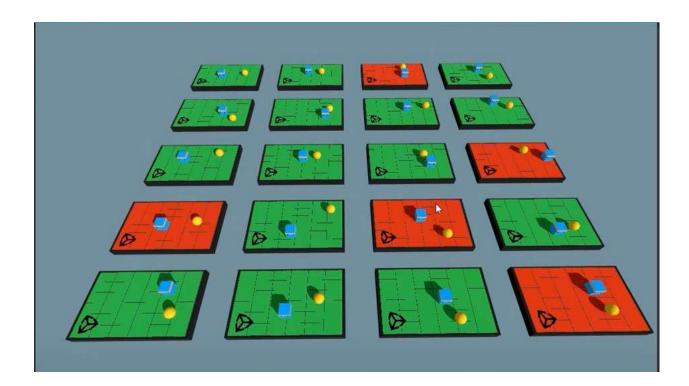
1. Introduction to ML-Agents

ML-Agents (Machine Learning Agents) is a Unity toolkit that allows developers to train intelligent agents using reinforcement learning, imitation learning, and heuristic methods.



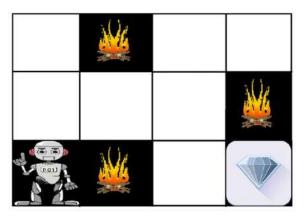
ML-Agents Approaches or Learning Methods

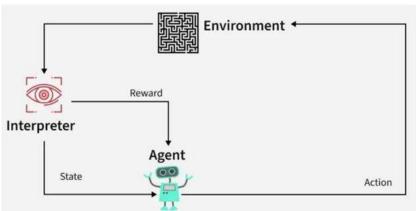
ML-Agents is a Unity toolkit that allows developers to train intelligent agents using different learning approaches. The main methods include:

- Reinforcement Learning (RL) Learning through trial and error.
- Imitation Learning (IL) Learning by mimicking an expert.

• **Heuristic Methods** - Using manually coded behaviors.







1. Reinforcement Learning (RL)

How It Works:

- The agent interacts with the environment and receives **rewards** for good actions and **penalties** for bad ones.
- Over time, the agent improves by maximizing the rewards.
- Uses Proximal Policy Optimization (PPO) to train.

Example:

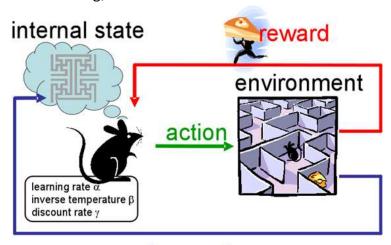
A self-driving car learns to stay on the road by receiving positive rewards for staying on track and negative rewards for going off-road.

ML-Agents Setup:

- **Behavior Type:** Default (Agent trains when learning is active)
- Training Command:

mlagents-learn config.yaml --run-id=MyTraining --no-graphics

- Saving & Using the Model:
 - o After training, the model is saved and can be used for inference.



observation

2. Imitation Learning (IL)

How It Works:

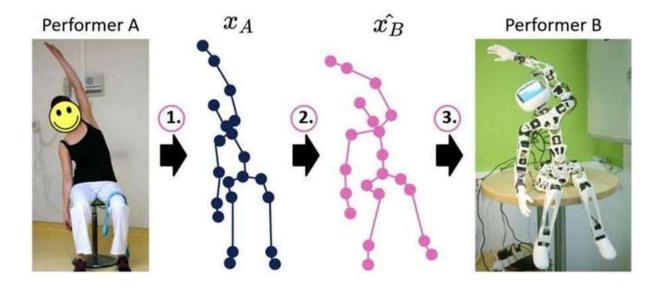
- The agent learns from a pre-recorded expert's actions instead of trial and error.
- Uses Behavioral Cloning (BC) or Generative Adversarial Imitation Learning (GAIL).

Example:

A robot arm learns to pick up objects by imitating a human's recorded actions.

ML-Agents Setup:

- **Behavior Type:** Player (During recording) → Default (During training)
- Steps:
 - o **Record Expert Actions** using the heuristic method.
 - o **Train the agent** using the recorded data.



3. Heuristic Methods

How It Works:

- The agent does **not learn** but follows **predefined rules** in the script.
- A **Heuristic () function** is manually written to control the agent's behavior.

Example:

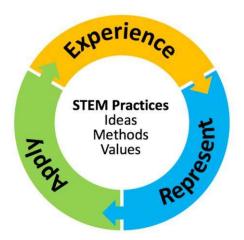
An NPC enemy follows a fixed patrol path without adapting to the player's movements.

ML-Agents Setup:

- **Behavior Type:** Heuristic Only
- Example Script:

```
public override void Heuristic(in ActionBuffers actionsOut)
{
    var continuousActions = actionsOut.ContinuousActions;
    continuousActions[0] = Input.GetAxis("Horizontal");
    continuousActions[1] = Input.GetAxis("Vertical");
}
```

- **Heuristic learning** is mainly for **debugging and manual control**, not for Al-driven decision-making and it allows you to **manually define** how the agent behaves using the Heuristic () method in the script.
- It does **not** use **machine learning** or training, and it is useful for **testing sensors**, **physics**, **and input controls** before switching to Al training.
- For training an Al model to make decisions and move toward a target, you need Reinforcement Learning (RL) or Imitation Learning (IL).







4. Installation & Setup

Installing ML-Agents

- 1. Install Python (3.8+ recommended).
- 2. Create a virtual environment:

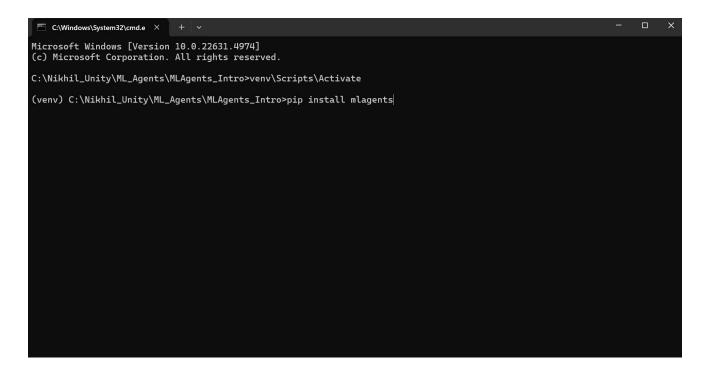
```
python -m venv venv
source venv/bin/activate (Linux/Mac)
venv\Scripts\activate (Windows)
```

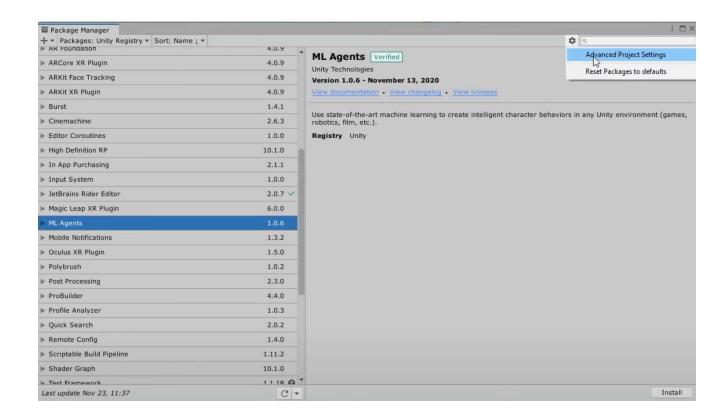
3. Install ML-Agents package:

pip install mlagents==0.29.0

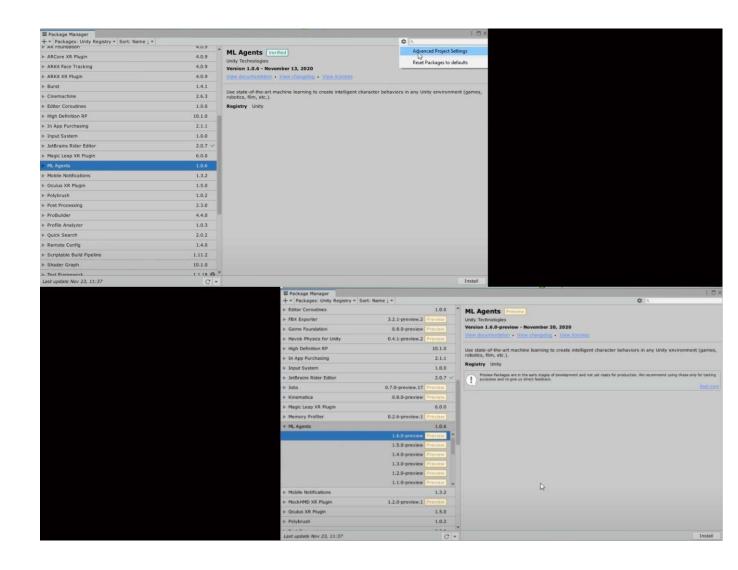
4. Verify installation:

mlagents-learn -help



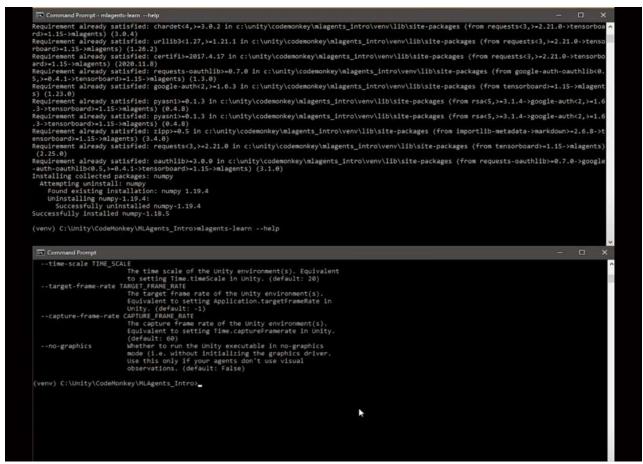


5.Additional Information to Preview Old Versions of Packages:

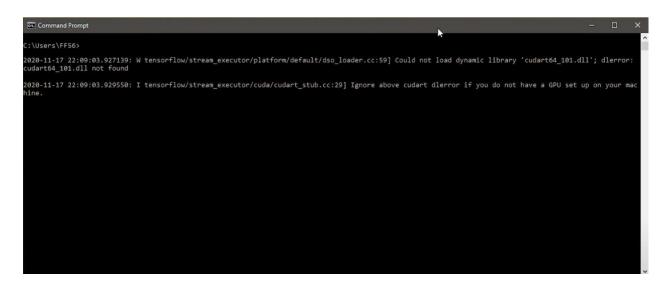


6. Pytorch Packages Installation & Setup

7. Verification, Installation & Setup of mlagents package



8. mlagents Dependencies Not Necessary Otherwise) CUDA || Cudnn Libraries Shown In This Case:



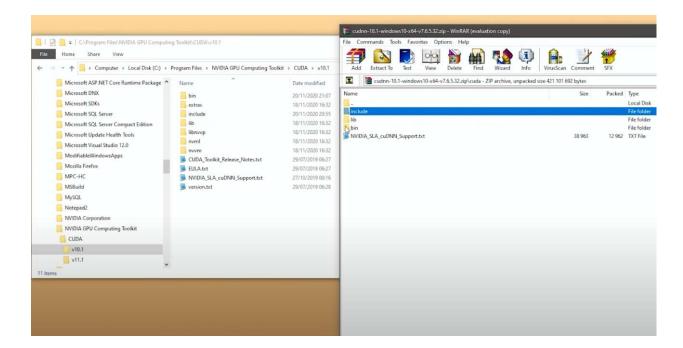
9. mlagents Dependencies Verification, Installation: CUDA || Cudnn Libraries As Shown Use These Content

Install Recommended Versions Package To Not Facing Dependecies Conflicts

- 1. Install Python (3.8.10 recommended): Python 3.8.10
- 2. Install CUDA (11.8.0 recommended): CUDA 11.8.0
- 3. Install Cudnn (Cudnn v8.9.7 recommended): Cudnn v8.9.7

Setup Cudnn into CUDA Package After Installation as Shown in Image:

When installing **cuDNN** for **CUDA**, you need to manually copy specific files from the **cuDNN package** into the **CUDA installation folder**. Follow these steps:



1 Locate Your CUDA and cuDNN Installation Paths

• CUDA Installation Path (Default):

makefile
CopyEdit

C:\Program Files\NVIDIA GPU Computing Toolkit\CUDA\vXX.X

(Replace XX. X with your CUDA version, e.g., v12.2)

• cuDNN Extracted Folder:

makefile

CopyEdit

C:\Users\YourUsername\Downloads\cudnn-windows-x86_64-X.X.X.X_cudaXX.X

(Replace X.X.X.X with your cuDNN version and XX.X with your CUDA version)

2 Copy These Files From cuDNN to CUDA

cuDNN Folder	Copy To (CUDA Folder)	Files to Copy
bin\	<pre>C:\Program Files\NVIDIA GPU Computing</pre>	cudnn64_8.dll
	Toolkit\CUDA\vXX.X\bin	(or latest version)
include\	<pre>C:\Program Files\NVIDIA GPU Computing</pre>	cudnn.h
	Toolkit\CUDA\vXX.X\include	
lib\x64\	<pre>C:\Program Files\NVIDIA GPU Computing</pre>	cudnn.lib
	Toolkit\CUDA\vXX.X\lib\x64	Cuuliii. 11D

3 Verify Installation

After copying the files, verify that CUDA and cuDNN are correctly installed:

• Check CUDA Version

cmd
CopyEdit
nvcc --version

• Check If PyTorch Detects CUDA & cuDNN

```
cmd
CopyEdit
python -c "import torch; print(torch.cuda.is_available())"
cmd
CopyEdit
```

```
python -c "import torch; print(torch.backends.cudnn.version())"
```

4 Set Environment Variables (If Needed)

If CUDA/cuDNN files are not detected, add these paths to the **Windows Environment Variables**:

- 1. Press Win + $R \rightarrow Type sysdm.cpl \rightarrow Go to Advanced > Environment Variables$.
- 2. Under **System Variables**, find Path → Click **Edit** → Add these:

vbnet

CopyEdit

C:\Program Files\NVIDIA GPU Computing Toolkit\CUDA\vXX.X\bin
C:\Program Files\NVIDIA GPU Computing Toolkit\CUDA\vXX.X\lib\x64

3. Click OK and Restart Your PC.

5 Final Verification

Run this command again to ensure everything is set up properly:

```
\mathsf{cmd}
```

```
CopyEdit
```

```
python -c "import torch; print(torch.cuda.is_available(),
torch.backends.cudnn.version())"
```

4. Install ML-Agents (0.29.0 recommended):
 Just by this command: pip install mlagents==0.29.0

Verify Installed Packages with Commands:

- 1. Python: python -version
- 2. CUDA: nvcc -version
- 3. Cudnn Didn't Need As We have to copy some of constraints from these into CUDA Package Folder

4. ML-Agents: pip show mlagents

Behavior Type Options in Unity

Behavior Type	Description	When to Use
Default	Trains during mlagents-learn, switches to	Recommended for
	inference when not training.	training agents.
Inference Only	Uses a trained model, does not learn.	For testing a trained
		model in Unity.
Heuristic Only	Uses manually defined rules in the script.	For debugging or full
		manual control.
		Useful for imitation
Player	Allows human control.	learning setups.

Key Training Commands

Command	Description
mlagents-learn config.yamlrun-	Starts training in headless
id=MyTrainingno-graphics	mode.
mlagents-learn config.yamlinitialize-	Resumes training from a
from=MyTrainingrun-id=MyTraining2	previous run.
mlagents-learn config.yamlrun-	Forces overwriting of
id=MyTrainingforce	previous training data.
mlagents-learn config.yamlrun-	Runs the trained model in
id=MyTraininginference	inference mode.

Best Practices for Training ML-Agents

- Longer training improves performance but takes time.
- Use **reward shaping** to guide learning more efficiently.

- Test different hyperparameters (e.g., batch size, learning rate).
- Use Curriculum Learning for complex tasks.
- Save trained models periodically to avoid loss of progress.

10. Behavior Types in Unity

In the Unity Inspector, the **Behavior Type** setting determines how the agent acts:

- **Default**: Trains when learning is active; switches to inference when training stops.
- Inference Only: Uses a pre-trained model (no learning).
- **Heuristic Only**: Uses manual logic (custom script in Heuristic () function).
- Player: Controlled by a human (useful for imitation learning).

11. Training an Agent

Basic Command for Training

mlagents-learn config.yaml --run-id=MoveToGoal --no-graphics

- --run-id=MoveToGoal: Saves training data under this ID.
- --no-graphics: Runs Unity in headless mode (no rendering for faster training).

Resume Training from Checkpoint

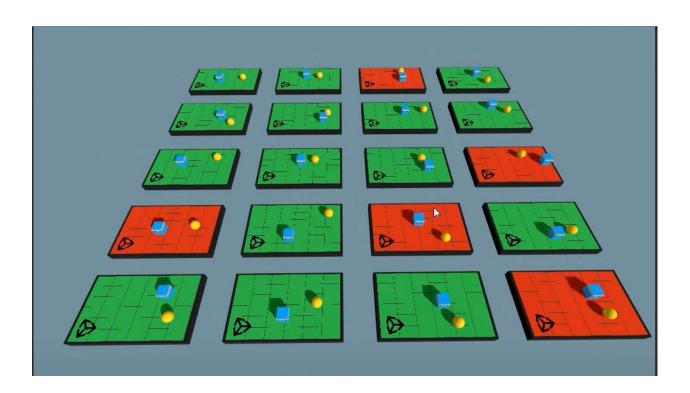
mlagents-learn config.yaml --run-id=MoveToGoal --resume --no-graphics

Override Previous Training Data

mlagents-learn config.yaml --run-id=MoveToGoal --force --no-graphics



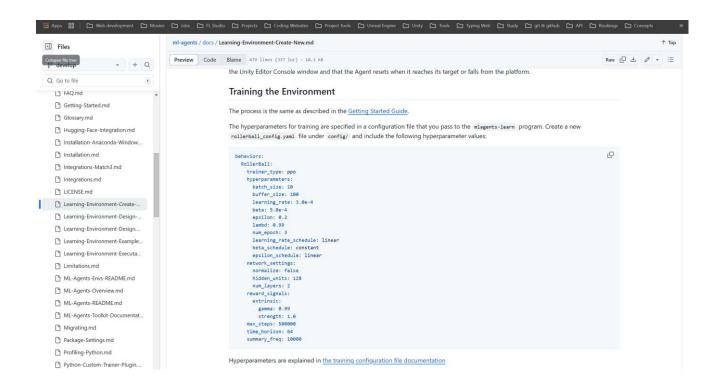
While Training And Saving Model And Make him agent Like (Agent Vinod):



But before this we are facing issues like if we move the target object itself then the ai still having follow the previous training path which is not correct. So, the below recommendation point is a resolver for this issue, we have to update config, yaml that give like trainer to train specific model in given recommendations.

12. ML-Agents Configuration (config.yaml) File Update:

The below yaml file is taken from this that the developer has made and post it on to resolve previous issue and the link is: <u>Link: Go To Training the Environment and copy the below yaml file code</u>



Key parameters in YAML configuration file:

We have to replace "RollerBall" With "MoveToGoal" and We are done.

behaviors:

MoveToGoal:

trainer_type: ppo # Proximal Policy Optimization (RL Algorithm)
hyperparameters:

batch_size: 1024
buffer_size: 10240
learning_rate: 0.0003
max_steps: 500000 # Total training steps
summary_freq: 50000 # How often to log progress
checkpoint_interval: 500000 # When to save models
network_settings:
 hidden_units: 128
 num_layers: 2
reward_signals:
 extrinsic:
 gamma: 0.99
 strength: 1.0

Now we have another issue as we know about ${f ML-Agents\ Approaches\ or}$

Learning Methods, so the thing is if we make a focus on this references:

1.



2. Agent Class Script: Which we are inheriting that inherited monobehaviour with some interfaces and with constraints that needed to work with mlagents in unity to train Al based Models:

```
using System;
using System.Collections.Generic;
using System.Collections.ObjectModel;
using Votem.Collections.ObjectModel;
using Unity.Barracuda;
using Unity.MLAgents.Actuators;
using Unity.MLAgents.Demonstrations;
using Unity.MLAgents.Policies;
using Unity.MLAgents.Sensors;
using Unity.MLAgents.Sensors.Reflection;
using UnityEngine;
using UnityEngine;
using UnityEngine;
 amespace Unity.MLAgents;
[Serializable]
[HelpURL("https://github.com/Unity-Technologies/ml-agents/blob/release_17_docs/docs/Learning-Environment-Design-Agents.md")]
[RequireComponent(typeof(BehaviorParameters))]
public class Agent : MonoBehaviour, ISerializationCallbackReceiver, IActionReceiver, IHeuristicProvider
      internal struct AgentParameters
            public int maxStep;
      private enum DoneReason
            DoneCalled,
MaxStepReached,
Disabled
      private IPolicy m_Brain;
      private BehaviorParameters m_PolicyFactory;
     [HideInInspector]
internal AgentParameters agentParameters;
      [SerializeField]
      [HideInInspector]
internal bool hasUpgradedFromAgentParameters;
      [FormerlySerializedAs("maxStep")]
     [HideInInspector]
public int MaxStep;
      private AgentInfo m_Info;
     private float m_Reward;
      private float m_GroupReward;
     private float m_CumulativeReward;
     private bool m_RequestAction;
     private bool m_RequestDecision;
      private int m_StepCount;
```

3. Custom Agent Script:

Main Objectives of the Custom Agent Script

Your custom agent script does the following:

1. Observing the Environment (CollectObservations)

- a. The agent **perceives** the environment by collecting observations (e.g., position, velocity, surroundings).
- b. This data is passed to the neural network for decision-making.

2. Taking Actions (OnActionReceived)

- c. The agent receives actions from the ML model (or a heuristic function).
- d. It then applies those actions to control movement, rotation, jumping, etc.

3. Defining Rewards (SetReward & AddReward)

- e. Assign rewards for good behavior (e.g., moving towards a goal).
- f. Assign penalties for bad behavior (e.g., colliding with walls).
- g. This helps the agent learn optimal strategies.

4. Handling Resets (OnEpisodeBegin)

- h. Reset conditions at the start of each training episode.
- i. This ensures fair training by reinitializing positions, parameters, etc

Custom Agent Script: MoveToGoalAgent:

```
MoveToGoalAgent.cs ₹ ×
                                                                                                                  → ੴ MoveToGoalAgent
 Assembly-CSharp
                using Unity.MLAgents.Actuators;
using Unity.MLAgents.Sensors;
                using UnityEngine;
using UnityEngine.Assertions.Must;
               ⊕ Unity Script (2 asset references) | O references
∨public class MoveToGoalAgent : Agent
{
                      [SerializeField] private Transform targetTransform;
                      [SerializeField] private float moveSpeed;
[SerializeField] private Material winMaterial;
[SerializeField] private Material lossMaterial;
                      [SerializeField] private MeshRenderer floorMeshRenderer;
                     public override void OnEpisodeBegin()
        170
                          transform.localPosition = Vector3.zero;
                      public override void CollectObservations(VectorSensor sensor)
                           sensor.AddObservation(transform.localPosition)
                           sensor.AddObservation(targetTransform.localPosition);
                      O references
public override void OnActionReceived(ActionBuffers actions)
        26
27
28
29
                           float moveX = actions.ContinuousActions[0];
float moveZ = actions.ContinuousActions[1];
        30
31
                           transform.localPosition += new Vector3(moveX, 0, moveZ) * Time.deltaTime * moveSpeed;
        34
35
36
                      public override void Heuristic(in ActionBuffers actionsOut)
                           ActionSegment<float> continuousActions = actionsOut.ContinuousActions;
                           continuousActions[0] = Input.GetAxis("Horizontal");
continuousActions[1] = Input.GetAxis("Vertical");
                     © Unity Message | O references
private void OnTriggerEnter(Collider other)
                           if (other.TryGetComponent<Goal>(out Goal goal))
                                SetReward(+1f);
                                 floorMeshRenderer.material = winMaterial;
                                EndEpisode():
                           if (other.TryGetComponent<Wall>(out Wall wall))
                                SetReward(-1f);
                                 floorMeshRenderer.material = lossMaterial;
                                EndEpisode();
```

Now, Remember the highlighted mark of blue color that will create issue like:

Let's assume I moved target that follow up by player or agent at runtime and that will not follow that path because the AI model is created or trained as we resetting target or agent or player position to vector zero, so it will not configuring all directions path or exploring more paths if object goes to other positions,

So, the thing is we have to give random positions to target or player or agent so it will take also random paths because as objects positions changes then paths exploring possibilities increases with respect to it and the complex model will be ready to

solve complex paths efficiently.

So, the given change we made to custom agent script is this to resolve this issue by giving random reset positions to both target or player or agent to eat as random observation and take actions and follow up life cycle approaches or methods AI Based algorithms as in given reference.

Updated Custom Agent Script: MoveToGoalAgent:

```
MoveToGoalAgent.cs 😃 🗙
Assembly-CSharp

→ <sup>Ag</sup> MoveToGoalAgent

               | using Unity.MLAgents;
| using Unity.MLAgents.Actuators;
| using Unity.MLAgents.Sensors;
| using UnityEngine.Assertions.Must;
| using UnityEngine.Assertions.Must;
                   © Unity Script (2 asset references) | 0 references

✓public class MoveToGoalAgent : Agent

f
                            [SerializeField] private Transform targetTransform;
[SerializeField] private float moveSpeed;
[SerializeField] private Material winMaterial;
[SerializeField] private Material lossMaterial;
[SerializeField] private MeshRenderer floorMeshRenderer;
                             public override void OnEpisodeBegin()
                                 transform.localPosition = new Vector3(Random.Range(3.8f, -3.8f), transform.localPosition.y, Random.Range(3.8f, -3.8f));
targetTransform.localPosition = new Vector3(Random.Range(3.8f, -3.8f), targetTransform.localPosition.y, Random.Range(3.8f, -3.8f));
                             public override void CollectObservations(VectorSensor sensor)
                                   sensor.AddObservation(transform.localPosition);
sensor.AddObservation(targetTransform.localPosition);
                             public override void OnActionReceived(ActionBuffers actions)
                                   float moveX = actions.ContinuousActions[0];
float moveZ = actions.ContinuousActions[1];
                                   transform.localPosition += new Vector3(moveX, 0, moveZ) * Time.deltaTime * moveSpeed;
                             public override void Heuristic(in ActionBuffers actionsOut)
                                   ActionSegment<float> continuousActions = actionsOut.ContinuousActions;
continuousActions[0] = Input.GetAxis("Horizontal");
continuousActions[1] = Input.GetAxis("Vertical");
                                   if (other.TryGetComponent<Goal>(out Goal goal))
                                         SetReward(+1f);
floorMeshRenderer.material = winMaterial;
                                          EndEpisode();
                                         SetReward(-1f);
floorMeshRenderer.material = lossMaterial;
EndEpisode();
```

How It Works in ML-Agents Training

1. Training Phase:

- a. The agent starts at a random position.
- b. It tries different movements using reinforcement learning.
- c. It gets rewarded for moving towards the target.
- d. It gets penalized for moving away.
- e. The model updates based on rewards, improving over time.

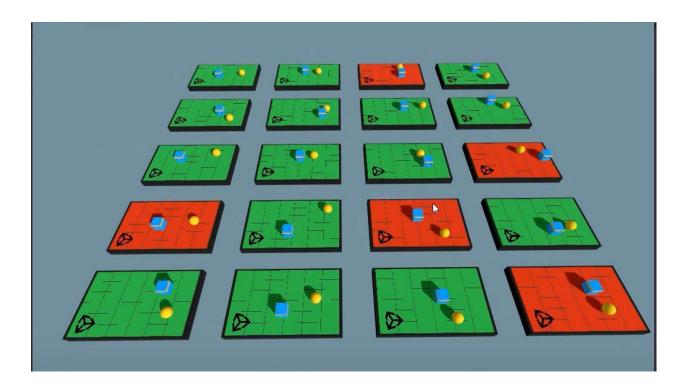
2. **Inference Phase** (Using Trained Model):

- a. After training, the agent applies its learned model to navigate toward the goal optimally.
- b. No further learning occurs; it simply executes decisions based on past training.

♦ Summary

Your custom agent script defines what the agent sees, how it acts, how it learns, and how it resets. This is essential for training Al-controlled characters in Unity ML-Agents.

After 5 Steps We have to setup to train AI Based Model || While Training It looks like this:



13. Understanding Training Phases

Each **step** of ML-Agent training goes through:

- 1. **Observation**: Agent receives inputs from environment.
- 2. **Decision Making:** ML model predicts best action.

- 3. **Action Execution**: Agent performs the action.
- 4. **Reward Assignment**: Agent gets feedback on performance.
- 5. **Learning Update**: Model is updated to improve future decisions.

14. Reinforcement Learning (RL) vs Imitation Learning

- **Reinforcement Learning (RL)**: The agent learns from trial and error based on rewards.
- Imitation Learning: The agent learns by mimicking human-controlled actions.

Which One to Use?

- Use **RL** if the agent needs to discover solutions on its own.
- Use **Imitation Learning** if you want to guide the agent's behavior based on expert demonstrations.

15. Deploying a Trained Model

- 1. Train the model.
- 2. Locate the trained .onnx file in results/MoveToGoal/.
- 3. In Unity, assign the .onnx file to the **Behavior Parameters** of your agent.
- 4. Set Behavior Type to Inference Only.

16. Optimizing Training Performance

🔽 Increase training time to allow the agent to fully learn. 🔽 Improve the reward function
to encourage perfect behaviors. 🗹 Reduce exploration as training progresses to refine
decisions. 🔽 Use curriculum learning to gradually increase difficulty.

17. Useful Commands

Command

Description

```
mlagents-learn config.yaml --run-id=MoveToGoal
--no-graphics
mlagents-learn config.yaml --run-id=MoveToGoal
--resume --no-graphics
mlagents-learn config.yaml --run-id=MoveToGoal
--force --no-graphics
mlagents-learn config.yaml --run-id=MoveToGoal
--initialize-from=MoveToGoal2

Start training

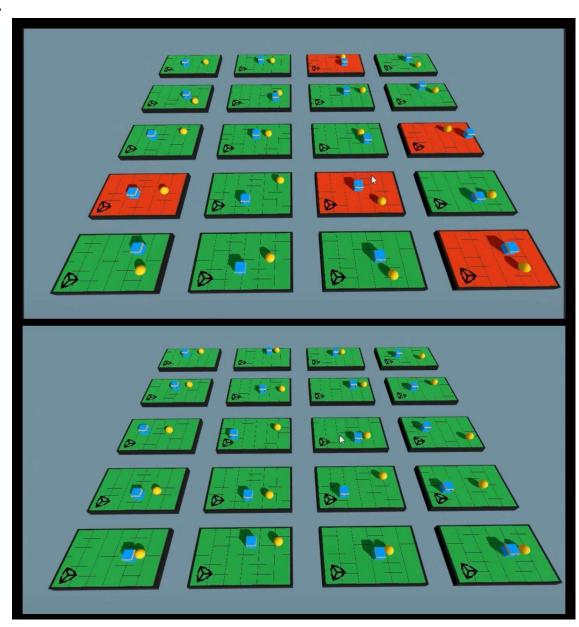
Coverride old training data
Initialize training from a
previous model
```

Conclusion

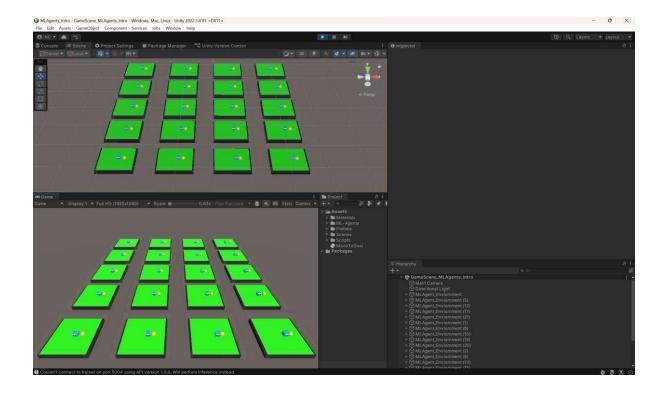
By understanding and tweaking these settings, you can train ML-Agents efficiently and optimize their performance for different use cases in Unity.

17. Results:

1.



2. In My System Implementation of AI Based Trained Model Setup and Ready to Make Decisions on its Own:



18. Link of Project If Wanted a Just Basic Setup for Work with MLAgents Instant In Unity:

- 1. GitHub Link: MLAgentBasicStartup Demo
- 2. MLAgents Package Requirements GoogelDrive Link:

MLAgents Packages Requirements



For attending the Orientation Session

Your participation is what make us Successful