

Anexo: código de aprendizaje por refuerzo en planeación de movimiento para coche autónomo

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La indentación en este documento ha sido modificada para una mejor visualización.

1 Función de recompensa

```
# Mariana Hernandez
# 150845
import math

def producto_punto(a,b):
    return (a[0]*b[0])+(a[1]*b[1])

def get_norma(a):
    return math.sqrt((a[0]**2)+(a[1]**2))

def normaliza(a):
    if (a[0]==0) and (a[1]==0):
        return [1,0] # para evitar divisiones problematicas
    else:
        norma = get_norma(a)
        return [a[0]/norma,a[1]/norma]

def reward_function(params):

    # recibir parámetros de entrada
    x = params['x'] #actual
    y = params['y'] #actual
    heading = params['heading']
    waypoints = params['waypoints']
    speed = params['speed']
    all_wheels_on_track = params['all_wheels_on_track']
    distance_from_center = params['distance_from_center']
```

```

track_width = params['track_width']
steering = abs(params['steering_angle'])
closest_points = params['closest_waypoints']

# encontrar los vecinos de adelante y atras sobre la pista
next_point = waypoints[closest_points[1]]
prev_point = waypoints[closest_points[0]]

# obtener vectores entre puntos vecinos
next_vector = [next_point[0]-x, next_point[1]-y]
prev_vector = [x-prev_point[0], y-prev_point[1]]

# normalizar vectores
next_vector = normaliza(next_vector)
prev_vector = normaliza(prev_vector)

# calcular angulo entre puntos
producto = producto_punto(next_vector,prev_vector)
norma_next = get_norma(next_vector)
norma_prev = get_norma(prev_vector)
theta = math.acos(producto/(abs(norma_next)*abs(norma_prev)))
theta = theta*(180/math.pi)

# la recompensa empieza con un valor muy baja
reward = 1e-3

# marcas que delimitan posiciones a lo ancho de la pista
marca1 = 0.1 * track_width
marca2 = 0.25 * track_width
marca3 = 0.5 * track_width

# la primer condicion es que el coche se mantenga
# dentro de la pista
if all_wheels_on_track and (0.5*track_width - distance_from_center) >= 0.05:

    # la segunda condicion es que el coche se mantenga
    # cerca de la linea central
    if distance_from_center <= marca1:
        reward = 1
    elif distance_from_center <= marca2:
        reward = 0.5
    elif distance_from_center <= marca3:
        reward = 0.1
    else:
        reward = 1e-3

```

```

# si hay un ángulo entre los puntos
# previo, actual y futuro, tal vez exista una curva
if theta<=20 :

    # tolerancia de steering
    # depende del espacio de accion
    steering_tol = 20

    # si gira mucho el volante se penaliza
    # esto para evitar movimientos zig zag
    if (steering > steering_tol):
        reward *=0.8

elif (theta > 20) and (speed>0.3):

    # viene curva y va rapido
    reward *= 0.4

# evitar que conduzca en reversa
else:
    reward = params['progress']

return float(reward)

```

2 Código para lanzar proyecto

```

# Mariana Hernandez 150845
# codigo para inicializar block de notas AWS
# inicializacion del entrenamiento

#####
### Prerequisites AWS
#####
# #
# # Run these commands if you want to modify the simapp
# #
# # Clean the build directory if present
# !python3 sim_app_bundler.py --clean

# # Download Robomaker simApp from the deepracer public s3 bucket
# simulation_application_bundle_location =
# "s3://deepracer-managed-resources-us-east-1/deepracer-simapp-notebook.tar.gz"
# !aws s3 cp {simulation_application_bundle_location} ./

# # Untar the simapp bundle

```

```

# !python3 sim_app_bundler.py --untar ./deepracer-simapp.tar.gz

# # Now modify the simapp from build directory and run this command.

# # Most of the simapp files can be found here (Robomaker changes)
# # bundle/opt/install/sagemaker_rl_agent/lib/python3.5/site-packages/
# # bundle/opt/install/deepracer_simulation_environment/share/
# deepracer_simulation_environment/
# # bundle/opt/install/deepracer_simulation_environment/lib/
# deepracer_simulation_environment/

# # Copying the notebook src/markov changes to the simapp
# (For sagemaker container)
# !rsync -av ./src/markov/ ./build/simapp/bundle/opt/install/sagemaker_rl_agent/
# lib/python3.5/site-packages/markov

# !python3 sim_app_bundler.py --tar

#####
### Imports
#####

import boto3
import sagemaker
import sys
import os
import re
import numpy as np
import subprocess
sys.path.append("common")
from misc import get_execution_role, wait_for_s3_object
from docker_utils import build_and_push_docker_image
from sagemaker.rl import RLEstimator, RLToolkit, RLFramework
from time import gmtime, strftime
import time
# from IPython.display import Markdown
# from markdown_helper import *

#####
# Inicializar parametros basicso
#####

# Select the instance type
instance_type = "ml.c4.2xlarge"
#instance_type = "ml.p2.xlarge"
#instance_type = "ml.c5.4xlarge"

```

```

# Starting SageMaker session
sage_session = sagemaker.session.Session()

# Create unique job name.
job_name_prefix = 'deepracer-notebook'

# Duration of job in seconds (1 hours)
job_duration_in_seconds = 3600

# AWS Region
aws_region = sage_session.boto_region_name
if aws_region not in ["us-west-2", "us-east-1", "eu-west-1"]:
    raise Exception("This notebook uses RoboMaker which is available
        only in US East (N. Virginia), "US West (Oregon) and EU (Ireland).
        Please switch to one of these regions.")

#####
# Setup S3 bucket
#####

# S3 bucket
s3_bucket = sage_session.default_bucket()

# SDK appends the job name and output folder
s3_output_path = 's3://{}/'.format(s3_bucket)

#Ensure that the S3 prefix contains the keyword 'sagemaker'
s3_prefix = job_name_prefix + "-sagemaker-" + strftime("%y%m%d-%H%M%S", gmtime())

# Get the AWS account id of this account
sts = boto3.client("sts")
account_id = sts.get_caller_identity()['Account']

print("Using s3 bucket {}".format(s3_bucket))
print("Model checkpoints and other metadata will be stored at:
\ns3://{}/{}".format(s3_bucket, s3_prefix))

#####3
# Create an IAM role
#####

try:
    sagemaker_role = sagemaker.get_execution_role()
except:
    sagemaker_role = get_execution_role('sagemaker')

```

```

print("Using Sagemaker IAM role arn: \n{}".format(sagemaker_role))

#####
# Permission setup for invoking AWS RoboMaker from this notebook
#####

display(Markdown(generate_help_for_robomaker_trust_relationship(sagemaker_role)))

#####
# Permission setup for Sagemaker to S3 bucket
#####

display(Markdown(generate_s3_write_permission_for_sagemaker_role(sagemaker_role)))

#####
# Permission setup for Sagemaker to create KinesisVideoStreams
#####

display(Markdown(generate_kinesis_create_permission_for_sagemaker_role(sagemaker_role)))

#####
# Build and push docker image
#####

%%time
from copy_to_sagemaker_container import get_sagemaker_docker,
copy_to_sagemaker_container, get_custom_image_name
cpu_or_gpu = 'gpu' if instance_type.startswith('ml.p') else 'cpu'
repository_short_name = "sagemaker-docker-%s" % cpu_or_gpu
custom_image_name = get_custom_image_name(repository_short_name)
try:
    print("Copying files from your notebook to existing sagemaker container")
    sagemaker_docker_id = get_sagemaker_docker(repository_short_name)
    copy_to_sagemaker_container(sagemaker_docker_id, repository_short_name)
except Exception as e:
    print("Creating sagemaker container")
    docker_build_args = {
        'CPU_OR_GPU': cpu_or_gpu,
        'AWS_REGION': boto3.Session().region_name,
    }
    custom_image_name = build_and_push_docker_image(repository_short_name,
    build_args=docker_build_args)
    print("Using ECR image %s" % custom_image_name)

```

```

#####33
# Configure VPC
#####3

ec2 = boto3.client('ec2')

#
# Check if the user has Deepracer-VPC and use that if its present.
# This will have all permission.
# This VPC will be created when you have used the Deepracer console and
# created one model atleast
# If this is not present. Use the default VPC connection
#
deepracer_security_groups = [group["GroupId"] for group in
ec2.describe_security_groups()
['SecurityGroups']\
if group['GroupName'].startswith("aws-deepracer-")]

# deepracer_security_groups = False
if(deepracer_security_groups):
    print("Using the DeepRacer VPC stacks. This will be created if you run
one training job from console.")
    deepracer_vpc = [vpc['VpcId'] for vpc in ec2.describe_vpcs()['Vpcs'] \
        if "Tags" in vpc for val in vpc['Tags'] \
        if val['Value'] == 'deepracer-vpc'][0]
    deepracer_subnets = [subnet["SubnetId"] for subnet in
ec2.describe_subnets()["Subnets"] \
        if subnet["VpcId"] == deepracer_vpc]
else:
    print("Using the default VPC stacks")
    deepracer_vpc = [vpc['VpcId'] for vpc in ec2.describe_vpcs()['Vpcs'] if
vpc["IsDefault"] == True][0]

    deepracer_security_groups = [group["GroupId"] for
group in ec2.describe_security_groups()['SecurityGroups'] \

    if 'VpcId' in group and group["GroupName"] == "default" and
group["VpcId"] == deepracer_vpc]

    deepracer_subnets = [subnet["SubnetId"] for
subnet in ec2.describe_subnets()["Subnets"] \
    if subnet["VpcId"] == deepracer_vpc and subnet['DefaultForAz']==True]

print("Using VPC:", deepracer_vpc)
print("Using security group:", deepracer_security_groups)
print("Using subnets:", deepracer_subnets)

```

```

#####
# Create Route Table
#####

#TODO: Explain to customer what CREATE_ROUTE_TABLE is doing
CREATE_ROUTE_TABLE = True

def create_vpc_endpoint_table():
    print("Creating ")
    try:
        route_tables = [route_table["RouteTableId"] for route_table in
            ec2.describe_route_tables()['RouteTables']\
                if route_table['VpcId'] == deepracer_vpc]
    except Exception as e:
        if "UnauthorizedOperation" in str(e):
            display(Markdown(generate_help_for_s3_endpoint_permissions(sagemaker_role)))
        else:
            display(Markdown(create_s3_endpoint_manually(aws_region, deepracer_vpc)))
        raise e

    print("Trying to attach S3 endpoints to the following route tables:", route_tables)

    if not route_tables:
        raise Exception(("No route tables were found.
            Please follow the VPC S3 endpoint creation "
            "guide by clicking the above link."))

    try:
        ec2.create_vpc_endpoint(DryRun=False,
            VpcEndpointType="Gateway",
            VpcId=deepracer_vpc,
            ServiceName="com.amazonaws.{}.s3".format(aws_region),
            RouteTableIds=route_tables)
        print("S3 endpoint created successfully!")
    except Exception as e:
        if "RouteAlreadyExists" in str(e):
            print("S3 endpoint already exists.")
        elif "UnauthorizedOperation" in str(e):
            display(Markdown(generate_help_for_s3_endpoint_permissions(role)))
            raise e
        else:
            display(Markdown(create_s3_endpoint_manually(aws_region, deepracer_vpc)))
            raise e

if CREATE_ROUTE_TABLE:
    create_vpc_endpoint_table()

```



```

# Mariana Hernandez 150845
# configuracion del entrenamiento
# editar el archivo default.py con la funcion de recompensa deseada

#####
# Configure the preset for RL algorithm
#####

# Uncomment the pygmentize code lines to see the code

# Environmental File
#!pygmentize src/markov/environments/deepracer_racetrack_env.py

# Reward function
# modificar el archivo default.py con la funcion deseada
!pygmentize src/markov/rewards/default.py

# Action space
#!pygmentize src/markov/actions/model_metadata_10_state.json

# Preset File
#!pygmentize src/markov/presets/default.py
#!pygmentize src/markov/presets/preset_attention_layer.py

# Mariana Hernandez 150845
# entrenamiento del modelo

#####
# Copy custom files to S3 bucket so that sagemaker & robomaker can pick it up
#####

s3_location = "s3://%s/%s" % (s3_bucket, s3_prefix)
print(s3_location)

# Clean up the previously uploaded files
!aws s3 rm --recursive {s3_location}

# Make any changes to the environment and preset files below and upload these files
!aws s3 cp src/markov/environments/deepracer_racetrack_env.py
{s3_location}/environments/deepracer_racetrack_env.py

!aws s3 cp src/markov/rewards/default.py {s3_location}/rewards/reward_function.py

!aws s3 cp src/markov/actions/model_metadata_10_state.json

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{s3_location}/model_metadata.json

!aws s3 cp src/markov/presets/default.py
{s3_location}/presets/preset.py
#!aws s3 cp src/markov/presets/preset_attention_layer.py
{s3_location}/presets/preset.py

#####
# Train the RL model using the Python SDK Script mode
#####

metric_definitions = [
    # Training> Name=main_level/agent, Worker=0, Episode=19, Total
    # reward=-102.88, Steps=19019, Training iteration=1
    {'Name': 'reward-training',
     'Regex': '^Training>.*Total reward=(.*?),'},

    # Policy training> Surrogate loss=-0.32664725184440613, KL
    # divergence=7.255815035023261e-06, Entropy=2.83156156539917,
    # training epoch=0, learning_rate=0.00025
    {'Name': 'ppo-surrogate-loss',
     'Regex': '^Policy training>.*Surrogate loss=(.*?),'},
    {'Name': 'ppo-entropy',
     'Regex': '^Policy training>.*Entropy=(.*?),'},

    # Testing> Name=main_level/agent, Worker=0, Episode=19, Total
    # reward=1359.12, Steps=20015, Training iteration=2
    {'Name': 'reward-testing',
     'Regex': '^Testing>.*Total reward=(.*?),'},
]

estimator = RLEstimator(entry_point="training_worker.py",
                        source_dir='src',
                        image_name=custom_image_name,
                        dependencies=["common/"],
                        role=sagemaker_role,
                        train_instance_type=instance_type,
                        train_instance_count=1,
                        output_path=s3_output_path,
                        base_job_name=job_name_prefix,
                        metric_definitions=metric_definitions,
                        train_max_run=job_duration_in_seconds,
                        hyperparameters={
                            "s3_bucket": s3_bucket,
                            "s3_prefix": s3_prefix,
                            "aws_region": aws_region,
                        })

```

```

        "preset_s3_key": "%s/presets/preset.py"%
        s3_prefix,
        "model_metadata_s3_key":

        "%s/model_metadata.json" % s3_prefix,
        "environment_s3_key":
        "%s/environments/deepracer_racetrack_env.py"
        % s3_prefix,
    },
    subnets=deepracer_subnets,
    security_group_ids=deepracer_security_groups,
)

estimator.fit(wait=False)
job_name = estimator.latest_training_job.job_name
print("Training job: %s" % job_name)

#####
# Create the Kinesis video stream
#####

kvs_stream_name = "dr-kvs-{}".format(job_name)

!aws --region {aws_region} kinesismvideo create-stream --stream-name
{kvs_stream_name} --media-type video/h264 --data-retention-in-hours 24
print ("Created kinesis video stream {}".format(kvs_stream_name))

#####
# Start the Robomaker job
#####

robomaker = boto3.client("robomaker")

#####
# Create Simulation Application
#####

robomaker_s3_key = 'robomaker/simulation_ws.tar.gz'
robomaker_source = {'s3Bucket': s3_bucket,
                    's3Key': robomaker_s3_key,
                    'architecture': "X86_64"}
simulation_software_suite={'name': 'Gazebo',
                           'version': '7'}
robot_software_suite={'name': 'ROS',
                      'version': 'Kinetic'}
rendering_engine={'name': 'OGRE',

```

```

        'version': '1.x'}

if not os.path.exists('./build/output.tar.gz'):
    print("Using the latest simapp from public s3 bucket")
    # Download Robomaker simApp for the deepracer public s3 bucket
    simulation_application_bundle_location = "s3://deepracer-managed-resources-us-east-1/
    deepracer-simapp-notebook.tar.gz"
    !aws s3 cp {simulation_application_bundle_location} ./

    # Remove if the Robomaker sim-app is present in s3 bucket
    !aws s3 rm s3://{s3_bucket}/{robomaker_s3_key}

    # Uploading the Robomaker SimApp to your S3 bucket
    !aws s3 cp ./deepracer-simapp-notebook.tar.gz s3://{s3_bucket}/{robomaker_s3_key}

    # Cleanup the locally downloaded version of SimApp
    !rm deepracer-simapp-notebook.tar.gz
else:
    print("Using the simapp from build directory")
    !aws s3 cp ./build/output.tar.gz s3://{s3_bucket}/{robomaker_s3_key}

    app_name = "deepracer-notebook-application" + strftime("%y%m%d-%H%M%S", gmtime())

print(app_name)
try:
    response = robomaker.create_simulation_application(name=app_name,
                                                    sources=[robomaker_source],
                                                    simulationSoftwareSuite=
                                                    simulation_software_suite,
                                                    robotSoftwareSuite=
                                                    robot_software_suite,
                                                    renderingEngine=
                                                    rendering_engine)

    simulation_app_arn = response["arn"]
    print("Created a new simulation app with ARN:", simulation_app_arn)
except Exception as e:
    if "AccessDeniedException" in str(e):
        display(Markdown(generate_help_for_robomaker_all_permissions(role)))
        raise e
    else:
        raise e

#####
# Launch the Simulation job on RoboMake
#####

```

```

num_simulation_workers = 1

environ_vars = {
    "WORLD_NAME": "reinvent_base",
    "KINESIS_VIDEO_STREAM_NAME": kvs_stream_name,
    "SAGEMAKER_SHARED_S3_BUCKET": s3_bucket,
    "SAGEMAKER_SHARED_S3_PREFIX": s3_prefix,
    "TRAINING_JOB_ARN": job_name,
    "APP_REGION": aws_region,
    "METRIC_NAME": "TrainingRewardScore",
    "METRIC_NAMESPACE": "AWSDeepRacer",
    "REWARD_FILE_S3_KEY": "%s/rewards/reward_function.py" % s3_prefix,
    "MODEL_METADATA_FILE_S3_KEY": "%s/model_metadata.json" % s3_prefix,
    "METRICS_S3_BUCKET": s3_bucket,
    "METRICS_S3_OBJECT_KEY": s3_bucket + "/training_metrics.json",
    "TARGET_REWARD_SCORE": "None",
    "NUMBER_OF_EPISODES": "0",
    "ROBOMAKER_SIMULATION_JOB_ACCOUNT_ID": account_id
}

simulation_application = {"application":simulation_app_arn,
    "launchConfig": {"packageName":
        "deepracer_simulation_environment",
        "launchFile":
            "distributed_training.launch",
        "environmentVariables": environ_vars}
    }

vpcConfig = {"subnets": deepracer_subnets,
    "securityGroups": deepracer_security_groups,
    "assignPublicIp": True}

responses = []
for job_no in range(num_simulation_workers):
    client_request_token = strftime("%Y-%m-%d-%H-%M-%S", gmtime())
    response = robomaker.create_simulation_job(iamRole=sagemaker_role,
        clientRequestToken=client_request_token,
        maxJobDurationInSeconds=
            job_duration_in_seconds,
        failureBehavior="Continue",
        simulationApplications=
            [simulation_application],
        vpcConfig=vpcConfig
    )

    responses.append(response)

```

```

print("Created the following jobs:")
job_arns = [response["arn"] for response in responses]
for response in responses:
    print("Job ARN", response["arn"])

#####
# Visualizing the simulations in RoboMaker
#####

display(Markdown(generate_robomaker_links(job_arns, aws_region)))

#####
# Creating temporary folder top plot metrics
#####

tmp_dir = "/tmp/{}".format(job_name)
os.system("mkdir {}".format(tmp_dir))
print("Create local folder {}".format(tmp_dir))

#####
# Plot metrics for training job
#####

%matplotlib inline
import pandas as pd
import json

training_metrics_file = "training_metrics.json"
training_metrics_path = "{}/{}/{}".format(s3_bucket, training_metrics_file)
wait_for_s3_object(s3_bucket, training_metrics_path, tmp_dir)

json_file = "{}/{}/{}".format(tmp_dir, training_metrics_file)
with open(json_file) as fp:
    data = json.load(fp)

df = pd.DataFrame(data['metrics'])
x_axis = 'episode'
y_axis = 'reward_score'

plt = df.plot(x=x_axis, y=y_axis, figsize=(12,5), legend=True, style='b-')
plt.set_ylabel(y_axis);
plt.set_xlabel(x_axis);

#####
# Evaluation - ReInvent Track

```

```
#####

sys.path.append("./src")

num_simulation_workers = 1

environ_vars = {
    "WORLD_NAME": "reinvent_base",
    "KINESIS_VIDEO_STREAM_NAME": "SilverstoneStream",
    "MODEL_S3_BUCKET": s3_bucket,
    "MODEL_S3_PREFIX": s3_prefix,
    "APP_REGION": aws_region,
    "MODEL_METADATA_FILE_S3_KEY": "%s/model_metadata.json" % s3_prefix,
    "METRICS_S3_BUCKET": s3_bucket,
    "METRICS_S3_OBJECT_KEY": s3_bucket + "/evaluation_metrics.json",
    "NUMBER_OF_TRIALS": "5",
    "ROBOMAKER_SIMULATION_JOB_ACCOUNT_ID": account_id
}

simulation_application = {
    "application": simulation_app_arn,
    "launchConfig": {
        "packageName": "deepracer_simulation_environment",
        "launchFile": "evaluation.launch",
        "environmentVariables": environ_vars
    }
}

vpcConfig = {"subnets": deepracer_subnets,
             "securityGroups": deepracer_security_groups,
             "assignPublicIp": True}

responses = []
for job_no in range(num_simulation_workers):
    response = robomaker.create_simulation_job(clientRequestToken=
        strftime("%Y-%m-%d-%H-%M-%S", gmtime()),
        outputLocation={
            "s3Bucket": s3_bucket,
            "s3Prefix": s3_prefix
        },
        maxJobDurationInSeconds=
            job_duration_in_seconds,
        iamRole=sagemaker_role,
        failureBehavior="Continue",
        simulationApplications=
            [simulation_application],
```

```

vpcConfig=vpcConfig)

responses.append(response)

# print("Created the following jobs:")
for response in responses:
    print("Job ARN", response["arn"])

#####
# Creating temporary folder to plot metrics
#####

evaluation_metrics_file = "evaluation_metrics.json"
evaluation_metrics_path = "{}/{}/{}".format(s3_bucket, evaluation_metrics_file)
wait_for_s3_object(s3_bucket, evaluation_metrics_path, tmp_dir)

json_file = "{}/{}/{}".format(tmp_dir, evaluation_metrics_file)
with open(json_file) as fp:
    data = json.load(fp)

df = pd.DataFrame(data['metrics'])
# Converting milliseconds to seconds
df['elapsed_time'] = df['elapsed_time_in_milliseconds']/1000
df = df[['trial', 'completion_percentage', 'elapsed_time']]

display(df)

```