# Imitation Learning (Behavior Cloning)

Fabian Caba Heilbron IVUL – February 16

http://rll.berkeley.edu/deeprlcourse/docs/week 2 lecture 1 behavior cloning.pdf



#### Imitation vs Reinforcement



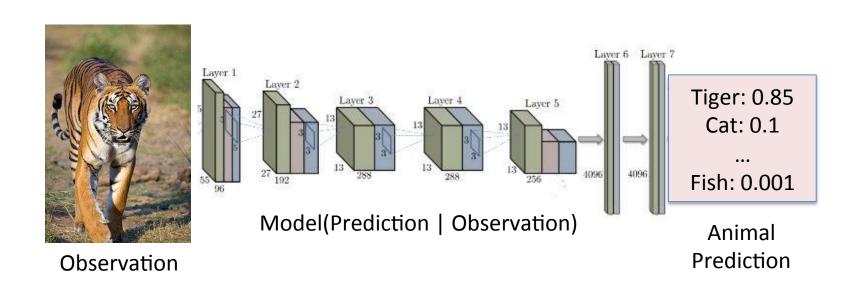


Define a reward for each task!

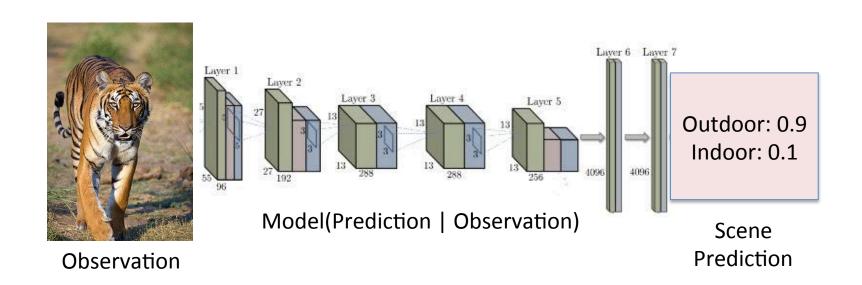
# Today's talk

- Definition of sequential problem
- Imitation learning: supervised learning for decision making
  - Does imitation learning works?
  - How can we make it work more often?
- Case studies of recent work in (deep) imitation learning

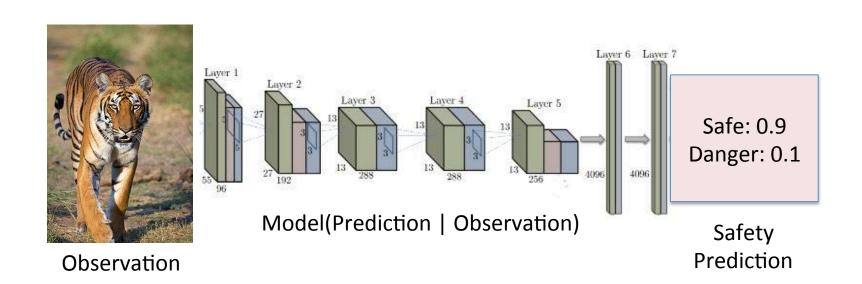
# Supervised Learning

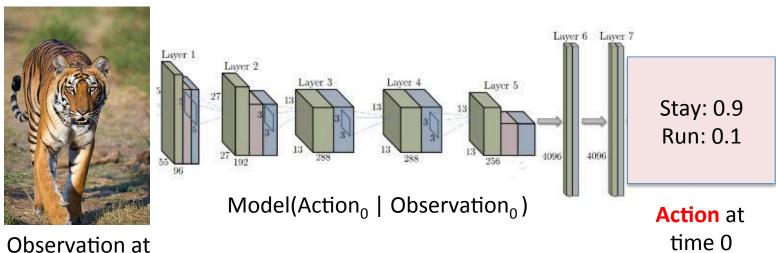


# Supervised Learning

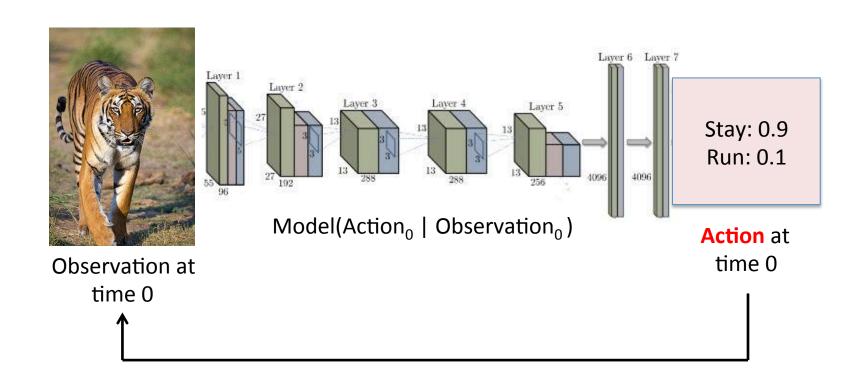


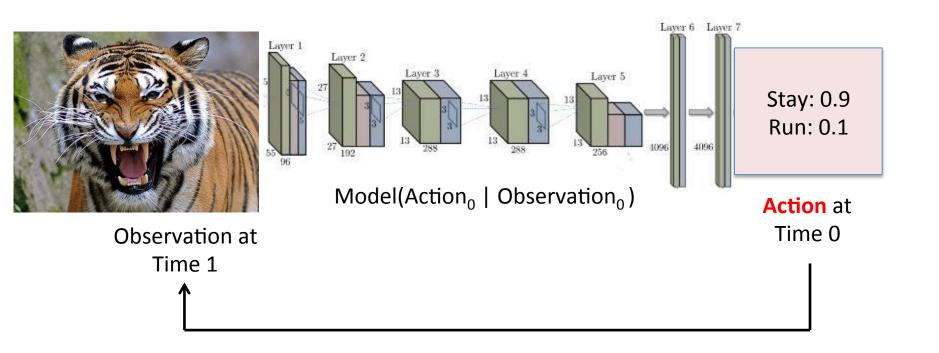
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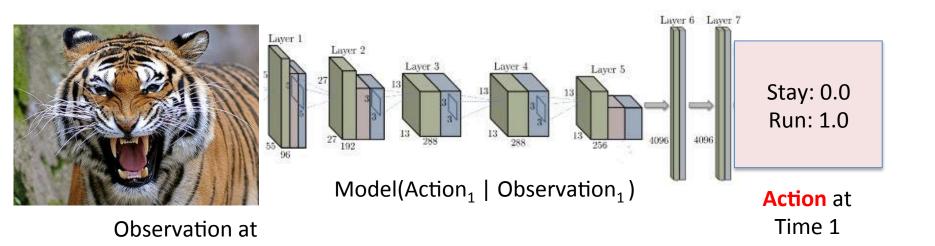




Observation a time 0



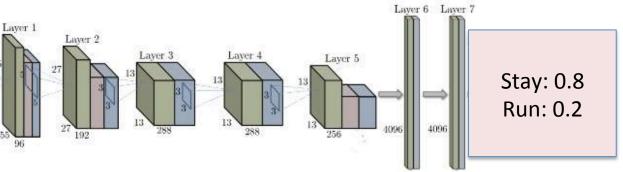




Time 1



Observation at Time t

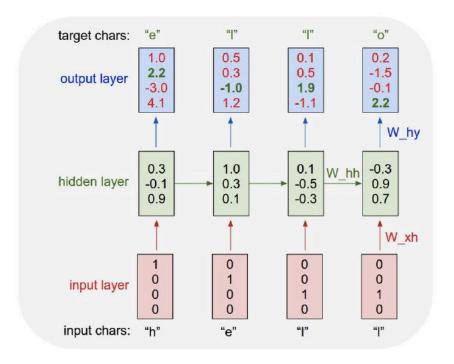


 $Model(Action_t \mid (Observation_{t,}, Action_{t-1}))$ 

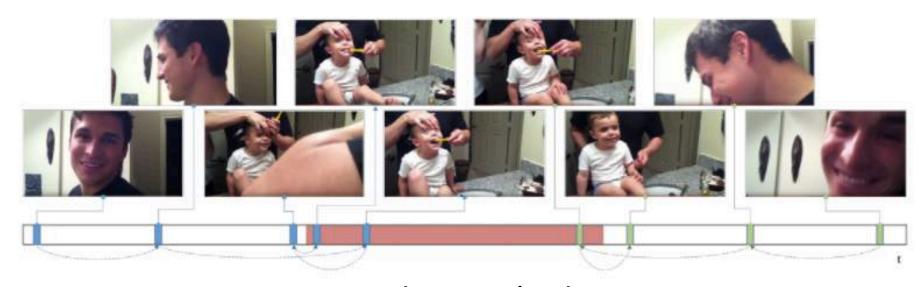
Action at Time t

## Sequential decision problem examples





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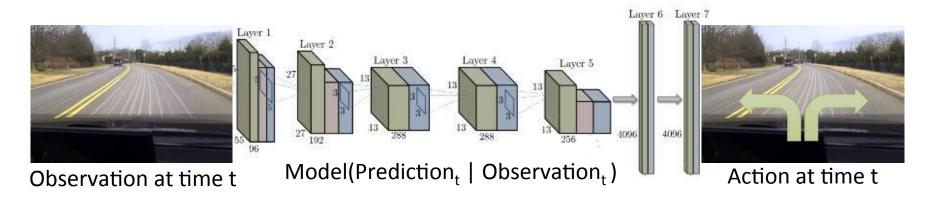


**Learning to search actions** 

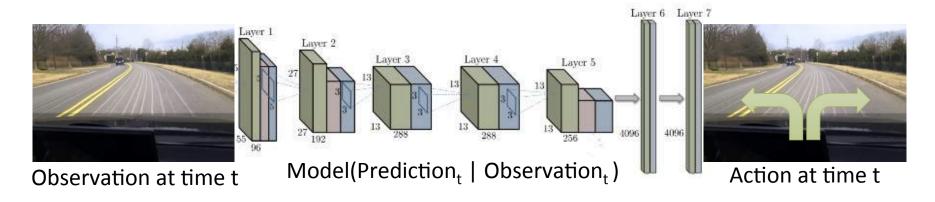
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# Steering wheel prediction



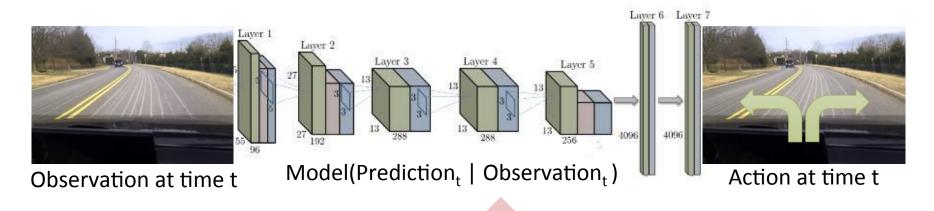
# Steering wheel prediction





Collect observation and actions

# Steering wheel prediction



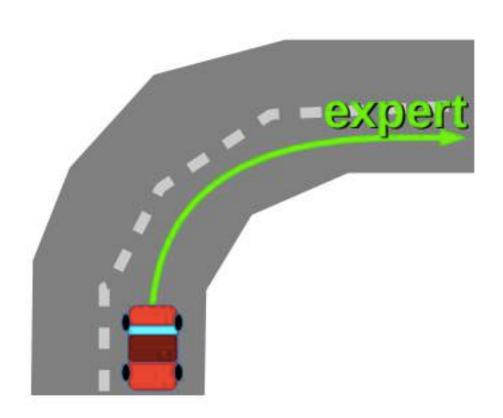




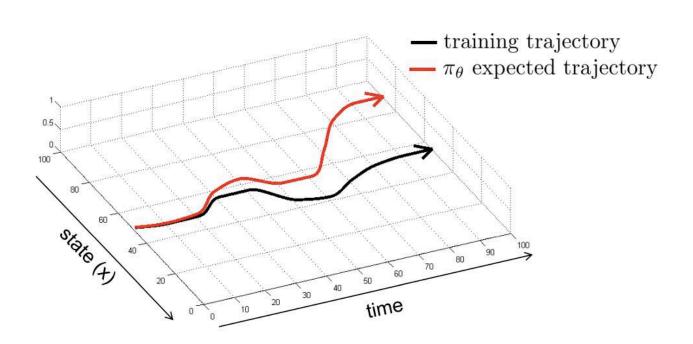
Collect observation and actions

# Should it work?

# Should it work? NO

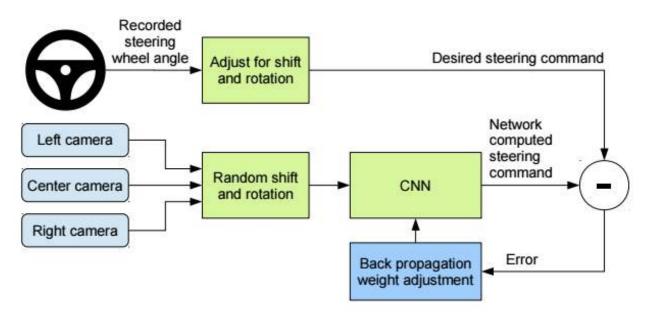


#### Should it work? NO

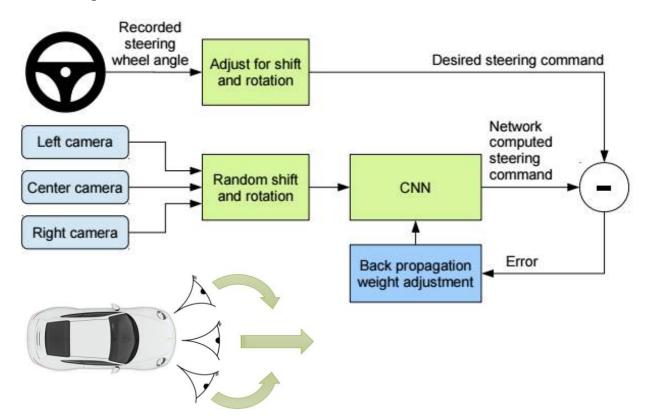


# Does it work? Yes! NVidia

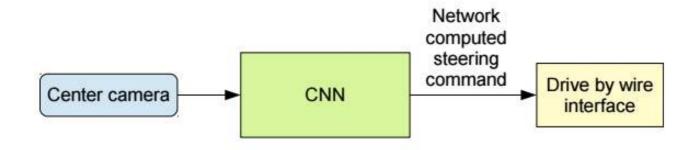
# Why it works?



# Why it works? It's a Hack!

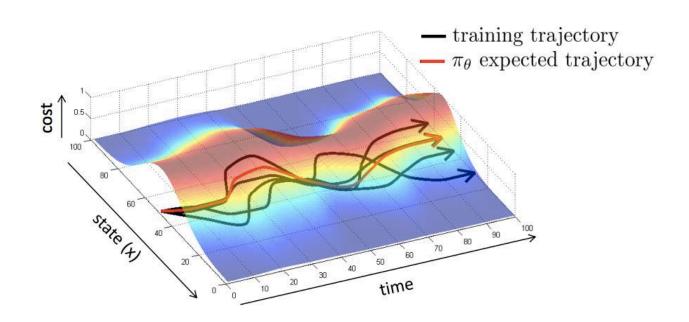


# Why it works? It's a Hack!





#### Can we make it work more often?



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#### **DAgger:** Dataset Aggregation

goal: collect training data from  $p_{\pi_{\theta}}(\mathbf{o}_t)$  instead of  $p_{\text{data}}(\mathbf{o}_t)$ 

how? just run  $\pi_{\theta}(\mathbf{u}_t|\mathbf{o}_t)$ 

but need labels  $\mathbf{u}_t$ !

- 1. train  $\pi_{\theta}(\mathbf{u}_t|\mathbf{o}_t)$  from human data  $\mathcal{D} = \{\mathbf{o}_1, \mathbf{u}_1, \dots, \mathbf{o}_N, \mathbf{u}_N\}$
- 2. run  $\pi_{\theta}(\mathbf{u}_t|\mathbf{o}_t)$  to get dataset  $\mathcal{D}_{\pi} = \{\mathbf{o}_1, \dots, \mathbf{o}_M\}$
- 3. Ask human to label  $\mathcal{D}_{\pi}$  with actions  $\mathbf{u}_t$
- 4. Aggregate:  $\mathcal{D} \leftarrow \mathcal{D} \cup \mathcal{D}_{\pi}$

# DAgger: Watch!

#### Can we make it work more often?

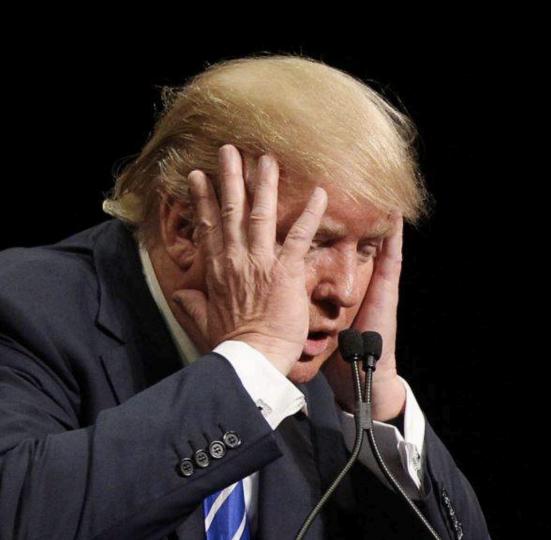
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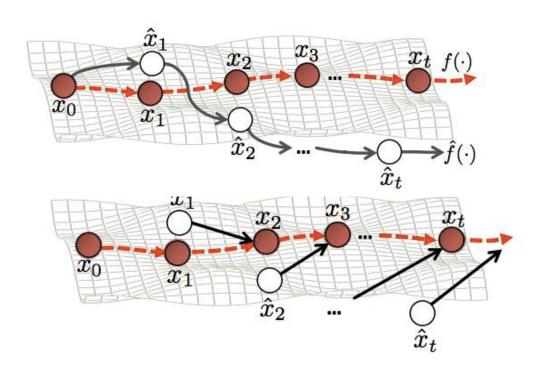
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#### Data as Demonstrator



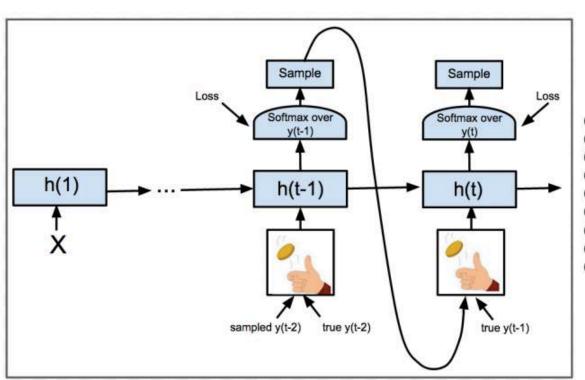
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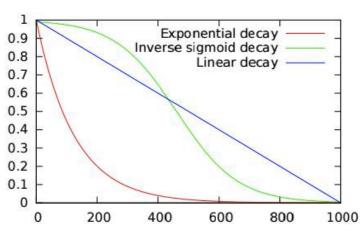
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# Case Study 1: Trail following as classification

Watch!

# Case Study 2: Scheduled Sampling





# **Imitation Learning Summary**

- Some times works well
  - Hacks
  - Intelligent way of getting training samples
  - Really diverse dataset that summarizes the data distribution
- Problems
  - Humans needs to provide data, which is usually finite
  - Humans are not good at provided some kind of actions

# That's all Folks!