# Dyna Style Planning with Linear Function Approximation



Alborz Geramifard April, 2010



# Acknowledgments



## Richard Sutton

#### Csaba Szepesvari





#### Michael Bowling

Cosmin Paduraru



## Outline



## Background

## Linear Prioritized Sweeping

## Empirical Results

#### Discussion



#### Reinforcement Learning



[Sutton, Barto 1998]



#### Planning



#### [Sutton, Barto 1998]



#### Function Approximation



#### [Sutton, Barto 1998]



Why planning ?

Expensive data

Trade off between data and time

Tracking ...















[Moore, Atkeson 1993]





[Moore, Atkeson 1993]





[Moore, Atkeson 1993]







## Background

#### Linear Prioritized Sweeping

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Discussion



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#### Similar to Prioritized Sweeping





Policy Independent Model





#### Similar to Prioritized Sweeping





Policy Independent Model













#### PQueue ¢i ∼∆





















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## Transition Model: $F_a$

 $F_a\phi = \phi'$ 

## Reward Model: $b_a$

 $b_a^T \phi = r$ 

Obtain initial  $\phi, \theta, F, b$ 

For each time step:

Take action a according to the policy. Receive  $r,\phi'$ 



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$$\begin{split} \delta &\leftarrow r + \gamma \theta^{\top} \phi' - \theta^{\top} \phi \\ \theta &\leftarrow \theta + \alpha \delta \phi \end{split}$$



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$$\delta \leftarrow r + \gamma \theta^{\top} \phi' - \theta^{\top} \phi$$
$$\theta \leftarrow \theta + \alpha \delta \phi$$
$$F \leftarrow F + \alpha (\phi' - F \phi) \phi^{\top}$$
$$b \leftarrow b + \alpha (r - b^{\top} \phi) \phi$$



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Algorithm 2 : Linear Dyna with PWMA prioritized sweeping (policy evaluation) Obtain initial  $\phi, \theta, F, b$ For each time step: Take action a according to the policy. Receive  $r, \phi'$ value/policy  $\delta \leftarrow r + \gamma \theta^\top \phi' - \theta^\top \phi$  $\theta \leftarrow \theta + \alpha \delta \phi$ acting planning direct  $F \leftarrow F + \alpha(\phi' - F\phi)\phi^{\top}$  $b \leftarrow b + \alpha (r - b^{\top} \phi) \phi$ model experience For all *i* such that  $\phi(i) \neq 0$ : For all j such that  $F^{ij} \neq 0$ : mode learning Put j on the PQueue with priority  $|F^{ij}\delta\phi(i)|$ Repeat p times while PQueue is not empty:  $i \leftarrow \text{pop the PQueue}$  $\delta \leftarrow b(i) + \gamma \theta^{+} F e_i - \theta(i)$  $\theta(i) \leftarrow \theta(i) + \alpha \delta$ For all j such that  $F^{ij} \neq 0$ : Put j on the queue with priority  $|F^{ij}\delta|$  $\phi \leftarrow \phi'$ 

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PQueue φ*i* Δi







![](_page_42_Picture_0.jpeg)

![](_page_43_Picture_0.jpeg)

## Background

#### Linear Prioritized Sweeping

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Discussion

![](_page_44_Picture_0.jpeg)

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![](_page_45_Picture_0.jpeg)

#### Settings

# 30 runs, same set of trajectories Best decay parameters in the set Results are shifted a bit

![](_page_46_Picture_0.jpeg)

#### N = 98 states

![](_page_46_Figure_2.jpeg)

## Boyan Chain (No control)

![](_page_47_Figure_1.jpeg)

![](_page_48_Picture_0.jpeg)

#### Tile coding (10000 tiles, 10 tillings)

![](_page_48_Figure_2.jpeg)

## Mountain Car (PE)

![](_page_49_Figure_1.jpeg)

## Mountain Car (Control)

![](_page_50_Figure_1.jpeg)

![](_page_51_Picture_0.jpeg)

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![](_page_52_Picture_0.jpeg)

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![](_page_53_Picture_0.jpeg)

The whole story is nice, but does it really work? Theory?

Geared towards sparse features

![](_page_53_Picture_3.jpeg)

![](_page_53_Picture_4.jpeg)

![](_page_53_Picture_5.jpeg)

![](_page_54_Picture_0.jpeg)

#### More planning on larger problems

#### Using Sigmoid function

#### Correlation with SPPI and LSPI

#### Convergence Proof

# Questions?