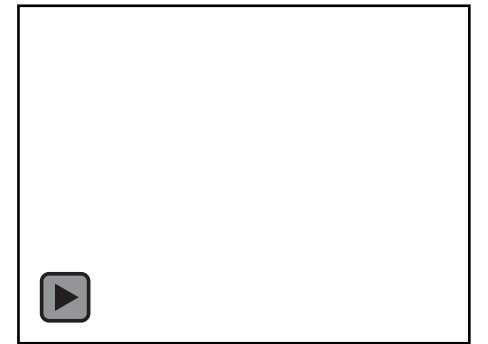
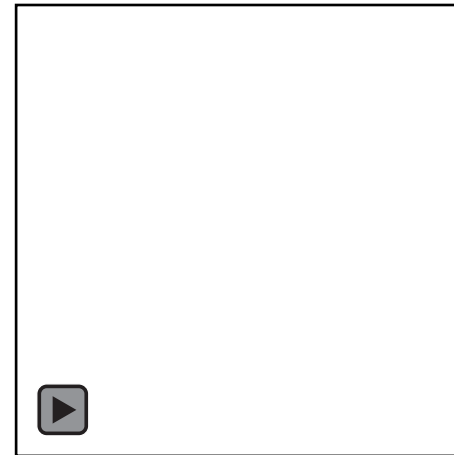
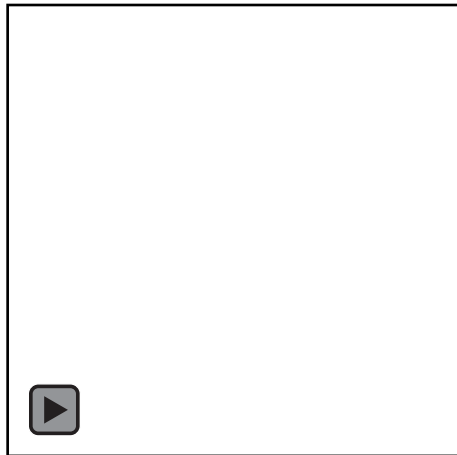


Behavior Cloning

Kyungjae Lee, Yunho Choi, Timothy Ha
RLLAB

[Exercise 1] Gym Environment

- Behavior Cloning Problem
 - Given control data $\{s, a\}$
 - Find controller from data (Supervised learning!)
 - Today, we will use a Gaussian process regression (GPR)
- Gym Environment



[Exercise 1] Gym Environment

- Install gym Environment
 - Pip install gym
- Install JSAnimation
 - git clone <https://github.com/jakevdp/JSAnimation>
 - cd JSAnimation
 - python setup.py install

[Exercise 1] Gym Environment

- Gym Environment

- Import gym `import gym`
- Gym has many simulators

List of Environment

```
print("This simulator has following environments")
envs = [spec.id for spec in envs.registry.all()]
for env in sorted(envs):
    print(env)
```

- We will use MountainCar and Penulum

[Exercise 1] Gym Environment

- Mountain Car Environment

Make simulator

```
env = gym.make('MountainCarContinuous-v0')  
obs = env.reset()
```

State and action

```
obs_space = env.observation_space  
print('Observation space')  
print(type(obs_space))  
print(obs_space.shape)  
print("Dimension:{}".format(obs_space.shape[0]))  
print("High: {}".format(obs_space.high))  
print("Low: {}".format(obs_space.low))  
print()
```

```
act_space = env.action_space  
print('Action space')  
print(type(act_space))  
print("Dimension:{}".format(act_space.shape[0]))  
print("High: {}".format(act_space.high))  
print("Low: {}".format(act_space.low))  
print()
```

Observation space
<class 'gym.spaces.box.Box'>
(2,)
Dimension:2
High: [0.6 0.07]
Low: [-1.2 -0.07]

Action space
<class 'gym.spaces.box.Box'>
Dimension:1
High: [1.]
Low: [-1.]

[Exercise 1] Gym Environment

- Mountain Car Environment

Make simulator

```
env = gym.make('MountainCarContinuous-v0')  
obs = env.reset()
```

Continuous state
3 dimension

Continuous action
1 dimension

State and action

```
obs_space = env.observation_space  
print('Observation space')  
print(type(obs_space))  
print(obs_space.shape)  
print("Dimension:{}".format(obs_space.shape[0]))  
print("High: {}".format(obs_space.high))  
print("Low: {}".format(obs_space.low))  
print()
```

```
act_space = env.action_space  
print('Action space')  
print(type(act_space))  
print("Dimension:{}".format(act_space.shape[0]))  
print("High: {}".format(act_space.high))  
print("Low: {}".format(act_space.low))  
print()
```

Observation space
<class 'gym.spaces.box.Box'>
(2,)
Dimension:2
High: [0.6 0.07]
Low: [-1.2 -0.07]

Action space
<class 'gym.spaces.box.Box'>
Dimension:1
High: [1.]
Low: [-1.]

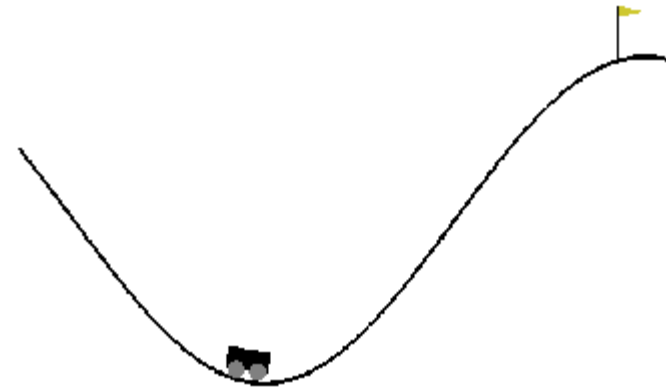
[Exercise 1] Gym Environment

- Mountain Car Environment
 - Visualize simulation
 - `env.render(mode = 'rgb_array')`

Visualization

```
env_img =  
env.close()  
  
plt.title('Environment',{'fontsize':35})  
plt.imshow(env_img)  
plt.axis('off')  
plt.show()
```

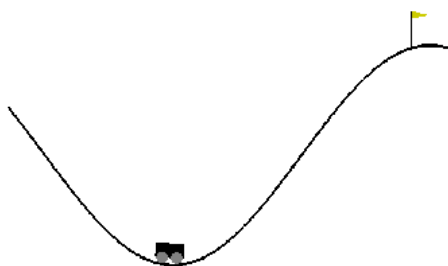
Environment



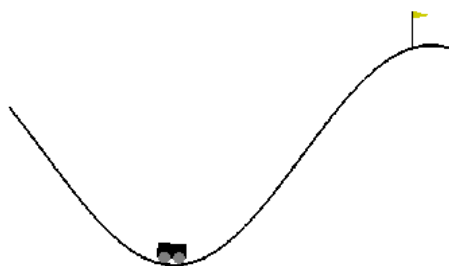
[Exercise 1] Gym Environment

- Mountain Car Environment
 - Control the agent using step() function
 - `action = env.action_space.sample()` (Random action sampler)
 - `env.step(action)`
- When using GPR controller
- `action = GPR(state)`

Environment (t=0)



Environment (t=5)



Control the agent ¶

```
env = gym.make('MountainCarContinuous-v0')
obs = env.reset()
env_img0 = env.render(mode = 'rgb_array')
action = env.action_space.sample()
env.step( )
env.render(mode = 'rgb_array')
action = env.action_space.sample()
env.step( )
env.render(mode = 'rgb_array')
action = env.action_space.sample()
env.step( )
env.render(mode = 'rgb_array')
action = env.action_space.sample()
env.step( )
env.render(mode = 'rgb_array')
action = env.action_space.sample()
env_img1 = env.render(mode = 'rgb_array')
env.close()

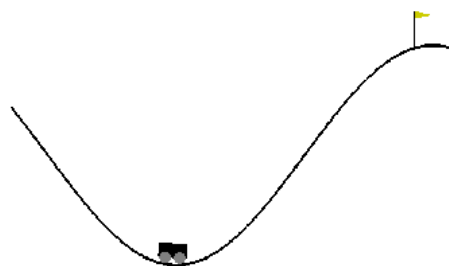
plt.figure()
plt.title('Environment (t=0)', {'fontsize':35})
plt.imshow(env_img0)
plt.axis('off')

plt.figure()
plt.title('Environment (t=5)', {'fontsize':35})
plt.imshow(env_img1)
plt.axis('off')
plt.show()
```

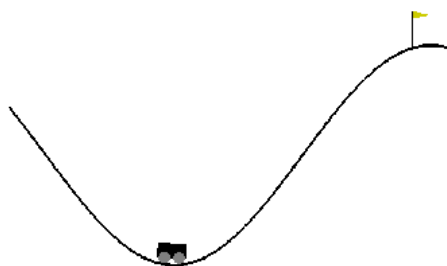

[Exercise 1] Gym Environment

- Mountain Car Environment
 - Control the agent using step() function
 - action = env.action_space.sample() (Random action sampler)
 - env.step(action)
- When using GPR controller
 - ation = GPR(state)

Environment (t=0)



Environment (t=5)



Control the agent ¶

```
env = gym.make('MountainCarContinuous-v0')
obs = env.reset()
env_img0 = env.render(mode = 'rgb_array')
action = env.action_space.sample()
env.step( )
env.render(mode = 'rgb_array')
action = env.action_space.sample()
env.step( )
env.render(mode = 'rgb_array')
action = env.action_space.sample()
env.step( )
env.render(mode = 'rgb_array')
action = env.action_space.sample()
env.step( )
env.render(mode = 'rgb_array')
action = env.action_space.sample()
env_img1 = env.render(mode = 'rgb_array')
env.close()

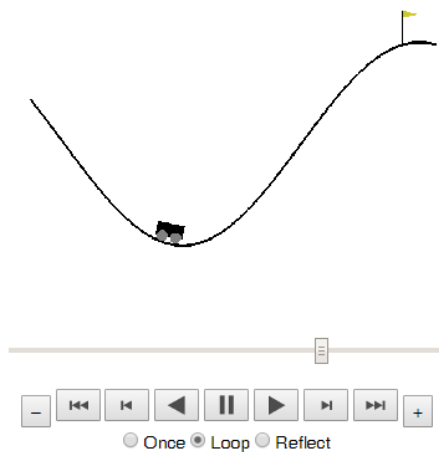
plt.figure()
plt.title('Environment (t=0)', {'fontsize':35})
plt.imshow(env_img0)
plt.axis('off')

plt.figure()
plt.title('Environment (t=5)', {'fontsize':35})
plt.imshow(env_img1)
plt.axis('off')
plt.show()
```

[Exercise 1] Gym Environment

- Mountain Car Environment
 - Output of step function is as follows
 - Next observation
 - Reward
 - Done
 - Info
 - Play it

```
env = gym.make('MountainCarContinuous-v0')
env.reset()
cum_reward = 0
frames = []
for t in range(10000):
    # Render into buffer.
    frames.append(env.render(mode = 'rgb_array'))
    action = env.action_space.sample()
    observation, reward, done, info = env.step(action)
    if done:
        break
env.close()
display_frames_as_gif(frames)
```



[Exercise 2] Behavior Cloning with GPR

- We provide you control data
 - demo_observes : input data
 - 100 by 3 matrix
 - demo_actions : output data
 - 100 by 1 matrix

Load data from pickle file

Data file has (s,a) pairs

```
envname="MountainCarContinuous-v0"

# Load demonstrations
demo_file = open('./'+envname+'_expert_demo.pkl', 'rb')
demonstrations, = pickle.load(demo_file)
demonstrations = shuffle(demonstrations)

# Check expert's performance
exp_ret = np.mean([np.sum(d['rewards']) for d in demonstrations])
print('Expert's Average Cumulative Rewards {:.3f}'.format(exp_ret))

demo_observes = []
demo_actions = []
for demonstration in demonstrations:
    for obs in demonstration['observes']:
        demo_observes.append(obs)
    for act in demonstration['actions']:
        demo_actions.append(act)
demo_observes=np.asarray(demo_observes)
demo_actions=np.asarray(demo_actions)

demo_observes, demo_actions = shuffle(demo_observes, demo_actions)

demo_observes=demo_observes[:100,:]
demo_actions=demo_actions[:100,]
```

Expert's Average Cumulative Rewards 92.459

[Exercise 2] Behavior Cloning with GPR

- Find dimension of state and action
 - `Obs_dim = demo_observations.shape[1]`
 - `Act_dim = demo_actions.shape[1]`
- Define kernel and GP using sklearn
 - `kernel =`
`C(1.0, (1e-3, 1e3)) * RBF(1, (1e-2, 1e2))`
 - `gp =`
`GaussianProcessRegressor(kernel=kernel, n_restarts_optimizer=9)`
 - `gpr =`
`gp.fit(nz_demo_observations, demo_actions)`

Find observation dimension and action dimension

```
obs_dim =  
act_dim =  
  
print("Observation data has shape {}".format(demo_observations.shape))  
print("Action data has shape {}".format(demo_actions.shape))
```

```
Observation data has shape (100, 2)  
Action data has shape (100, 1)
```

Run Gaussian Process Regression ¶

```
kernel =  
gp =  
  
demo_obs_mean = np.mean(demo_observations, axis=0, keepdims=True)  
demo_obs_std = np.std(demo_observations, axis=0, keepdims=True)  
nz_demo_observations = (demo_observations - demo_obs_mean) / demo_obs_std  
  
gpr =
```

[Exercise 2] Behavior Cloning with GPR

- Find dimension of state and action
 - `Obs_dim = demo_observations.shape[1]`
 - `Act_dim = demo_actions.shape[1]`
- Define kernel and GP using sklearn
 - `kernel =`
`C(1.0, (1e-3, 1e3)) * RBF(1, (1e-2, 1e2))`
 - `gp =`
`GaussianProcessRegressor(kernel=kernel, n_restarts_optimizer=9)`
 - `gpr =`
`gp.fit(nz_demo_observations, demo_actions)`

Find observation dimension and action dimension

```
obs_dim =  
act_dim =  
  
print("Observation data has shape {}".format(demo_observations.shape))  
print("Action data has shape {}".format(demo_actions.shape))
```

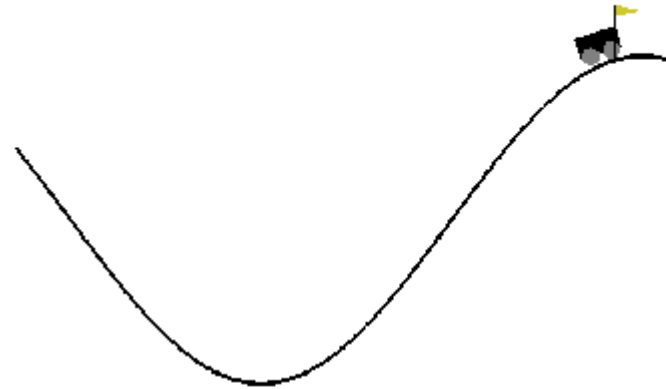
```
Observation data has shape (100, 2)  
Action data has shape (100, 1)
```

Run Gaussian Process Regression ¶

```
kernel =  
gp =  
  
demo_obs_mean = np.mean(demo_observations, axis=0, keepdims=True)  
demo_obs_std = np.std(demo_observations, axis=0, keepdims=True)  
nz_demo_observations = (demo_observations - demo_obs_mean) / demo_obs_std  
  
gpr =
```

[Exercise 2] Behavior Cloning with GPR

- Test GPR controller
 - `action = gp.predict(nz_obs)`

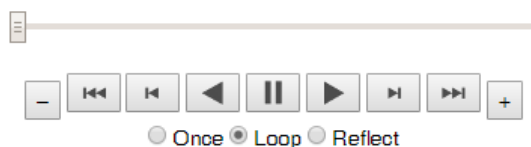


Test GPR controller!

```
env = gym.make(envname)
obs = env.reset()
obs = np.reshape(obs, [1, -1])
total_reward = 0
frames = []
for t in range(10000):
    # Render into buffer.
    frames.append(env.render(mode = 'rgb_array'))
    nz_obs = (obs - demo_obs_mean) / demo_obs_std
    action = |
    obs, reward, done, info = env.step(action)
    obs = np.reshape(obs, [1, -1])
    total_reward += reward
    if done:
        break
env.close()
print('Total Reward : %.2f'%total_reward)
display_frames_as_gif(frames)
```

[Exercise 3] Solve Pendulum-v0

- Load data
- Define GPR using sklearn
- Fit GPR hyper parameters
- Run the trained controller



Load data from pickle file

```
envname="Pendulum-v0"

# Load demonstrations
demo_file = open('./'+envname+'_expert_demo.pkl', 'rb')
demonstrations = pickle.load(demo_file)
demonstrations = shuffle(demonstrations)

# Check expert's performance
exp_ret = np.mean([np.sum(d['rewards']) for d in demonstrations])
print('Expert's Average Cumulative Rewards {:.3f}'.format(exp_ret))
```

Run Gaussian Process Regression

```
kernel =
gp =

demo_obs_mean =
demo_obs_std =
nz_demo_observes =

gpr =
```

Test GPR controller!

```
env = gym.make(envname)
obs = env.reset()
obs = np.reshape(obs, [1, -1])
total_reward = 0
frames = []
for t in range(10000):
    # Render into buffer.
    frames.append(env.render(mode = 'rgb_array'))
    nz_obs =
    action =
    obs, reward, done, info =
    obs =
    total_reward += reward
    if done:
        break
env.close()
print('Total Reward : {:.2f}%total_reward)
display_frames_as_gif(frames)
```