

# Teach a Quadcopter How to Fly

## REVIEW

### Meets Specifications

Hello,

You've done a great work by implementing the DDPG and to teach quadcopter how to fly. I have provided some feedback on your agent code and the **reward** function. Please go through it. I am including this [resource](#) which is about DDPG algorithm. It would be nice if you can have a look at it.

Since you have completed this project, Now if you want to know how to apply these techniques to real-world problems. For that, I would advise you to go through [Deep Reinforcement Learning for Self Driving Car by MIT](#). So that you can know more about reinforcement learning algorithms in a broader and real-world perspective. As you asked for more feedback, I have provided them in Suggestion section.

### Define the Task, Define the Agent, and Train Your Agent!



The `agent.py` file contains a functional implementation of a reinforcement learning algorithm.

#### Awesome

- Successful implementation of the DDPG algorithm.
- Actor and critic network are correctly implemented.
- It was good to use target network for both actor and critic network as suggested by the author of DDPG Paper.
- Further using soft updates for the target network was a good choice.



The `Quadcopter_Project.ipynb` notebook includes code to train the agent.

### Plot the Rewards



A plot of **rewards** per episode is used to illustrate how the agent learns over time.

### Reflections



The submission describes the task and **reward** function, and the description lines up with the implementation in `task.py`. It is clear how the **reward** function can be used to guide the agent to accomplish the task.

## Awesome

- Task and **Reward** function are explained properly.

## Suggestion

Try using a small number of units and compare the performance. For example:

- [128, 128, 128]
- [128, 64, 128]
- [64, 128, 256]

Use batch normalization for all the layers. Deep networks are prone to gradient vanishing and exploding problems.



The submission provides a detailed description of the agent in `agent.py`.

## Awesome

- Good job by choosing DDPG algorithm for continuous state space problem.
- Good job by including hyperparameters and network architecture in the notebook.

## Suggestion

Exploration parameters are very important for a good learning. Please refer to this [Link](#) to know more about the exploration strategies in reinforcement learning.



The submission discusses the **rewards** plot. Ideally, the plot shows that the agent has learned (with episode **rewards** that are gradually increasing). If not, the submission describes in detail various attempted settings (hyperparameters and architectures, etc) that were tested to teach the agent.

## Awesome

- It is quite common to see agent not learning ideally in `Quadcopter` project since it is not an easy project. And you have done a good job so far.
- Good job providing the details of various hyperparameters and **reward** settings.

## Suggestion

- You can add z-axis velocity in reward function to encourage quadcopter to fly towards the target.

- You can subtract angular velocity from the reward to make sure quadcopter flies straight up.
- You can include some large bonus and penalty rewards also. Such as a bonus on achieving the target height and a penalty on crashing.
- Clip your final reward between  $(-1, 1)$ . It will definitely help in better performance.



A brief overall summary of the experience working on the project is provided, with ideas for further improving the project.