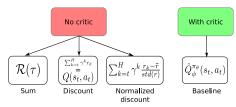
# Policy Gradient in practice Don't become an alchemist :)

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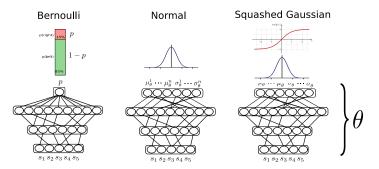
#### Outline



- Investigation of basic REINFORCE phenomena and issues
- Using:
  - ▶ gym "classic control": CartPole, Continuous MountainCar, Pendulum
  - Bernoulli, Normal and squashed Gaussian policies
- Visualization of policies, critics, learning curves
- A prerequisite before going to SOTA deep RL algorithms and harder benchmarks
- Understanding phenomena is better than using black-box algorithms
- Github repo: https://github.com/osigaud/Basic-Policy-Gradient-Labs



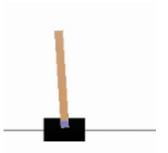
Stochastic policies



- Bernoulli: binary choice between two actions
- Normal: continuous actions, Gaussian, no bounds
- Squashed Gaussian: Normal with bounds



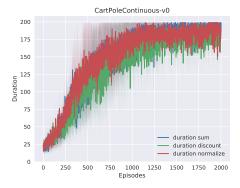
### The CartPole-v0 environment



- The easiest gym classic control environment
- 4 state dimensions:  $x, \dot{x}, \theta, \dot{\theta}$
- Binary action: push left or right. Use discrete or Bernoulli policy
- Custom continuous CartPole to study Gaussian policies (action in [-1,1]
- ▶ 200 steps, +1 at each step  $\rightarrow$  utility in [0, 200]

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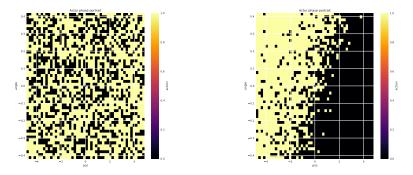
#### Results: Policy Gradient with Bernoulli policy and no baseline



- Variance over 10 runs
- Sum, discounted sum and normalized advantage work well
- No need for additional exploration
- Stochasticity of the binary policy is enough



## ${\sf Initial}/{\sf Final \ policy}$

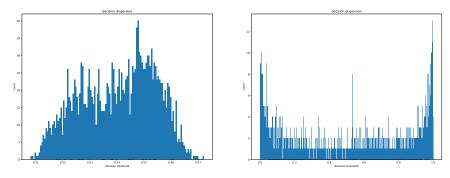


- 4 dimensions:  $x, \dot{x}, \theta, \dot{\theta}$
- FeatureInverter wrapper to show x and  $\theta$  (see video about coding)
- black = push left, yellow = push right
- General idea: push left when right, right when left, then manage pole



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#### Initial/Final randomness



- Mind the scope on x-axis: initially very small (0.5  $\rightarrow$  0.58, not centered)
- At the end of training, the policy is much less stochastic (more 0 and 1)
- Looking for optimality pushes towards less exploration

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Policy Gradient in practice

# Any question?



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