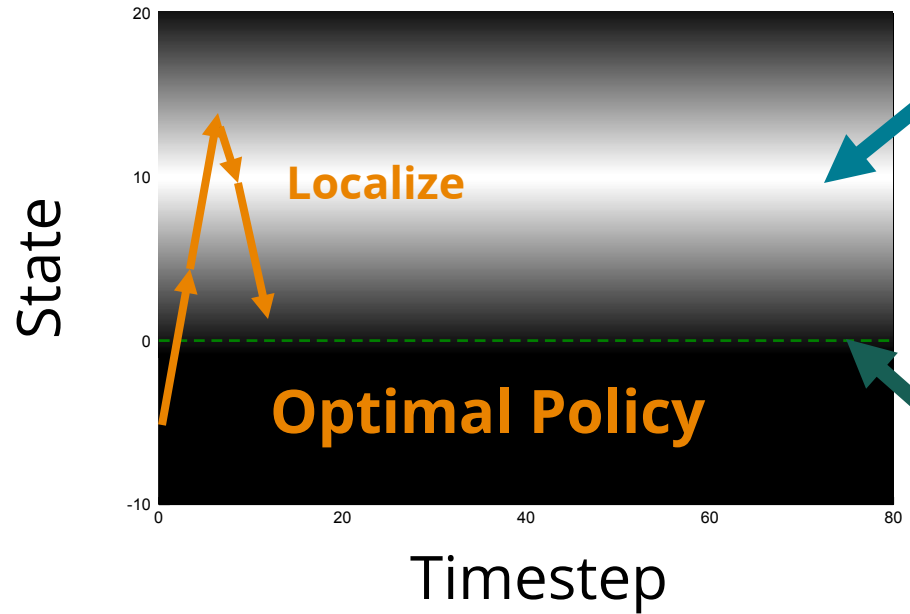
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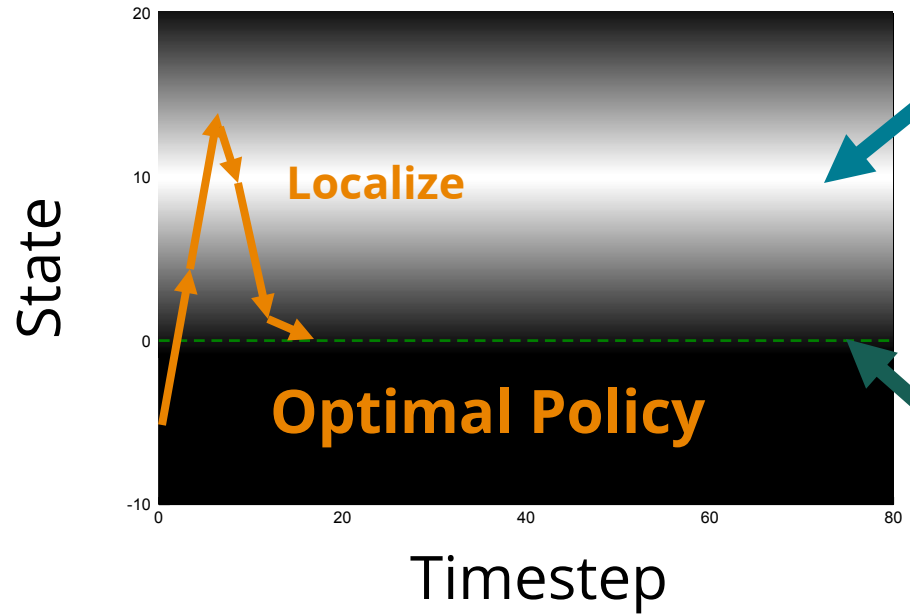
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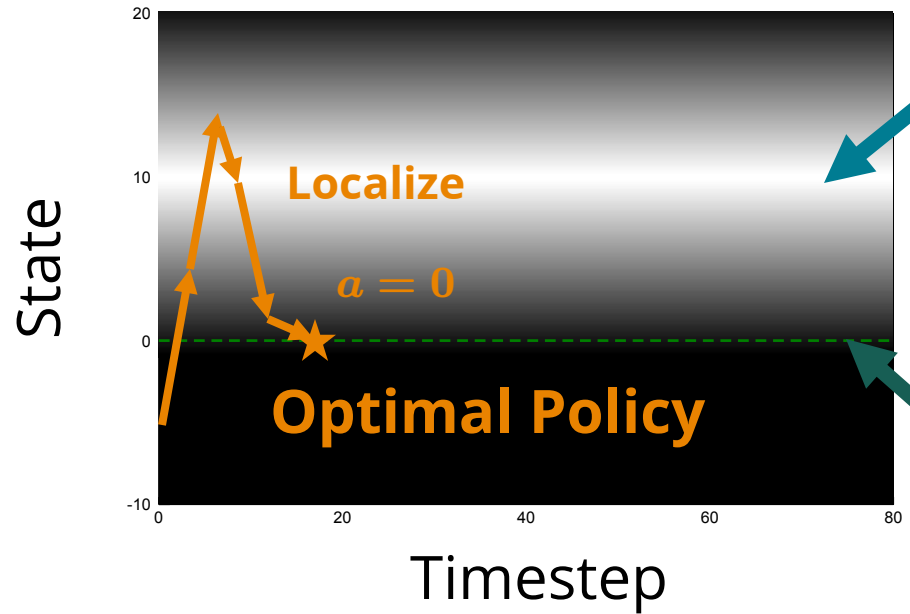
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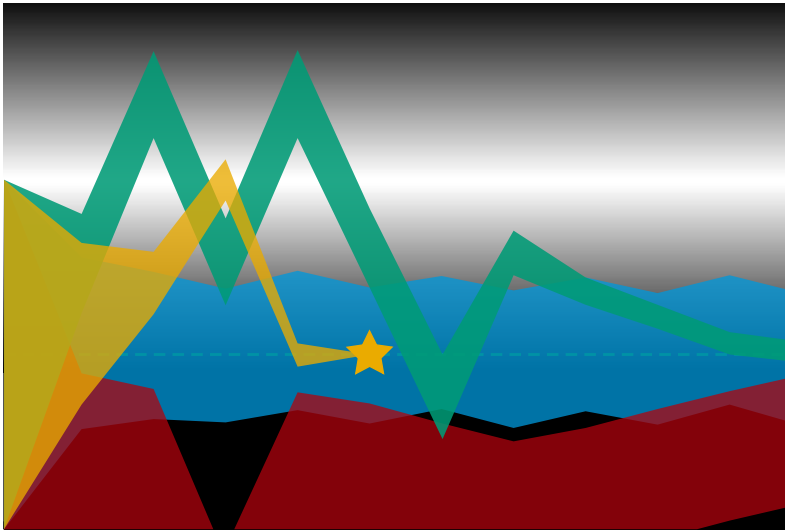
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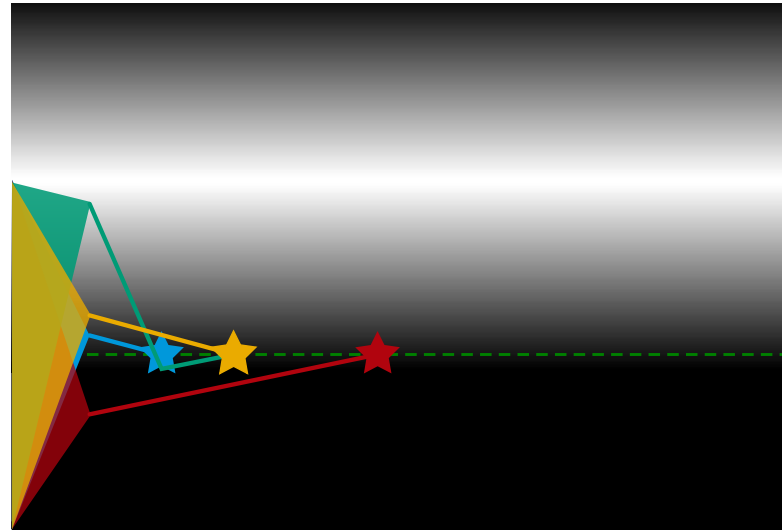
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## POMDP Solution



## QMDP

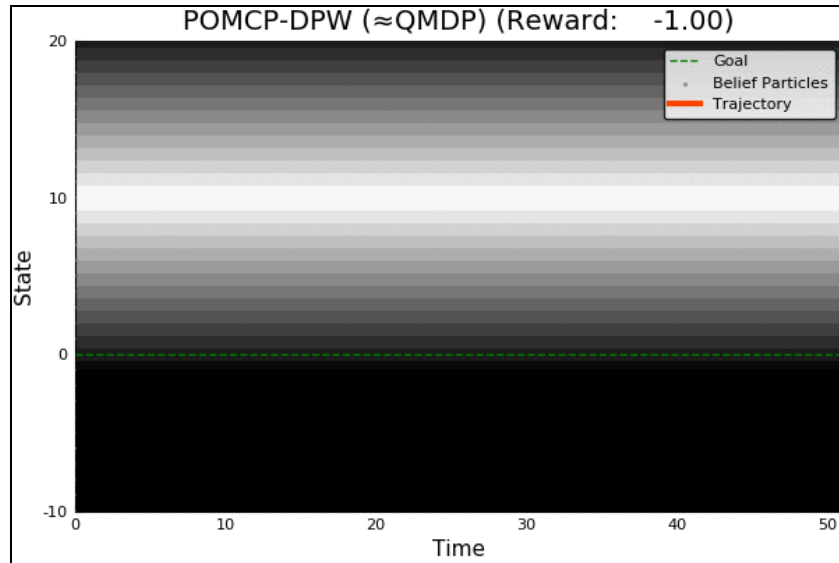


Same as **full observability**  
on the next step

# Information Gathering

QMDP

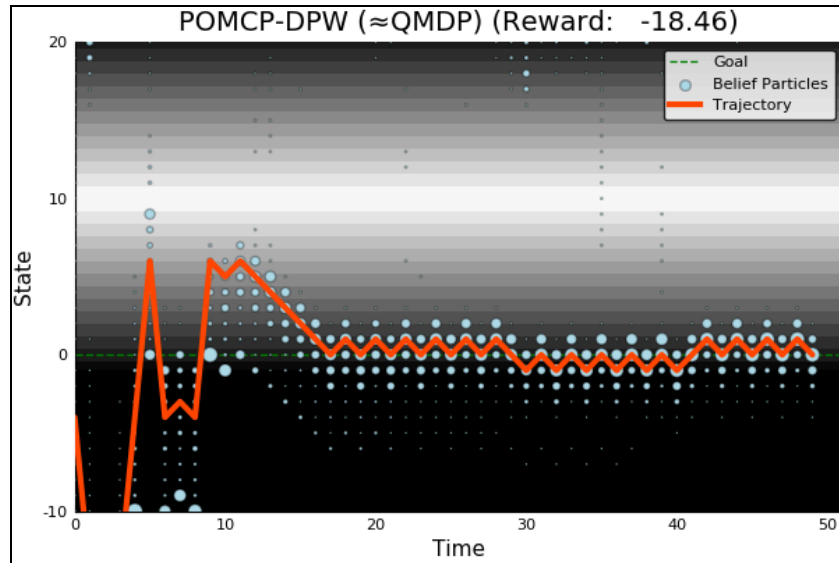
Full POMDP



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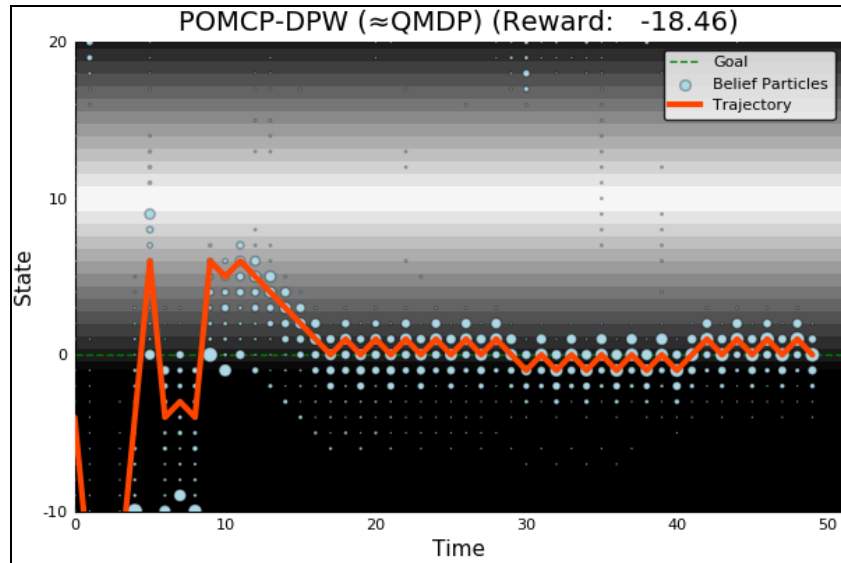
QMDP

Full POMDP

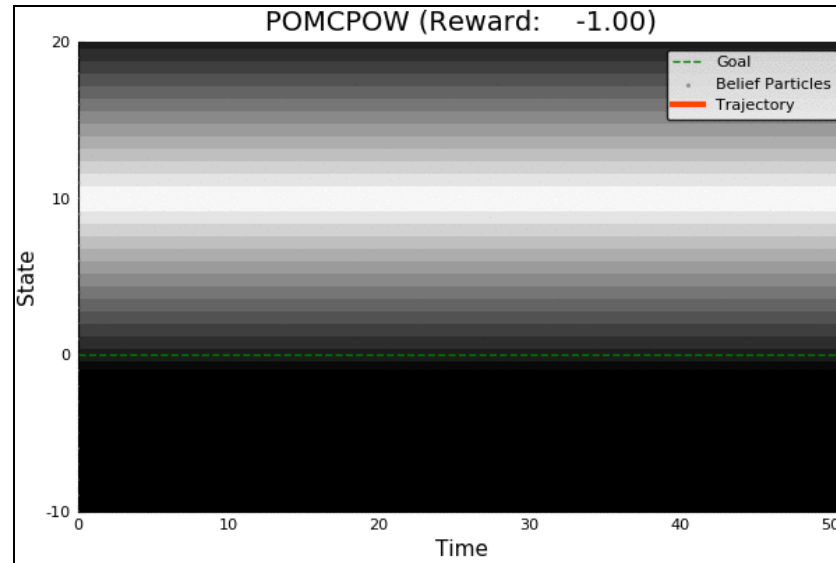


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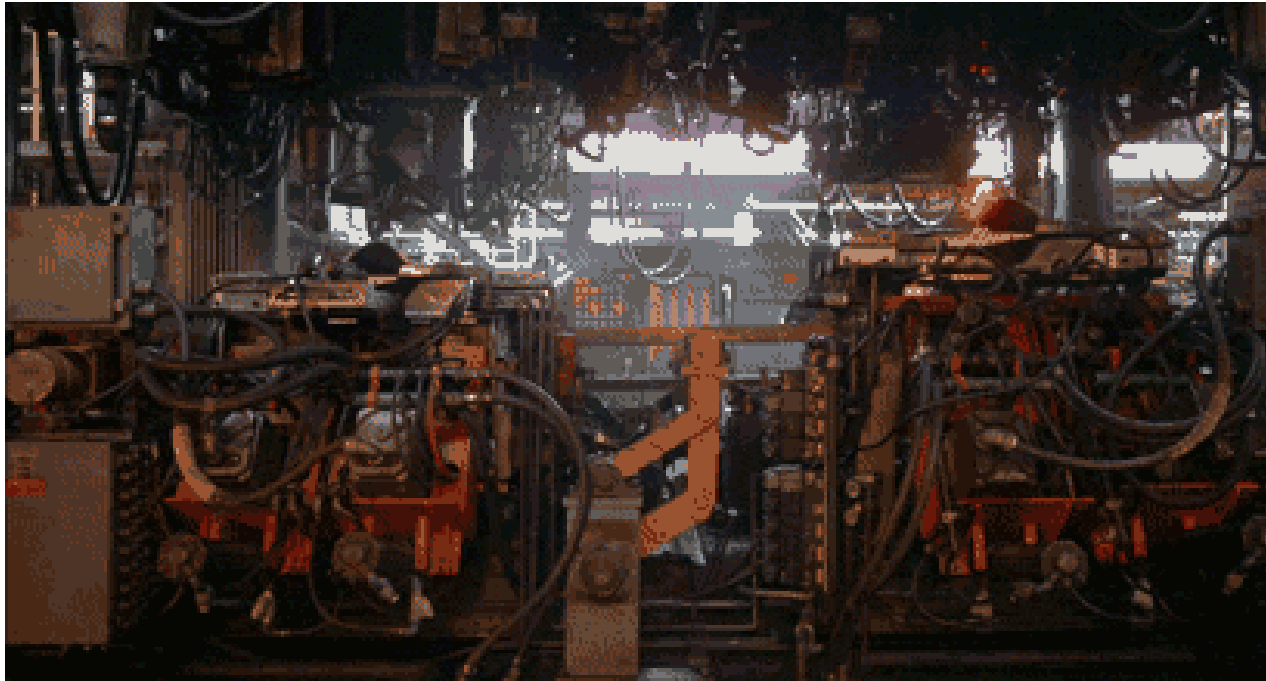


Full POMDP



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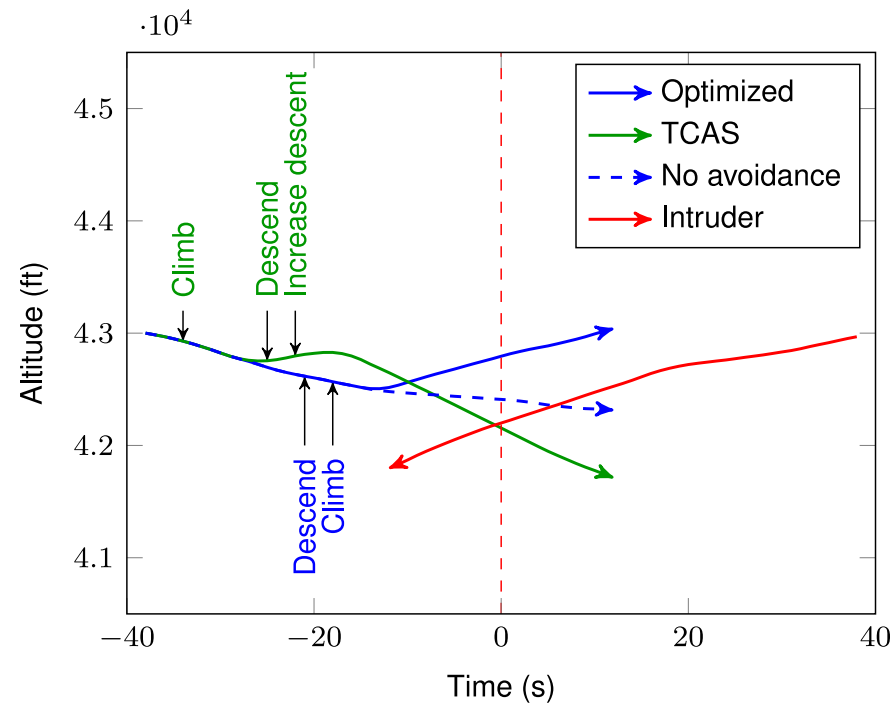
***INDUSTRIAL GRADE***



# QMDP

## ACAS X

[Kochenderfer, 2011]



# Hindsight Optimization

## POMDP Objective

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# FIB

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# k-Markov

## POMDP Objective

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# Open Loop

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