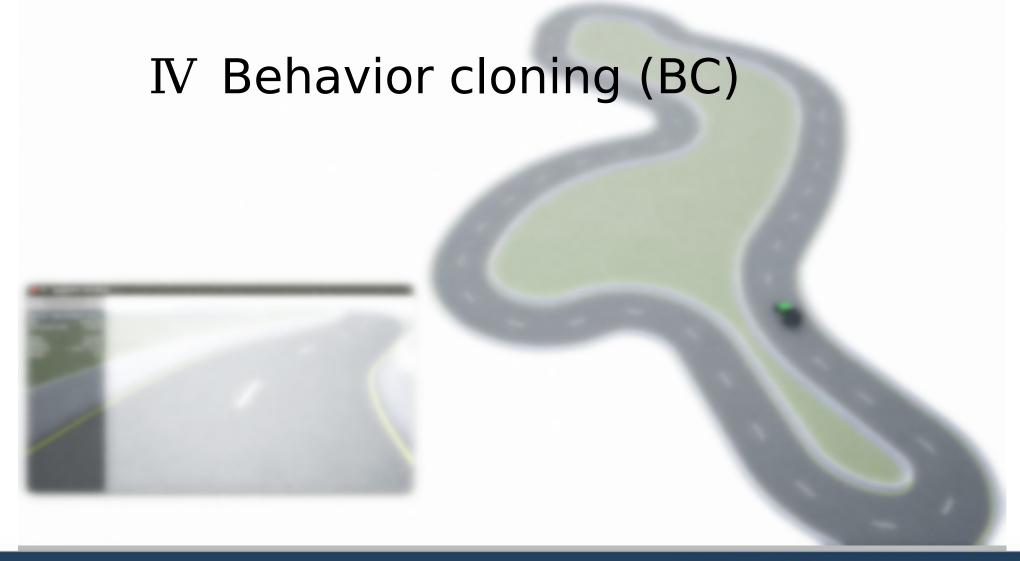


## **MoCAD:** Carla-python Experimental Course







## **Outline**





- 1. Imitation learning—Behavior Cloning
- 2. Collecting data: Autopilot or manual control
- 3. Training Model: Supervise learning
- 4. Applying the model to control vehicle



## 1. Imitation learning—Behavior Cloning



## Imitation learning and why?

An expert (typically a human) provides us with a set of demonstrations. The agent then tries to learn the optimal policy by following, **imitating the expert's** decisions.

Reinforcement learning (RL) reward sparse or hard to design

### IL interacts with the environment

Markov Decision Process (MDP)

- ① an A set of actions
- ② an S set of states
- 3 a P(s'|s,a) transition
- ④ an unknown R(s,a) reward function

The expert's demonstrations  $\tau = (s0, a0, s1, a1, ...)$ , actions are based on the expert's ("optimal")  $\pi^*$  policy

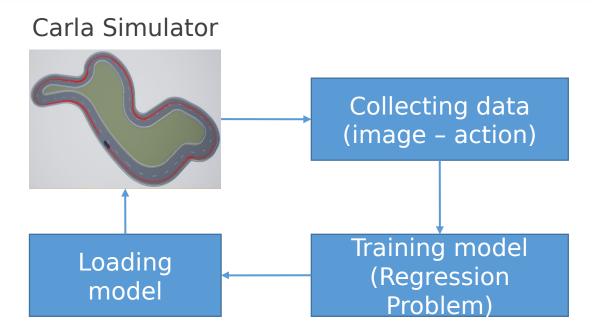


## 1. Imitation learning—Behavior Cloning



## Imitation learning classification

- ① Behavior Cloning
- ② Direct Policy Learning
- ③ Inverse Reinforcement Learning



## Behavior Cloning learning the expert's policy using supervised learning.

- 1. Collect demonstrations ( $\tau^*$  trajectories) from expert
- 2. Treat the demonstrations as i.i.d. state-action pairs:  $(s_0^*, a_0^*), (s_1^*, a_1^*), \dots$
- 3. Learn  $\pi_{\theta}$  policy using supervised learning by minimizing the loss function  $L(a^*, \pi_{\theta}(s))$

**End to end learning** 



## 2. Collecting data: Autopilot or manual control



### Carla Simulator

Mode: Synchronous or **Asynchronous** 

Sensor: **RGB-image**, Depth-image, Lidar

Vehicle: Autopilot or manual control

Action: **Steer**, throttle, brake



## Moving the vehicle and Collecting Data

Control method:

Autopilot | PID

Manual control ☐ Keyboard

Data:

RGB-image & Vehicle States



## 2. Collecting data: Autopilot or manual control



Data: image - action pair

### **Action:** Fixed throttle, steer

```
if keys[K_UP] or keys[K_w]:
    self._control.throttle = min(self._control.throttle + 0.01, 1)
else:
    # self._control.throttle = 0.0
    # fix the velocity
    self._control.throttle = 0.40
```

```
steer_increment = 5e-2 * milliseconds
if keys[K_LEFT] or keys[K_a]:
    if self._steer_cache > 0:
        self._steer_cache = 0
    else:
        self._steer_cache -= steer_increment
elif keys[K_RIGHT] or keys[K_d]:
    if self._steer_cache < 0:
        self._steer_cache = 0
    else:
        self._steer_cache += steer_increment
else:
        self._steer_cache = 0.0
self._steer_cache = min(0.7, max(-0.7, self._steer_cache));
self._control.steer = round(self._steer_cache, 7)</pre>
```

### **RGB-image:** position, rotation, size

```
self.camera transforms = [
   (carla.Transform(carla.Location(x=-5.5, z=2.5), carla.Rotation(pitch=8.0)), Attachment.SpringArm),
   (carla.Transform(
      carla.Location(x=1.6, z=3.0), carla.Rotation(pitch=-30.0)), Attachment.Rigid),
   (carla.Transform(
      carla.Location(x=1, z=2.0), carla.Rotation(pitch=-20.0)), Attachment.Rigid), # see the car
  (carla.Transform(carla.Location(x=5.5, y=1.5, z=1.5)), Attachment.SpringArm),
   (carla.Transform(carla.Location(x=-8.0, z=6.0), carla.Rotation(pitch=6.0)), Attachment.SpringArm),
   (carla.Transform(carla.Location(x=-1, y=-bound_y, z=0.5)), Attachment.Rigid)]
world = self. parent.get world()
bp_library = world.get_blueprint_library()
for item in self.sensors:
      bp = bp library.find(item[0])
      if item[0].startswith('sensor.camera'):
           bp.set attribute('image size x', str(hud.dim[0]));
         bp.set_attribute('image_size_y', str(hud.dim[1]));
           if bp.has attribute('qamma'):
                bp.set_attribute('gamma', str(gamma_correction))
           for attr_name, attr_value in item[3].items():
                bp.set attribute(attr name, attr value)
```



## 2. Collecting data: Autopilot or manual control



### Carla Class

### carla.VehicleControl

Manages the basic movement of a vehicle using typical driving controls.

#### **Instance Variables**

throttle (float)

A scalar value to control the vehicle throttle [0.0, 1.0]. Default is 0.0.

steer (float)

A scalar value to control the vehicle steering [-1.0, 1.0]. Default is 0.0.

brake (float)

A scalar value to control the vehicle brake [0.0, 1.0]. Default is 0.0.

• hand brake (bool)

Determines whether hand brake will be used. Default is **False**.

reverse (bool)

Determines whether the vehicle will move backwards. Default is **False**.

manual\_gear\_shift (bool)

Determines whether the vehicle will be controlled by changing gears manually. Default is **False**.

• gear (int)

States which gear is the vehicle running on.

#### carla.Image

Inherited from carla.SensorData

Class that defines an image of 32-bit BGRA colors that will be used as initial data retrieved by camera sensors. There are different camera sensors (currently three, RGB, depth and semantic segmentation) and each of these makes different use for the images. Learn more about them here.

#### **Instance Variables**

- fov (float degrees)
   Horizontal field of view of the image.
- height (int)
   Image height in pixels.
- width (int)
   Image width in pixels.
- raw\_data (bytes)

#### **Methods**

convert(self, color\_converter)

Converts the image following the color\_converter pattern.

- Parameters:
  - color\_converter (carla.ColorConverter)
- save\_to\_disk(self, path, color\_converter=Raw)

Saves the image to disk using a converter pattern stated as **color\_converter**. The default conversion pattern is **Raw** that will make no changes to the image.

- Parameters:
  - path (str) Path that will contain the image.
  - color\_converter (carla.ColorConverter) Default Raw will make no changes.



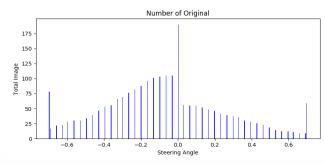
## 3. Training Model: Supervise learning



## Preprocess data:

### **RGB-Image:** x\_train

- ① Resize;
- ② RGB-image to gray-image;
- 3 Crop;
- ① Distribution;



### Vehicle States: y\_label

```
Timestamp
   Y_label: steer
               Throttle
```



## 3. Training Model: Supervise learning

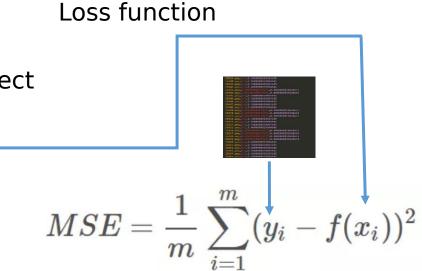


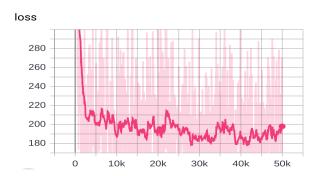
## Training

### X inputs

Model: CNN+BatchNormalization+Fully connect

```
(conv1): Conv2d(1, 24, kernel_size=(5, 5), stride=(2, 2))
(conv1_bn): BatchNorm2d(24, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
(conv2): Conv2d(24, 36, kernel_size=(5, 5), stride=(2, 2))
(conv2_bn): BatchNorm2d(36, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
(conv3): Conv2d(36, 48, kernel_size=(5, 5), stride=(2, 2))
(conv3_bn): BatchNorm2d(48, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
(conv4): Conv2d(48, 64, kernel_size=(3, 3), stride=(1, 1))
(conv4_bn): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
(conv5): Conv2d(64, 64, kernel_size=(3, 3), stride=(1, 1))
(fc1): Linear(in_features=1280, out_features=256, bias=True)
(fc2): Linear(in_features=10, out_features=10, bias=True)
(fc3): Linear(in_features=10, out_features=1, bias=True)
```



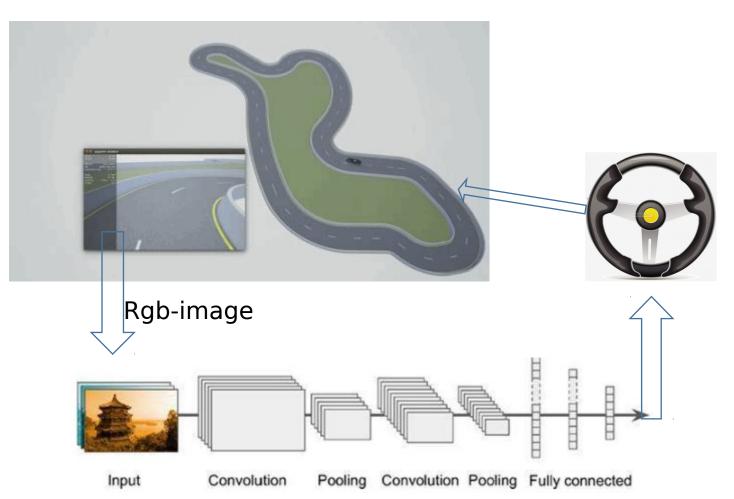




## 4. Control the vehicle



### Demo



```
# get the image from the camera
rgb_image = world.camera_manager.rgb_image
t, s, b = agent.run_step(rgb_image)
# scale [-1, 1]
s /= 100.0  # scaling based on the training model
control = carla.VehicleControl(throttle=t, steer=s, brake=b)
world.player.apply_control(control)
```

```
def run_step(self, sensor_data):
    return self._compute_action(sensor_data)

def _compute_action(self, rgb_image):
    """
    camera_rgb --> model --> action
    """
    # resize 1280*720 --> 256*128
    # rgb->gray
    gray_image = cv2.cvtcolor(rgb_image, cv2.ColoR_BGR2GRAY)
    gray_image = gray_image.reshape(gray_image.shape[0], gray_image.shape[1], 1)
    gray_image = cv2.resize(gray_image, dsize=(256, 128), interpolation=cv2.INTER_CUBIC)
    # cv2.imshow('image', gray_image)
    # cv2.waitKey(0)
    # see the image and execute the action by policy model batch*channel*height*width
    input_image = torch.from_numpy(gray_image).\
         to(device=torch.device('cuda'), dtype=torch.float32).reshape(1, 1, 128, 256)
    lsteer = self.bc_model(input_image).item()
    # throttle and brake set by experiment
    brake = 0.0
    acc = 0.4
    return (acc, steer, brake)
```



## **MoCAD: Carla-python Experimental Course**



# Behavior cloning (BC)

- Testing scenario image does not in the collecting data;
  - Velocity is faster or slower;
  - How to get the perfect performance;

Carla over! But ....



## Intermediate project: Leader-follower instance



Control a vehicle: Traditional or learning

Traditional: Keyboard and self-driving PID

Leader-follower instance: keyboard - PID

### Tips:

- ① How to get the state information of leader?
- ② What does the follower follow?
- 3 How to control two vehicles?

