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**MineRL**

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# TUTORIALS AND GUIDES

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Welcome to documentation for the [MineRL](#) project and its related repositories and components!



## WHAT IS MINERL

MineRL is a research project started at Carnegie Mellon University aimed at developing various aspects of artificial intelligence within Minecraft. In short MineRL consists of several major components:

- [MineRL-v0 Dataset](#) – One of the largest imitation learning datasets with over **60 million frames** of recorded human player data. The dataset includes a [set of environments](#) which highlight many of the hardest problems in modern-day Reinforcement Learning: sparse rewards and hierarchical policies.
- `minerl` – A rich python3 package for doing artificial intelligence research in Minecraft. This includes two major submodules. We develop `minerl` in our spare time, [please consider supporting us on Patreon](#)
  - `minerl.env` – A growing set of OpenAI Gym environments in Minecraft. These environments leverage a **synchronous, stable**, and **fast** fork of Microsoft Malmo called *MineREnv*.
  - `minerl.data` – The main python module for ext with the *MineRL-v0* dataset

### 1.1 Installation

Welcome to MineRL! This guide will get you started.

To start using the MineRL dataset and Gym environments comprising MineRL, you'll need to install the main python package, `minerl`.

1. First **make sure you have JDK 1.8** installed on your system.
  - a. [Windows installer](#) – On windows go this link and follow the instructions to install JDK 8.
  - b. On Mac, you can install java8 using homebrew and AdoptOpenJDK (an open source mirror, used here to get around the fact that Java8 binaries are no longer available directly from Oracle):

```
brew tap AdoptOpenJDK/openjdk
brew install --cask adoptopenjdk8
```

- c. On Debian based systems (Ubuntu!) you can run the following:

```
sudo add-apt-repository ppa:openjdk-r/ppa
sudo apt-get update
sudo apt-get install openjdk-8-jdk
```

2. Now install the `minerl` package!:

```
pip3 install --upgrade minerl
```

**Note:** You may need the user flag: `pip3 install --upgrade minerl --user` to install properly.

---

## 1.2 Hello World: Your First Agent

With the `minerl` package installed on your system you can now make your first agent in Minecraft!

To get started, let's first import the necessary packages

```
import gym
import minerl
```

### 1.2.1 Creating an environment

Now we can choose any one of the [many environments](#) included in the `minerl` package. To learn more about the environments [checkout the environment documentation](#).

For this tutorial we'll choose the `MineRLNavigateDense-v0` environment. In this task, the agent is challenged with using a first-person perspective of a random Minecraft map and navigating to a target.

To create the environment, simply invoke `gym.make`

```
env = gym.make('MineRLNavigateDense-v0')
```

**Caution:** Currently `minerl` only supports environment rendering in **headed environments** (servers with monitors attached).

**In order to run `minerl` environments without a head use a software renderer such as `xvfb`:**

```
xvfb-run python3 <your_script.py>
```

Alternatively, you can use an environment variable which automatically adds `xvfb` when launching MineRL:

```
MINERL_HEADLESS=1 python3 <your_script.py>
```

**Note:** If you're worried and want to make sure something is happening behind the scenes install a logger **before** you create the environment.

```
import logging
logging.basicConfig(level=logging.DEBUG)

env = gym.make('MineRLNavigateDense-v0')
```



## 1.2.2 Taking actions

### As a warm up let's create a random agent.

Now we can reset this environment to its first position and get our first observation from the agent by resetting the environment.

**Note:** The first time you run this command to complete, it will take a while as it is recompiling Minecraft with the MineRL simulator mod (can be as long as 15-30 minutes)!

```
obs = env.reset()
```

The `obs` variable will be a dictionary containing the following observations returned by the environment. In the case of the `MineRLNavigate-v0` environment, three observations are returned: `pov`, an RGB image of the agent's first person perspective; `compassAngle`, a float giving the angle of the agent to its (approximate) target; and `inventory`, a dictionary containing the amount of 'dirt' blocks in the agent's inventory (this is useful for climbing steep inclines).

```
{
  'pov': array([[ 63,  63,  68],
               [ 63,  63,  68],
               [ 63,  63,  68],
               ...,
               [ 92,  92, 100],
               [ 92,  92, 100],
               [ 92,  92, 100]],
               ...),
  'compass': {'angle': array(-63.48639)},
  'inventory': {'dirt': 0}
}
```

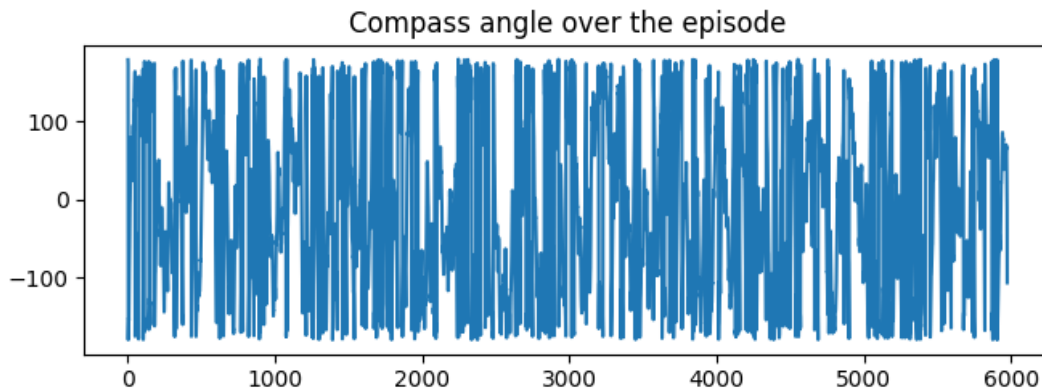
**Note:** To see the exact format of observations returned from and the exact action format expected by `env.step` for any environment refer to [the environment reference documentation](#)!

Now let's take actions through the environment until time runs out or the agent dies. To do this, we will use the normal OpenAI Gym `env.step` method.

```
done = False

while not done:
    action = env.action_space.sample()
    obs, reward, done, _ = env.step(action)
```

After running this code the agent should move sporadically until `done` flag is set to true. If you see a Minecraft window, it does not update while agent is playing, which is intended behaviour. To confirm that our agent is at least qualitatively acting randomly, on the right is a plot of the compass angle over the course of the experiment.



### 1.2.3 No-op actions and a better policy

Now let's make a hard-coded agent that actually runs towards the target.

To do this at every step of the environment we will take the *noop* action with a few modifications; in particular, we will only move forward, jump, attack, and change the agent's direction to minimize the angle between the agent's movement direction and it's target, `compassAngle`.

```
import minerl
import gym
env = gym.make('MineRLNavigateDense-v0')

obs = env.reset()
done = False
net_reward = 0

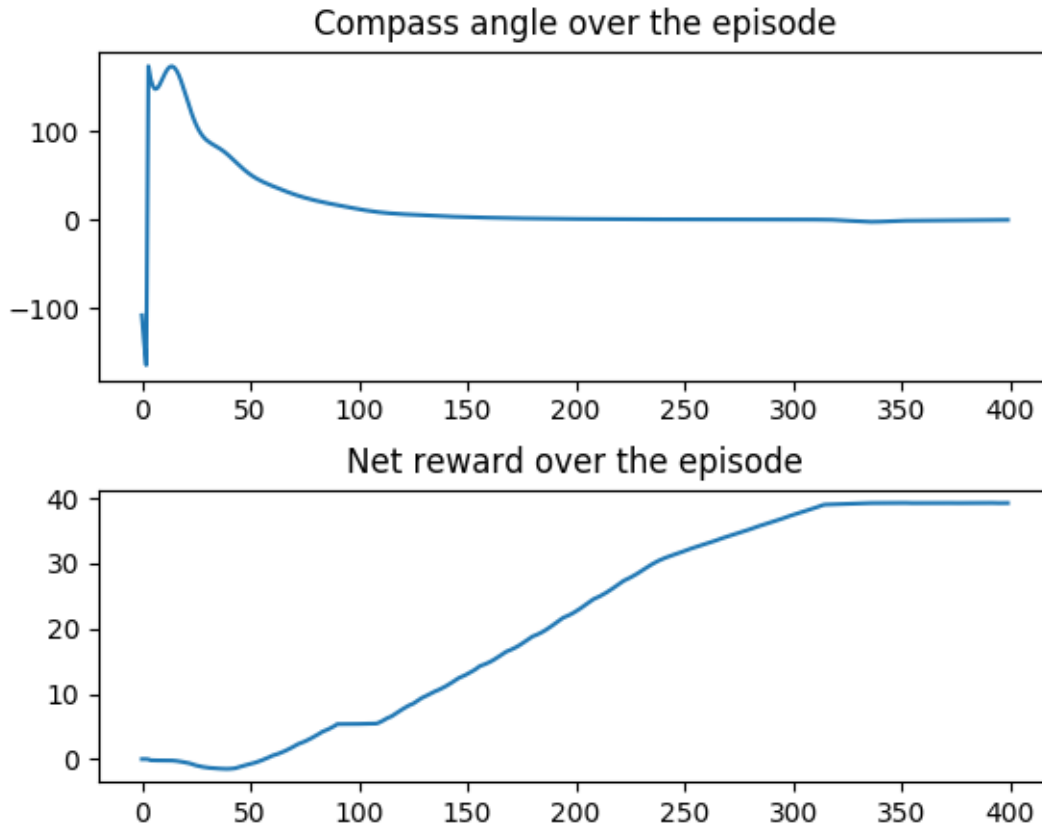
while not done:
    action = env.action_space.noop()

    action['camera'] = [0, 0.03*obs["compass"]["angle"]]
    action['back'] = 0
    action['forward'] = 1
    action['jump'] = 1
    action['attack'] = 1

    obs, reward, done, info = env.step(
        action)

    net_reward += reward
    print("Total reward: ", net_reward)
```

After running this agent, you should notice markedly less sporadic behaviour. Plotting both the `compassAngle` and the net reward over the episode confirm that this policy performs better than our random policy.



Congratulations! You've just made your first agent using the `minerl` framework!

## 1.3 Downloading and Sampling The Dataset

### 1.3.1 Introduction

Now that your agent can act in the environment, we should show it how to leverage human demonstrations.

To get started, let's download the minimal version of the dataset (two demonstrations from every environment). Since there are over 20 MineRL environments, this is still a sizeable download, at about 2 GB.

Then we will sample a few state-action-reward-done tuples from the `MineRLObtainDiamond-v0` dataset.

### 1.3.2 Setting up environment variables

The `minerl` package uses the `MINERL_DATA_ROOT` environment variable to locate the data directory. Please export `MINERL_DATA_ROOT=/your/local/path/`.

(Here are some tutorials on how to set environment variables on [Linux/Mac](#) and [Windows](#) computers.)

### 1.3.3 Downloading the MineRL Dataset with `minerl.data.download`

To download the minimal dataset into `MINERL_DATA_ROOT`, run the command:

```
python3 -m minerl.data.download
```

**Note:** The full dataset for a particular environment, or for a particular competition (Diamond or Basalt) can be downloaded using the `--environment ENV_NAME` and `--competition COMPETITION` flags.

`ENV_NAME` is any Gym environment name from the *documented environments*.

`COMPETITION` is `basalt` or `diamond`.

For more information, run `python3 -m minerl.data.download --help`.

As an example, to download the full dataset for “MineRLObtainDiamond-v0”, you can run

```
python3 -m minerl.data.download --environment "MineRLObtainDiamond-v0"
```

### 1.3.4 Sampling the Dataset with `buffered_batch_iter`

Now we can build the dataset for `MineRLObtainDiamond-v0`

There are two ways of sampling from the MineRL dataset: the deprecated but still supported `batch_iter`, and `buffered_batch_iter`. `batch_iter` is the legacy method, which we’ve kept in the code to avoid breaking changes, but we have recently realized that, when using `batch_size > 1`, `batch_iter` can fail to return a substantial portion of the data in the epoch.

**If you are not already using `data_pipeline.batch_iter`, we recommend against it, because of these issues**

The recommended way of sampling from the dataset is:

```
from minerl.data import BufferedBatchIter
data = minerl.data.make('MineRLObtainDiamond-v0')
iterator = BufferedBatchIter(data)
for current_state, action, reward, next_state, done \
    in iterator.buffered_batch_iter(batch_size=1, num_epochs=1):

    # Print the POV @ the first step of the sequence
    print(current_state['pov'][0])

    # Print the final reward pf the sequence!
    print(reward[-1])

    # Check if final (next_state) is terminal.
    print(done[-1])

    # ... do something with the data.
    print("At the end of trajectories the length"
          "can be < max_sequence_len", len(reward))
```

### 1.3.5 Moderate Human Demonstrations

MineRL-v0 uses community driven demonstrations to help researchers develop sample efficient techniques. Some of these demonstrations are less than optimal, however others could feature bugs with the client, server errors, or adversarial behavior.

Using the MineRL viewer, you can help curate this dataset by viewing these demonstrations manually and reporting bad streams by submitting an issue to github with the following information:

1. The stream name of the stream in question
2. The reason the stream or segment needs to be modified
3. The sample / frame number(s) (shown at the bottom of the viewer)

## 1.4 K-means exploration

With the 2020 MineRL competition, we introduced [vectorized obfuscated environments](#) which abstract non-visual state information as well as the action space of the agent to be continuous vector spaces. See [MineRLObtainDiamondVectorObf-v0](#) for documentation on the evaluation environment for that competition.

To use techniques in the MineRL competition that require discrete actions, we can use [k-means](#) to quantize the human demonstrations and give our agent n discrete actions representative of actions taken by humans when solving the environment.

To get started, let's download the [MineRLTreechopVectorObf-v0](#) environment.

```
python -m minerl.data.download 'MineRLTreechopVectorObf-v0'
```

**Note:** If you are unable to download the data ensure you have the `MINERL_DATA_ROOT` env variable set as demonstrated in [data sampling](#).

Now we load the dataset for [MineRLTreechopVectorObf-v0](#) and find 32 clusters using sklearn learn

```
from sklearn.cluster import KMeans

dat = minerl.data.make('MineRLTreechopVectorObf-v0')

# Load the dataset storing 1000 batches of actions
act_vectors = []
for _, act, _, _, _ in tqdm.tqdm(dat.batch_iter(16, 32, 2, preload_buffer_size=20)):
    act_vectors.append(act['vector'])
    if len(act_vectors) > 1000:
        break

# Reshape these the action batches
acts = np.concatenate(act_vectors).reshape(-1, 64)
kmeans_acts = acts[:100000]

# Use sklearn to cluster the demonstrated actions
kmeans = KMeans(n_clusters=32, random_state=0).fit(kmeans_acts)
```

Now we have 32 actions that represent reasonable actions for our agent to take. Let's take these and improve our random hello world agent from before.

```

i, net_reward, done, env = 0, 0, False, gym.make('MineRLTreechopVectorObf-v0')
obs = env.reset()

while not done:
    # Let's use a frame skip of 4 (could you do better than a hard-coded frame skip?)
    if i % 4 == 0:
        action = {
            'vector': kmeans.cluster_centers_[np.random.choice(NUM_CLUSTERS)]
        }

        obs, reward, done, info = env.step(action)
        env.render()

        if reward > 0:
            print("+{} reward!".format(reward))
            net_reward += reward
            i += 1

print("Total reward: ", net_reward)

```

Putting this all together we get:

#### Full snippet

```

import gym
import tqdm
import minerl
import numpy as np

from sklearn.cluster import KMeans

dat = minerl.data.make('MineRLTreechopVectorObf-v0')

act_vectors = []
NUM_CLUSTERS = 30

# Load the dataset storing 1000 batches of actions
for _, act, _, _, _ in tqdm.tqdm(dat.batch_iter(16, 32, 2, preload_buffer_size=20)):
    act_vectors.append(act['vector'])
    if len(act_vectors) > 1000:
        break

# Reshape these the action batches
acts = np.concatenate(act_vectors).reshape(-1, 64)
kmeans_acts = acts[:100000]

# Use sklearn to cluster the demonstrated actions
kmeans = KMeans(n_clusters=NUM_CLUSTERS, random_state=0).fit(kmeans_acts)

i, net_reward, done, env = 0, 0, False, gym.make('MineRLTreechopVectorObf-v0')
obs = env.reset()

```

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```

while not done:
    # Let's use a frame skip of 4 (could you do better than a hard-coded frame skip?)
    if i % 4 == 0:
        action = {
            'vector': kmeans.cluster_centers_[np.random.choice(NUM_CLUSTERS)]
        }

        obs, reward, done, info = env.step(action)
        env.render()

        if reward > 0:
            print("+{} reward!".format(reward))
            net_reward += reward
            i += 1

print("Total reward: ", net_reward)

```

Try comparing this k-means random agent with a random agent using `env.action_space.sample()`! You should see the human actions are a much more reasonable way to explore the environment!

## 1.5 Visualizing The Data `minerl.viewer`

To help you get familiar with the MineRL dataset, the `minerl` python package also provides a data trajectory viewer called `minerl.viewer`:

**Warning:** BASALT: `minerl.viewer` can load BASALT competition data, but is not yet updated to display the use or equip actions yet.

The `minerl.viewer` program lets you step through individual trajectories, showing the observation seen by the player, the action they took (including camera, movement, and any action described by an MineRL environment's action space), and the reward they received.

```

usage: python3 -m minerl.viewer [-h] environment [stream_name]

positional arguments:
  environment  The MineRL environment to visualize. e.g.
               MineRLObtainDiamondDense-v0
  stream_name  (optional) The name of the trajectory to visualize. e.g.
               v4_absolute_zucchini_basilisk-13_36805-50154.

optional arguments:
  -h, --help  show this help message and exit

```

Try it out on a random trajectory by running:

```

# Make sure your MINERL_DATA_ROOT is set!
export MINERL_DATA_ROOT='/your/local/path'

```

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```
# Visualizes a random trajectory of MineRLObtainDiamondDense-v0
python3 -m minerl.viewer MineRLObtainDiamondDense-v0
```

Try it out on a specific trajectory by running:

```
# Make sure your MINERL_DATA_ROOT is set!
export MINERL_DATA_ROOT='/your/local/path'
# Visualizes a specific trajectory. v4_absolute_zucch...
python3 -m minerl.viewer MineRLTreechop-v0 \
    v4_absolute_zucchini_basilisk-13_36805-50154
```

## 1.6 Interactive Mode `minerl.interactor`

**Warning:** The interactor does not function in MineRL version v0.4. If you wish to use this utility, install an older version of MineRL `pip install minerl=0.3.7`.

Once you have started training agents, the next step is getting them to interact with human players. To help achieve this, the `minerl` python package provides a interactive Minecraft client called `minerl.interactor`:

The `minerl.interactor` allows you to connect a human-controlled Minecraft client to the Minecraft world that your agent(s) is using and interact with the agent in real time.

**Note:** For observation-only mode hit the `t` key and type `/gamemode sp` to enter spectator mode and become invisible to your agent(s).

Enables human interaction with the environment.

To interact with the environment add `make_interactive` to your agent's evaluation code and then run the `minerl.interactor`.

For example:

```
env = gym.make('MineRL...')

# set the environment to allow interactive connections on port 6666
# and slow the tick speed to 6666.
env.make_interactive(port=6666, realtime=True)

# reset the env
env.reset()

# interact as normal.
...
```

Then while the agent is running, you can start the interactor with the following command.

```
python3 -m minerl.interactor 6666 # replace with the port above.
```



The interactor will disconnect when the mission resets, but you can connect again with the same command. If an interactor is already started, it won't need to be relaunched when running the command a second time.

## 1.7 General Information

The `minerl` package includes several environments as follows. This page describes each of the included environments, provides usage samples, and describes the exact action and observation space provided by each environment!

**Caution:** In the MineRL Competition, many environments are provided for training, however competition agents will only be evaluated in `MineRLObtainDiamondVectorObf-v0` which has **sparse** rewards. See [MineRLObtainDiamondVectorObf-v0](#).

**Note:** All environments offer a default no-op action via `env.action_space.no_op()` and a random action via `env.action_space.sample()`.

## 1.8 Environment Handlers

Minecraft is an extremely complex environment which provides players with visual, auditory, and informational observation of many complex data types. Furthermore, players interact with Minecraft using more than just embodied actions: players can craft, build, destroy, smelt, enchant, manage their inventory, and even communicate with other players via a text chat.

To provide a unified interface with which agents can obtain and perform similar observations and actions as players, we have provided first-class support for this multi-modality in the environment: **the observation and action spaces of environments are `gym.spaces.Dict` spaces**. These observation and action dictionaries are comprised of individual fields we call *handlers*.

**Note:** In the documentation of every environment we provide a listing of the exact `gym.space` of the observations returned by and actions expected by the environment's step function. We are slowly building documentation for these handlers, and **you can click those highlighted with blue** for more information!

### 1.8.1 Environment Handlers

Minecraft is an extremely complex environment which provides players with visual, auditory, and informational observation of many complex data types. Furthermore, players interact with Minecraft using more than just embodied actions: players can craft, build, destroy, smelt, enchant, manage their inventory, and even communicate with other players via a text chat.

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handlers, and **you can click those highlighted with blue** for more information!

---

## 1.8.2 Spaces

### Enum Spaces

Some observation and action spaces are Enum types. Examples include the *equip observation* and the *equip action*.

Observation and action spaces that are Enum are encoded as strings by default (e.g. “none”, “log”, and “sandstone#2”) when they are returned from `env.step()` and `env.reset()`, or yielded from `minerl.data.DataPipeline.batch_iter()`.

When building an action to pass into `env.step(act)`, the Enum component of the action dict can be encoded as either a string or an integer.

---

**Tip:** The Enum integer value that corresponds to each Enum string value can be accessed via `Enum.values_map[string_value]`. For example, to get the integer value corresponding to the equip action “dirt” in `MineRLObtainDiamond` or `MineRLBasaltBuildVillageHouse`, you can call `env.action_space.spaces["equip"].values_map["dirt"]`.

---

## 1.8.3 Observations

### Visual Observations - pov, third-person

**pov** : `Box(width, height, nchannels)`

An RGB image observation of the agent’s first-person perspective.

**Type** `np.uint8`

**third-person** : `Box(width, height, nchannels)`

An RGB image observation of the agent’s third-person perspective.

**Warning:** This observation is not yet supported by any environment.

**Type** `np.uint8`

**compass-observation** : `Box(1)`

The current position of the *minecraft:compass* object from 0 (behind agent left) to 0.5 in front of agent to 1 (behind agent right)

---

**Note:** This observation uses the default Minecraft game logic which includes compass needle momentum. As such it may change even when the agent has stopped moving!

---

## Equip Observations - equipped\_items

**equipped\_items.mainhand.type** : Enum('none', 'air', ..., 'other'))

This observation is an Enum type. See *Enum Spaces* for more information.

The type of the item that the player has equipped in the mainhand slot. If the mainhand slot is empty then the value is 'air'. If the mainhand slot contains an item not inside this observation space, then the value is 'other'.

**Type** np.int64

**Shape** [1]

## 1.8.4 Actions

### Camera Control - camera

**camera** : Box(2) [delta\_pitch, delta\_yaw]

This action changes the orientation of the agent's head by the corresponding number of degrees. When the pov observation is available, the camera changes its orientation pitch by the first component and its yaw by the second component. Both delta\_pitch and delta\_yaw are limited to [-180, 180] inclusive

**Type** np.float32

**Shape** [2]

**attack** : Discrete(1) [attack]

This action causes the agent to attack.

**Type** np.float32

**Shape** [1]

### Tool Control - equip and use

**equip** : Enum('none', 'air', ..., 'other'))

This action is an Enum type. See *Enum Spaces* for more information.

This action equips the first instance of the specified item from the agent's inventory to the main hand if the specified item is present, otherwise does nothing. `air` matches any empty slot in an agent's inventory and functions as an un-equip, or equip-nothing action.

**Type** np.int64

**Shape** [1]

---

**Note:** `equip 'none'` and `equip 'other'` are both no-op actions. In other words, they leave the currently equipped item unchanged. However, in the MineRL dataset, `other` takes on a special meaning. `other` is the wildcard equip action that is recorded in the dataset whenever a player equipped an item that wasn't included in this action space's Enum.

---

**Warning:** `env.step(act)` typically will not process the equip action for two ticks (i.e., you will not see the observation value `equipped_items` change until two more calls to `env.step`.)

This is due to a limitation with the current version of Malmo, our Minecraft backend.

**use** : `Discrete(1)` [`use`]

This action is equivalent to right-clicking in Minecraft. It causes the agent to use the item it is holding in the *mainhand slot*, or to open doors or gates when it is facing an applicable Minecraft structure.

**Type** `np.int64`

**Shape** `[1]`

## 1.9 Basic Environments

**Warning:** The following Basic Environments are NOT part of the MineRL Diamond and BASALT competitions! Feel free to use them for personal exploration, but note that competitions agents may only be trained on their corresponding competition environments.

### 1.9.1 MineRLTreechop-v0

In treechop, the agent must collect 64 *minecraft:log*. This replicates a common scenario in Minecraft, as logs are necessary to craft a large amount of items in the game and are a key resource in Minecraft.

The agent begins in a forest biome (near many trees) with an iron axe for cutting trees. The agent is given +1 reward for obtaining each unit of wood, and the episode terminates once the agent obtains 64 units.

#### Observation Space

```
Dict({
  "pov": "Box(low=0, high=255, shape=(64, 64, 3))"
})
```

#### Action Space

```
Dict({
  "attack": "Discrete(2)",
  "back": "Discrete(2)",
  "camera": "Box(low=-180.0, high=180.0, shape=(2,))",
  "forward": "Discrete(2)",
  "jump": "Discrete(2)",
  "left": "Discrete(2)",
  "right": "Discrete(2)",
  "sneak": "Discrete(2)",
  "sprint": "Discrete(2)"
})
```

## Usage

```
import gym
import minerl

# Run a random agent through the environment
env = gym.make("MineRLTreechop-v0") # A MineRLTreechop-v0 env

obs = env.reset()
done = False

while not done:
    # Take a no-op through the environment.
    obs, rew, done, _ = env.step(env.action_space.noop())
    # Do something

#####

# Sample some data from the dataset!
data = minerl.data.make("MineRLTreechop-v0")

# Iterate through a single epoch using sequences of at most 32 steps
for obs, rew, done, act in data.batch_iter(num_epochs=1, batch_size=32):
    # Do something
```

### 1.9.2 MineRLNavigate-v0

In this task, the agent must move to a goal location denoted by a diamond block. This represents a basic primitive used in many tasks throughout Minecraft. In addition to standard observations, the agent has access to a “compass” observation, which points near the goal location, 64 meters from the start location. The goal has a small random horizontal offset from the compass location and may be slightly below surface level. On the goal location is a unique block, so the agent must find the final goal by searching based on local visual features.

The agent is given a sparse reward (+100 upon reaching the goal, at which point the episode terminates). **This variant of the environment is sparse.**

In this environment, the agent spawns on a random survival map.

## Observation Space

```
Dict({
  "compass": {
    "angle": "Box(low=-180.0, high=180.0, shape=())"
  },
  "inventory": {
    "dirt": "Box(low=0, high=2304, shape=())"
  },
  "pov": "Box(low=0, high=255, shape=(64, 64, 3))"
})
```

## Action Space

```
Dict({
  "attack": "Discrete(2)",
  "back": "Discrete(2)",
  "camera": "Box(low=-180.0, high=180.0, shape=(2,))",
  "forward": "Discrete(2)",
  "jump": "Discrete(2)",
  "left": "Discrete(2)",
  "place": "Enum(dirt,none)",
  "right": "Discrete(2)",
  "sneak": "Discrete(2)",
  "sprint": "Discrete(2)"
})
```

## Usage

```
import gym
import minerl

# Run a random agent through the environment
env = gym.make("MineRLNavigate-v0") # A MineRLNavigate-v0 env

obs = env.reset()
done = False

while not done:
    # Take a no-op through the environment.
    obs, rew, done, _ = env.step(env.action_space.noop())
    # Do something

#####

# Sample some data from the dataset!
data = minerl.data.make("MineRLNavigate-v0")

# Iterate through a single epoch using sequences of at most 32 steps
for obs, rew, done, act in data.batch_iter(num_epochs=1, batch_size=32):
    # Do something
```

### 1.9.3 MineRLNavigateDense-v0

In this task, the agent must move to a goal location denoted by a diamond block. This represents a basic primitive used in many tasks throughout Minecraft. In addition to standard observations, the agent has access to a “compass” observation, which points near the goal location, 64 meters from the start location. The goal has a small random horizontal offset from the compass location and may be slightly below surface level. On the goal location is a unique block, so the agent must find the final goal by searching based on local visual features.

The agent is given a sparse reward (+100 upon reaching the goal, at which point the episode terminates). **This variant of the environment is dense reward-shaped where the agent is given a reward every tick for how much closer (or negative reward for farther) the agent gets to the target.**

In this environment, the agent spawns on a random survival map.

#### Observation Space

```
Dict({
  "compass": {
    "angle": "Box(low=-180.0, high=180.0, shape=())"
  },
  "inventory": {
    "dirt": "Box(low=0, high=2304, shape=())"
  },
  "pov": "Box(low=0, high=255, shape=(64, 64, 3))"
})
```

#### Action Space

```
Dict({
  "attack": "Discrete(2)",
  "back": "Discrete(2)",
  "camera": "Box(low=-180.0, high=180.0, shape=(2,))",
  "forward": "Discrete(2)",
  "jump": "Discrete(2)",
  "left": "Discrete(2)",
  "place": "Enum(dirt,none)",
  "right": "Discrete(2)",
  "sneak": "Discrete(2)",
  "sprint": "Discrete(2)"
})
```

## Usage

```
import gym
import minerl

# Run a random agent through the environment
env = gym.make("MineRLNavigateDense-v0") # A MineRLNavigateDense-v0 env

obs = env.reset()
done = False

while not done:
    # Take a no-op through the environment.
    obs, rew, done, _ = env.step(env.action_space.noop())
    # Do something

#####

# Sample some data from the dataset!
data = minerl.data.make("MineRLNavigateDense-v0")

# Iterate through a single epoch using sequences of at most 32 steps
for obs, rew, done, act in data.batch_iter(num_epochs=1, batch_size=32):
    # Do something
```

### 1.9.4 MineRLNavigateExtreme-v0

In this task, the agent must move to a goal location denoted by a diamond block. This represents a basic primitive used in many tasks throughout Minecraft. In addition to standard observations, the agent has access to a “compass” observation, which points near the goal location, 64 meters from the start location. The goal has a small random horizontal offset from the compass location and may be slightly below surface level. On the goal location is a unique block, so the agent must find the final goal by searching based on local visual features.

The agent is given a sparse reward (+100 upon reaching the goal, at which point the episode terminates). **This variant of the environment is sparse.**

In this environment, the agent spawns in an extreme hills biome.



## Observation Space

```
Dict({
  "compass": {
    "angle": "Box(low=-180.0, high=180.0, shape=())"
  },
  "inventory": {
    "dirt": "Box(low=0, high=2304, shape=())"
  },
  "pov": "Box(low=0, high=255, shape=(64, 64, 3))"
})
```

## Action Space

```
Dict({
  "attack": "Discrete(2)",
  "back": "Discrete(2)",
  "camera": "Box(low=-180.0, high=180.0, shape=(2,))",
  "forward": "Discrete(2)",
  "jump": "Discrete(2)",
  "left": "Discrete(2)",
  "place": "Enum(dirt,none)",
  "right": "Discrete(2)",
  "sneak": "Discrete(2)",
  "sprint": "Discrete(2)"
})
```

## Usage

```
import gym
import minerl

# Run a random agent through the environment
env = gym.make("MineRLNavigateExtreme-v0") # A MineRLNavigateExtreme-v0 env

obs = env.reset()
done = False

while not done:
    # Take a no-op through the environment.
    obs, rew, done, _ = env.step(env.action_space.noop())
    # Do something

#####

# Sample some data from the dataset!
data = minerl.data.make("MineRLNavigateExtreme-v0")

# Iterate through a single epoch using sequences of at most 32 steps
for obs, rew, done, act in data.batch_iter(num_epochs=1, batch_size=32):
    # Do something
```

### 1.9.5 MineRLNavigateExtremeDense-v0

In this task, the agent must move to a goal location denoted by a diamond block. This represents a basic primitive used in many tasks throughout Minecraft. In addition to standard observations, the agent has access to a “compass” observation, which points near the goal location, 64 meters from the start location. The goal has a small random horizontal offset from the compass location and may be slightly below surface level. On the goal location is a unique block, so the agent must find the final goal by searching based on local visual features.

The agent is given a sparse reward (+100 upon reaching the goal, at which point the episode terminates). **This variant of the environment is dense reward-shaped where the agent is given a reward every tick for how much closer (or negative reward for farther) the agent gets to the target.**

In this environment, the agent spawns in an extreme hills biome.

#### Observation Space

```
Dict({
  "compass": {
    "angle": "Box(low=-180.0, high=180.0, shape=())"
  },
  "inventory": {
    "dirt": "Box(low=0, high=2304, shape=())"
  },
  "pov": "Box(low=0, high=255, shape=(64, 64, 3))"
})
```

#### Action Space

```
Dict({
  "attack": "Discrete(2)",
  "back": "Discrete(2)",
  "camera": "Box(low=-180.0, high=180.0, shape=(2,))",
  "forward": "Discrete(2)",
  "jump": "Discrete(2)",
  "left": "Discrete(2)",
  "place": "Enum(dirt,none)",
  "right": "Discrete(2)",
  "sneak": "Discrete(2)",
  "sprint": "Discrete(2)"
})
```

## Usage

```

import gym
import minerl

# Run a random agent through the environment
env = gym.make("MineRLNavigateExtremeDense-v0") # A MineRLNavigateExtremeDense-v0 env

obs = env.reset()
done = False

while not done:
    # Take a no-op through the environment.
    obs, rew, done, _ = env.step(env.action_space.noop())
    # Do something

#####

# Sample some data from the dataset!
data = minerl.data.make("MineRLNavigateExtremeDense-v0")

# Iterate through a single epoch using sequences of at most 32 steps
for obs, rew, done, act in data.batch_iter(num_epochs=1, batch_size=32):
    # Do something

```

### 1.9.6 MineRLObtainDiamond-v0

In this environment the agent is required to obtain a diamond. The agent begins in a random starting location on a random survival map without any items, matching the normal starting conditions for human players in Minecraft. The agent is given access to a selected summary of its inventory and GUI free crafting, smelting, and inventory management actions.

During an episode the agent is rewarded **every** time it obtains an item in the requisite item hierarchy to obtaining a diamond. The rewards for each item are given here:

```

<Item reward="1" type="log" />
<Item reward="2" type="planks" />
<Item reward="4" type="stick" />
<Item reward="4" type="crafting_table" />
<Item reward="8" type="wooden_pickaxe" />
<Item reward="16" type="cobblestone" />
<Item reward="32" type="furnace" />
<Item reward="32" type="stone_pickaxe" />
<Item reward="64" type="iron_ore" />
<Item reward="128" type="iron_ingot" />
<Item reward="256" type="iron_pickaxe" />
<Item reward="1024" type="diamond" />

```

## Observation Space

```
Dict({
  "equipped_items": {
    "mainhand": {
      "damage": "Box(low=-1, high=1562, shape=())",
      "maxDamage": "Box(low=-1, high=1562, shape=())",
      "type": "Enum(air,iron_axe,iron_pickaxe,none,other,stone_axe,stone_
↪pickaxe,wooden_axe,wooden_pickaxe)"
    }
  },
  "inventory": {
    "coal": "Box(low=0, high=2304, shape=())",
    "cobblestone": "Box(low=0, high=2304, shape=())",
    "crafting_table": "Box(low=0, high=2304, shape=())",
    "dirt": "Box(low=0, high=2304, shape=())",
    "furnace": "Box(low=0, high=2304, shape=())",
    "iron_axe": "Box(low=0, high=2304, shape=())",
    "iron_ingot": "Box(low=0, high=2304, shape=())",
    "iron_ore": "Box(low=0, high=2304, shape=())",
    "iron_pickaxe": "Box(low=0, high=2304, shape=())",
    "log": "Box(low=0, high=2304, shape=())",
    "planks": "Box(low=0, high=2304, shape=())",
    "stick": "Box(low=0, high=2304, shape=())",
    "stone": "Box(low=0, high=2304, shape=())",
    "stone_axe": "Box(low=0, high=2304, shape=())",
    "stone_pickaxe": "Box(low=0, high=2304, shape=())",
    "torch": "Box(low=0, high=2304, shape=())",
    "wooden_axe": "Box(low=0, high=2304, shape=())",
    "wooden_pickaxe": "Box(low=0, high=2304, shape=())"
  },
  "pov": "Box(low=0, high=255, shape=(64, 64, 3))"
})
```

## Action Space

```
Dict({
  "attack": "Discrete(2)",
  "back": "Discrete(2)",
  "camera": "Box(low=-180.0, high=180.0, shape=(2,))",
  "craft": "Enum(crafting_table,none,planks,stick,torch)",
  "equip": "Enum(air,iron_axe,iron_pickaxe,none,stone_axe,stone_pickaxe,wooden_axe,
↪wooden_pickaxe)",
  "forward": "Discrete(2)",
  "jump": "Discrete(2)",
  "left": "Discrete(2)",
  "nearbyCraft": "Enum(furnace,iron_axe,iron_pickaxe,none,stone_axe,stone_pickaxe,
↪wooden_axe,wooden_pickaxe)",
  "nearbySmelt": "Enum(coal,iron_ingot,none)",
  "place": "Enum(cobblestone,crafting_table,dirt,furnace,none,stone,torch)",
  "right": "Discrete(2)",
  "sneak": "Discrete(2)",
  "sprint": "Discrete(2)"
})
```

```
})
```

## Usage

```
import gym
import minerl

# Run a random agent through the environment
env = gym.make("MineRLObtainDiamond-v0") # A MineRLObtainDiamond-v0 env

obs = env.reset()
done = False

while not done:
    # Take a no-op through the environment.
    obs, rew, done, _ = env.step(env.action_space.noop())
    # Do something

#####

# Sample some data from the dataset!
data = minerl.data.make("MineRLObtainDiamond-v0")

# Iterate through a single epoch using sequences of at most 32 steps
for obs, rew, done, act in data.batch_iter(num_epochs=1, batch_size=32):
    # Do something
```

### 1.9.7 MineRLObtainDiamondDense-v0

In this environment the agent is required to obtain a diamond. The agent begins in a random starting location on a random survival map without any items, matching the normal starting conditions for human players in Minecraft. The agent is given access to a selected summary of its inventory and GUI free crafting, smelting, and inventory management actions.

During an episode the agent is rewarded **every** time it obtains an item in the requisite item hierarchy to obtaining a diamond. The rewards for each item are given here:

```
<Item reward="1" type="log" />
<Item reward="2" type="planks" />
<Item reward="4" type="stick" />
<Item reward="4" type="crafting_table" />
<Item reward="8" type="wooden_pickaxe" />
<Item reward="16" type="cobblestone" />
<Item reward="32" type="furnace" />
<Item reward="32" type="stone_pickaxe" />
```

(continues on next page)

```
<Item reward="64" type="iron_ore" />
<Item reward="128" type="iron_ingot" />
<Item reward="256" type="iron_pickaxe" />
<Item reward="1024" type="diamond" />
```

## Observation Space

```
Dict({
  "equipped_items": {
    "mainhand": {
      "damage": "Box(low=-1, high=1562, shape=())",
      "maxDamage": "Box(low=-1, high=1562, shape=())",
      "type": "Enum(air, iron_axe, iron_pickaxe, none, other, stone_axe, stone_
↪pickaxe, wooden_axe, wooden_pickaxe)"
    }
  },
  "inventory": {
    "coal": "Box(low=0, high=2304, shape=())",
    "cobblestone": "Box(low=0, high=2304, shape=())",
    "crafting_table": "Box(low=0, high=2304, shape=())",
    "dirt": "Box(low=0, high=2304, shape=())",
    "furnace": "Box(low=0, high=2304, shape=())",
    "iron_axe": "Box(low=0, high=2304, shape=())",
    "iron_ingot": "Box(low=0, high=2304, shape=())",
    "iron_ore": "Box(low=0, high=2304, shape=())",
    "iron_pickaxe": "Box(low=0, high=2304, shape=())",
    "log": "Box(low=0, high=2304, shape=())",
    "planks": "Box(low=0, high=2304, shape=())",
    "stick": "Box(low=0, high=2304, shape=())",
    "stone": "Box(low=0, high=2304, shape=())",
    "stone_axe": "Box(low=0, high=2304, shape=())",
    "stone_pickaxe": "Box(low=0, high=2304, shape=())",
    "torch": "Box(low=0, high=2304, shape=())",
    "wooden_axe": "Box(low=0, high=2304, shape=())",
    "wooden_pickaxe": "Box(low=0, high=2304, shape=())"
  },
  "pov": "Box(low=0, high=255, shape=(64, 64, 3))"
})
```

## Action Space

```
Dict({
  "attack": "Discrete(2)",
  "back": "Discrete(2)",
  "camera": "Box(low=-180.0, high=180.0, shape=(2,))",
  "craft": "Enum(crafting_table, none, planks, stick, torch)",
  "equip": "Enum(air, iron_axe, iron_pickaxe, none, stone_axe, stone_pickaxe, wooden_axe,
↪wooden_pickaxe)",
  "forward": "Discrete(2)",
  "jump": "Discrete(2)",
  "left": "Discrete(2)",
```

```

    "nearbyCraft": "Enum(furnace,iron_axe,iron_pickaxe,none,stone_axe,stone_pickaxe,
↪ wooden_axe,wooden_pickaxe)",
    "nearbySmelt": "Enum(coal,iron_ingot,none)",
    "place": "Enum(cobblestone,crafting_table,dirt,furnace,none,stone,torch)",
    "right": "Discrete(2)",
    "sneak": "Discrete(2)",
    "sprint": "Discrete(2)"
})

```

## Usage

```

import gym
import minerl

# Run a random agent through the environment
env = gym.make("MineRLObtainDiamondDense-v0") # A MineRLObtainDiamondDense-v0 env

obs = env.reset()
done = False

while not done:
    # Take a no-op through the environment.
    obs, rew, done, _ = env.step(env.action_space.noop())
    # Do something

#####

# Sample some data from the dataset!
data = minerl.data.make("MineRLObtainDiamondDense-v0")

# Iterate through a single epoch using sequences of at most 32 steps
for obs, rew, done, act in data.batch_iter(num_epochs=1, batch_size=32):
    # Do something

```

### 1.9.8 MineRLObtainIronPickaxe-v0

In this environment the agent is required to obtain an iron pickaxe. The agent begins in a random starting location, on a random survival map, without any items, matching the normal starting conditions for human players in Minecraft. The agent is given access to a selected view of its inventory and GUI free crafting, smelting, and inventory management actions.

During an episode **the agent is rewarded only once per item the first time it obtains that item in the requisite item hierarchy for obtaining an iron pickaxe.** The reward for each item is given here:

```

<Item amount="1" reward="1" type="log" />
<Item amount="1" reward="2" type="planks" />
<Item amount="1" reward="4" type="stick" />
<Item amount="1" reward="4" type="crafting_table" />
<Item amount="1" reward="8" type="wooden_pickaxe" />
<Item amount="1" reward="16" type="cobblestone" />
<Item amount="1" reward="32" type="furnace" />
<Item amount="1" reward="32" type="stone_pickaxe" />
<Item amount="1" reward="64" type="iron_ore" />
<Item amount="1" reward="128" type="iron_ingot" />
<Item amount="1" reward="256" type="iron_pickaxe" />

```

## Observation Space

```

Dict({
  "equipped_items": {
    "mainhand": {
      "damage": "Box(low=-1, high=1562, shape=())",
      "maxDamage": "Box(low=-1, high=1562, shape=())",
      "type": "Enum(air,iron_axe,iron_pickaxe,none,other,stone_axe,stone_
↪pickaxe,wooden_axe,wooden_pickaxe)"
    }
  },
  "inventory": {
    "coal": "Box(low=0, high=2304, shape=())",
    "cobblestone": "Box(low=0, high=2304, shape=())",
    "crafting_table": "Box(low=0, high=2304, shape=())",
    "dirt": "Box(low=0, high=2304, shape=())",
    "furnace": "Box(low=0, high=2304, shape=())",
    "iron_axe": "Box(low=0, high=2304, shape=())",
    "iron_ingot": "Box(low=0, high=2304, shape=())",
    "iron_ore": "Box(low=0, high=2304, shape=())",
    "iron_pickaxe": "Box(low=0, high=2304, shape=())",
    "log": "Box(low=0, high=2304, shape=())",
    "planks": "Box(low=0, high=2304, shape=())",
    "stick": "Box(low=0, high=2304, shape=())",
    "stone": "Box(low=0, high=2304, shape=())",
    "stone_axe": "Box(low=0, high=2304, shape=())",
    "stone_pickaxe": "Box(low=0, high=2304, shape=())",
    "torch": "Box(low=0, high=2304, shape=())",
    "wooden_axe": "Box(low=0, high=2304, shape=())",
    "wooden_pickaxe": "Box(low=0, high=2304, shape=())"
  },
  "pov": "Box(low=0, high=255, shape=(64, 64, 3))"
})

```



## Action Space

```
Dict({
  "attack": "Discrete(2)",
  "back": "Discrete(2)",
  "camera": "Box(low=-180.0, high=180.0, shape=(2,))",
  "craft": "Enum(crafting_table,none,planks,stick,torch)",
  "equip": "Enum(air,iron_axe,iron_pickaxe,none,stone_axe,stone_pickaxe,wooden_axe,
↪wooden_pickaxe)",
  "forward": "Discrete(2)",
  "jump": "Discrete(2)",
  "left": "Discrete(2)",
  "nearbyCraft": "Enum(furnace,iron_axe,iron_pickaxe,none,stone_axe,stone_pickaxe,
↪wooden_axe,wooden_pickaxe)",
  "nearbySmelt": "Enum(coal,iron_ingot,none)",
  "place": "Enum(cobblestone,crafting_table,dirt,furnace,none,stone,torch)",
  "right": "Discrete(2)",
  "sneak": "Discrete(2)",
  "sprint": "Discrete(2)"
})
```

## Usage

```
import gym
import minerl

# Run a random agent through the environment
env = gym.make("MineRLObtainIronPickaxe-v0") # A MineRLObtainIronPickaxe-v0 env

obs = env.reset()
done = False

while not done:
    # Take a no-op through the environment.
    obs, rew, done, _ = env.step(env.action_space.noop())
    # Do something

#####

# Sample some data from the dataset!
data = minerl.data.make("MineRLObtainIronPickaxe-v0")

# Iterate through a single epoch using sequences of at most 32 steps
for obs, rew, done, act in data.batch_iter(num_epochs=1, batch_size=32):
    # Do something
```

### 1.9.9 MineRLObtainIronPickaxeDense-v0

In this environment the agent is required to obtain an iron pickaxe. The agent begins in a random starting location, on a random survival map, without any items, matching the normal starting conditions for human players in Minecraft. The agent is given access to a selected view of its inventory and GUI free crafting, smelting, and inventory management actions.

During an episode **the agent is rewarded only once per item the first time it obtains that item in the requisite item hierarchy for obtaining an iron pickaxe.** The reward for each item is given here:

```
<Item amount="1" reward="1" type="log" />
<Item amount="1" reward="2" type="planks" />
<Item amount="1" reward="4" type="stick" />
<Item amount="1" reward="4" type="crafting_table" />
<Item amount="1" reward="8" type="wooden_pickaxe" />
<Item amount="1" reward="16" type="cobblestone" />
<Item amount="1" reward="32" type="furnace" />
<Item amount="1" reward="32" type="stone_pickaxe" />
<Item amount="1" reward="64" type="iron_ore" />
<Item amount="1" reward="128" type="iron_ingot" />
<Item amount="1" reward="256" type="iron_pickaxe" />
```

#### Observation Space

```
Dict({
  "equipped_items": {
    "mainhand": {
      "damage": "Box(low=-1, high=1562, shape=())",
      "maxDamage": "Box(low=-1, high=1562, shape=())",
      "type": "Enum(air,iron_axe,iron_pickaxe,none,other,stone_axe,stone_
↪pickaxe,wooden_axe,wooden_pickaxe)"
    }
  },
  "inventory": {
    "coal": "Box(low=0, high=2304, shape=())",
    "cobblestone": "Box(low=0, high=2304, shape=())",
    "crafting_table": "Box(low=0, high=2304, shape=())",
    "dirt": "Box(low=0, high=2304, shape=())",
    "furnace": "Box(low=0, high=2304, shape=())",
    "iron_axe": "Box(low=0, high=2304, shape=())",
    "iron_ingot": "Box(low=0, high=2304, shape=())",
    "iron_ore": "Box(low=0, high=2304, shape=())",
    "iron_pickaxe": "Box(low=0, high=2304, shape=())",
    "log": "Box(low=0, high=2304, shape=())",
    "planks": "Box(low=0, high=2304, shape=())",
    "stick": "Box(low=0, high=2304, shape=())",
    "stone": "Box(low=0, high=2304, shape=())",
```

```

    "stone_axe": "Box(low=0, high=2304, shape=())",
    "stone_pickaxe": "Box(low=0, high=2304, shape=())",
    "torch": "Box(low=0, high=2304, shape=())",
    "wooden_axe": "Box(low=0, high=2304, shape=())",
    "wooden_pickaxe": "Box(low=0, high=2304, shape=())"
  },
  "pov": "Box(low=0, high=255, shape=(64, 64, 3))"
})

```

## Action Space

```

Dict({
  "attack": "Discrete(2)",
  "back": "Discrete(2)",
  "camera": "Box(low=-180.0, high=180.0, shape=(2,))",
  "craft": "Enum(crafting_table,none,planks,stick,torch)",
  "equip": "Enum(air,iron_axe,iron_pickaxe,none,stone_axe,stone_pickaxe,wooden_axe,
↳ wooden_pickaxe)",
  "forward": "Discrete(2)",
  "jump": "Discrete(2)",
  "left": "Discrete(2)",
  "nearbyCraft": "Enum(furnace,iron_axe,iron_pickaxe,none,stone_axe,stone_pickaxe,
↳ wooden_axe,wooden_pickaxe)",
  "nearbySmelt": "Enum(coal,iron_ingot,none)",
  "place": "Enum(cobblestone,crafting_table,dirt,furnace,none,stone,torch)",
  "right": "Discrete(2)",
  "sneak": "Discrete(2)",
  "sprint": "Discrete(2)"
})

```

## Usage

```

import gym
import minerl

# Run a random agent through the environment
env = gym.make("MineRLObtainIronPickaxeDense-v0") # A MineRLObtainIronPickaxeDense-v0
↳ env

obs = env.reset()
done = False

while not done:
    # Take a no-op through the environment.
    obs, rew, done, _ = env.step(env.action_space.noop())
    # Do something

#####

# Sample some data from the dataset!
data = minerl.data.make("MineRLObtainIronPickaxeDense-v0")

```

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```
# Iterate through a single epoch using sequences of at most 32 steps
for obs, rew, done, act in data.batch_iter(num_epochs=1, batch_size=32):
    # Do something
```

## 1.10 MineRL Diamond Competition Environments

### 1.10.1 MineRLTreechopVectorObf-v0

In treechop, the agent must collect 64 *minecraft:log*. This replicates a common scenario in Minecraft, as logs are necessary to craft a large amount of items in the game and are a key resource in Minecraft.

The agent begins in a forest biome (near many trees) with an iron axe for cutting trees. The agent is given +1 reward for obtaining each unit of wood, and the episode terminates once the agent obtains 64 units.

#### Observation Space

```
Dict({
    "pov": "Box(low=0, high=255, shape=(64, 64, 3))",
    "vector": "Box(low=-1.20000000476837158, high=1.20000000476837158, shape=(64,))"
})
```

#### Action Space

```
Dict({
    "vector": "Box(low=-1.0499999523162842, high=1.0499999523162842, shape=(64,))"
})
```

#### Usage

```
import gym
import minerl

# Run a random agent through the environment
env = gym.make("MineRLTreechopVectorObf-v0") # A MineRLTreechopVectorObf-v0 env

obs = env.reset()
done = False

while not done:
    # Take a no-op through the environment.
    obs, rew, done, _ = env.step(env.action_space.noop())
```

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```

# Do something

#####

# Sample some data from the dataset!
data = minerl.data.make("MineRLTreechopVectorObf-v0")

# Iterate through a single epoch using sequences of at most 32 steps
for obs, rew, done, act in data.batch_iter(num_epochs=1, batch_size=32):
    # Do something

```

### 1.10.2 MineRLNavigateVectorObf-v0

In this task, the agent must move to a goal location denoted by a diamond block. This represents a basic primitive used in many tasks throughout Minecraft. In addition to standard observations, the agent has access to a “compass” observation, which points near the goal location, 64 meters from the start location. The goal has a small random horizontal offset from the compass location and may be slightly below surface level. On the goal location is a unique block, so the agent must find the final goal by searching based on local visual features.

The agent is given a sparse reward (+100 upon reaching the goal, at which point the episode terminates). **This variant of the environment is sparse.**

In this environment, the agent spawns on a random survival map.

#### Observation Space

```

Dict({
    "pov": "Box(low=0, high=255, shape=(64, 64, 3))",
    "vector": "Box(low=-1.2000000476837158, high=1.2000000476837158, shape=(64,))"
})

```

#### Action Space

```

Dict({
    "vector": "Box(low=-1.0499999523162842, high=1.0499999523162842, shape=(64,))"
})

```

## Usage

```
import gym
import minerl

# Run a random agent through the environment
env = gym.make("MineRLNavigateVectorObf-v0") # A MineRLNavigateVectorObf-v0 env

obs = env.reset()
done = False

while not done:
    # Take a no-op through the environment.
    obs, rew, done, _ = env.step(env.action_space.noop())
    # Do something

#####

# Sample some data from the dataset!
data = minerl.data.make("MineRLNavigateVectorObf-v0")

# Iterate through a single epoch using sequences of at most 32 steps
for obs, rew, done, act in data.batch_iter(num_epochs=1, batch_size=32):
    # Do something
```

### 1.10.3 MineRLNavigateDenseVectorObf-v0

In this task, the agent must move to a goal location denoted by a diamond block. This represents a basic primitive used in many tasks throughout Minecraft. In addition to standard observations, the agent has access to a “compass” observation, which points near the goal location, 64 meters from the start location. The goal has a small random horizontal offset from the compass location and may be slightly below surface level. On the goal location is a unique block, so the agent must find the final goal by searching based on local visual features.

The agent is given a sparse reward (+100 upon reaching the goal, at which point the episode terminates). **This variant of the environment is dense reward-shaped where the agent is given a reward every tick for how much closer (or negative reward for farther) the agent gets to the target.**

In this environment, the agent spawns on a random survival map.

## Observation Space

```
Dict({
  "pov": "Box(low=0, high=255, shape=(64, 64, 3))",
  "vector": "Box(low=-1.20000000476837158, high=1.20000000476837158, shape=(64,))"
})
```

## Action Space

```
Dict({
  "vector": "Box(low=-1.0499999523162842, high=1.0499999523162842, shape=(64,))"
})
```

## Usage

```
import gym
import minerl

# Run a random agent through the environment
env = gym.make("MineRLNavigateDenseVectorObf-v0") # A MineRLNavigateDenseVectorObf-v0
↳ env

obs = env.reset()
done = False

while not done:
    # Take a no-op through the environment.
    obs, rew, done, _ = env.step(env.action_space.noop())
    # Do something

#####

# Sample some data from the dataset!
data = minerl.data.make("MineRLNavigateDenseVectorObf-v0")

# Iterate through a single epoch using sequences of at most 32 steps
for obs, rew, done, act in data.batch_iter(num_epochs=1, batch_size=32):
    # Do something
```

### 1.10.4 MineRLNavigateExtremeVectorObf-v0

In this task, the agent must move to a goal location denoted by a diamond block. This represents a basic primitive used in many tasks throughout Minecraft. In addition to standard observations, the agent has access to a “compass” observation, which points near the goal location, 64 meters from the start location. The goal has a small random

horizontal offset from the compass location and may be slightly below surface level. On the goal location is a unique block, so the agent must find the final goal by searching based on local visual features.

The agent is given a sparse reward (+100 upon reaching the goal, at which point the episode terminates). **This variant of the environment is sparse.**

In this environment, the agent spawns in an extreme hills biome.

### Observation Space

```
Dict({
  "pov": "Box(low=0, high=255, shape=(64, 64, 3))",
  "vector": "Box(low=-1.2000000476837158, high=1.2000000476837158, shape=(64,))"
})
```

### Action Space

```
Dict({
  "vector": "Box(low=-1.0499999523162842, high=1.0499999523162842, shape=(64,))"
})
```

### Usage

```
import gym
import minerl

# Run a random agent through the environment
env = gym.make("MineRLNavigateExtremeVectorObf-v0") # A MineRLNavigateExtremeVectorObf-
↪v0 env

obs = env.reset()
done = False

while not done:
    # Take a no-op through the environment.
    obs, rew, done, _ = env.step(env.action_space.noop())
    # Do something

#####

# Sample some data from the dataset!
data = minerl.data.make("MineRLNavigateExtremeVectorObf-v0")

# Iterate through a single epoch using sequences of at most 32 steps
for obs, rew, done, act in data.batch_iter(num_epochs=1, batch_size=32):
    # Do something
```



### 1.10.5 MineRLNavigateExtremeDenseVectorObf-v0

In this task, the agent must move to a goal location denoted by a diamond block. This represents a basic primitive used in many tasks throughout Minecraft. In addition to standard observations, the agent has access to a “compass” observation, which points near the goal location, 64 meters from the start location. The goal has a small random horizontal offset from the compass location and may be slightly below surface level. On the goal location is a unique block, so the agent must find the final goal by searching based on local visual features.

The agent is given a sparse reward (+100 upon reaching the goal, at which point the episode terminates). **This variant of the environment is dense reward-shaped where the agent is given a reward every tick for how much closer (or negative reward for farther) the agent gets to the target.**

In this environment, the agent spawns in an extreme hills biome.

#### Observation Space

```
Dict({
  "pov": "Box(low=0, high=255, shape=(64, 64, 3))",
  "vector": "Box(low=-1.20000000476837158, high=1.20000000476837158, shape=(64,))"
})
```

#### Action Space

```
Dict({
  "vector": "Box(low=-1.0499999523162842, high=1.0499999523162842, shape=(64,))"
})
```

#### Usage

```
import gym
import minerl

# Run a random agent through the environment
env = gym.make("MineRLNavigateExtremeDenseVectorObf-v0") # A_
↳ MineRLNavigateExtremeDenseVectorObf-v0 env

obs = env.reset()
done = False

while not done:
    # Take a no-op through the environment.
    obs, rew, done, _ = env.step(env.action_space.noop())
    # Do something

#####
```

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```

# Sample some data from the dataset!
data = minerl.data.make("MineRLNavigateExtremeDenseVectorObf-v0")

# Iterate through a single epoch using sequences of at most 32 steps
for obs, rew, done, act in data.batch_iter(num_epochs=1, batch_size=32):
    # Do something

```

### 1.10.6 MineRLObtainDiamondVectorObf-v0

In this environment the agent is required to obtain a diamond. The agent begins in a random starting location on a random survival map without any items, matching the normal starting conditions for human players in Minecraft. The agent is given access to a selected summary of its inventory and GUI free crafting, smelting, and inventory management actions.

During an episode the agent is rewarded **every** time it obtains an item in the requisite item hierarchy to obtaining a diamond. The rewards for each item are given here:

```

<Item reward="1" type="log" />
<Item reward="2" type="planks" />
<Item reward="4" type="stick" />
<Item reward="4" type="crafting_table" />
<Item reward="8" type="wooden_pickaxe" />
<Item reward="16" type="cobblestone" />
<Item reward="32" type="furnace" />
<Item reward="32" type="stone_pickaxe" />
<Item reward="64" type="iron_ore" />
<Item reward="128" type="iron_ingot" />
<Item reward="256" type="iron_pickaxe" />
<Item reward="1024" type="diamond" />

```

### Observation Space

```

Dict({
    "pov": "Box(low=0, high=255, shape=(64, 64, 3))",
    "vector": "Box(low=-1.2000000476837158, high=1.2000000476837158, shape=(64,))"
})

```

## Action Space

```
Dict({
  "vector": "Box(low=-1.0499999523162842, high=1.0499999523162842, shape=(64,))"
})
```

## Usage

```
import gym
import minerl

# Run a random agent through the environment
env = gym.make("MineRLObtainDiamondVectorObf-v0") # A MineRLObtainDiamondVectorObf-v0_
↪env

obs = env.reset()
done = False

while not done:
    # Take a no-op through the environment.
    obs, rew, done, _ = env.step(env.action_space.noop())
    # Do something

#####

# Sample some data from the dataset!
data = minerl.data.make("MineRLObtainDiamondVectorObf-v0")

# Iterate through a single epoch using sequences of at most 32 steps
for obs, rew, done, act in data.batch_iter(num_epochs=1, batch_size=32):
    # Do something
```

### 1.10.7 MineRLObtainDiamondDenseVectorObf-v0

In this environment the agent is required to obtain a diamond. The agent begins in a random starting location on a random survival map without any items, matching the normal starting conditions for human players in Minecraft. The agent is given access to a selected summary of its inventory and GUI free crafting, smelting, and inventory management actions.

During an episode the agent is rewarded **every** time it obtains an item in the requisite item hierarchy to obtaining a diamond. The rewards for each item are given here:

```
<Item reward="1" type="log" />
<Item reward="2" type="planks" />
<Item reward="4" type="stick" />
```

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```

<Item reward="4" type="crafting_table" />
<Item reward="8" type="wooden_pickaxe" />
<Item reward="16" type="cobblestone" />
<Item reward="32" type="furnace" />
<Item reward="32" type="stone_pickaxe" />
<Item reward="64" type="iron_ore" />
<Item reward="128" type="iron_ingot" />
<Item reward="256" type="iron_pickaxe" />
<Item reward="1024" type="diamond" />

```

## Observation Space

```

Dict({
  "pov": "Box(low=0, high=255, shape=(64, 64, 3))",
  "vector": "Box(low=-1.20000000476837158, high=1.20000000476837158, shape=(64,))"
})

```

## Action Space

```

Dict({
  "vector": "Box(low=-1.0499999523162842, high=1.0499999523162842, shape=(64,))"
})

```

## Usage

```

import gym
import minerl

# Run a random agent through the environment
env = gym.make("MineRLObtainDiamondDenseVectorObf-v0") # A_
↳ MineRLObtainDiamondDenseVectorObf-v0 env

obs = env.reset()
done = False

while not done:
    # Take a no-op through the environment.
    obs, rew, done, _ = env.step(env.action_space.noop())
    # Do something

#####

# Sample some data from the dataset!
data = minerl.data.make("MineRLObtainDiamondDenseVectorObf-v0")

# Iterate through a single epoch using sequences of at most 32 steps
for obs, rew, done, act in data.batch_iter(num_epochs=1, batch_size=32):
    # Do something

```

### 1.10.8 MineRLObtainIronPickaxeVectorObf-v0

In this environment the agent is required to obtain an iron pickaxe. The agent begins in a random starting location, on a random survival map, without any items, matching the normal starting conditions for human players in Minecraft. The agent is given access to a selected view of its inventory and GUI free crafting, smelting, and inventory management actions.

During an episode **the agent is rewarded only once per item the first time it obtains that item in the requisite item hierarchy for obtaining an iron pickaxe.** The reward for each item is given here:

```
<Item amount="1" reward="1" type="log" />
<Item amount="1" reward="2" type="planks" />
<Item amount="1" reward="4" type="stick" />
<Item amount="1" reward="4" type="crafting_table" />
<Item amount="1" reward="8" type="wooden_pickaxe" />
<Item amount="1" reward="16" type="cobblestone" />
<Item amount="1" reward="32" type="furnace" />
<Item amount="1" reward="32" type="stone_pickaxe" />
<Item amount="1" reward="64" type="iron_ore" />
<Item amount="1" reward="128" type="iron_ingot" />
<Item amount="1" reward="256" type="iron_pickaxe" />
```

#### Observation Space

```
Dict({
  "pov": "Box(low=0, high=255, shape=(64, 64, 3))",
  "vector": "Box(low=-1.20000000476837158, high=1.20000000476837158, shape=(64,))"
})
```

#### Action Space

```
Dict({
  "vector": "Box(low=-1.0499999523162842, high=1.0499999523162842, shape=(64,))"
})
```

#### Usage

```
import gym
import minerl

# Run a random agent through the environment
env = gym.make("MineRLObtainIronPickaxeVectorObf-v0") # A
↳ MineRLObtainIronPickaxeVectorObf-v0 env

obs = env.reset()
```

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```

done = False

while not done:
    # Take a no-op through the environment.
    obs, rew, done, _ = env.step(env.action_space.noop())
    # Do something

#####

# Sample some data from the dataset!
data = minerl.data.make("MineRLObtainIronPickaxeVectorObf-v0")

# Iterate through a single epoch using sequences of at most 32 steps
for obs, rew, done, act in data.batch_iter(num_epochs=1, batch_size=32):
    # Do something

```

### 1.10.9 MineRLObtainIronPickaxeDenseVectorObf-v0

In this environment the agent is required to obtain an iron pickaxe. The agent begins in a random starting location, on a random survival map, without any items, matching the normal starting conditions for human players in Minecraft. The agent is given access to a selected view of its inventory and GUI free crafting, smelting, and inventory management actions.

During an episode **the agent is rewarded only once per item the first time it obtains that item in the requisite item hierarchy for obtaining an iron pickaxe.** The reward for each item is given here:

```

<Item amount="1" reward="1" type="log" />
<Item amount="1" reward="2" type="planks" />
<Item amount="1" reward="4" type="stick" />
<Item amount="1" reward="4" type="crafting_table" />
<Item amount="1" reward="8" type="wooden_pickaxe" />
<Item amount="1" reward="16" type="cobblestone" />
<Item amount="1" reward="32" type="furnace" />
<Item amount="1" reward="32" type="stone_pickaxe" />
<Item amount="1" reward="64" type="iron_ore" />
<Item amount="1" reward="128" type="iron_ingot" />
<Item amount="1" reward="256" type="iron_pickaxe" />

```

## Observation Space

```
Dict({
  "pov": "Box(low=0, high=255, shape=(64, 64, 3))",
  "vector": "Box(low=-1.20000000476837158, high=1.20000000476837158, shape=(64,))"
})
```

## Action Space

```
Dict({
  "vector": "Box(low=-1.0499999523162842, high=1.0499999523162842, shape=(64,))"
})
```

## Usage

```
import gym
import minerl

# Run a random agent through the environment
env = gym.make("MineRLObtainIronPickaxeDenseVectorObf-v0") # A_
↳ MineRLObtainIronPickaxeDenseVectorObf-v0 env

obs = env.reset()
done = False

while not done:
    # Take a no-op through the environment.
    obs, rew, done, _ = env.step(env.action_space.noop())
    # Do something

#####

# Sample some data from the dataset!
data = minerl.data.make("MineRLObtainIronPickaxeDenseVectorObf-v0")

# Iterate through a single epoch using sequences of at most 32 steps
for obs, rew, done, act in data.batch_iter(num_epochs=1, batch_size=32):
    # Do something
```

## 1.11 MineRL BASALT Competition Environments

### 1.11.1 MineRLBasaltFindCave-v0

After spawning in a plains biome, explore and find a cave. When inside a cave, throw a snowball to end episode.

## Observation Space

```
Dict({
  "equipped_items": {
    "mainhand": {
      "damage": "Box(low=-1, high=1562, shape=())",
      "maxDamage": "Box(low=-1, high=1562, shape=())",
      "type": "Enum(air,bucket,carrot,cobblestone,fence,fence_gate,none,
↪other,snowball,stone_pickaxe,stone_shovel,water_bucket,wheat,wheat_seeds)"
    }
  },
  "inventory": {
    "bucket": "Box(low=0, high=2304, shape=())",
    "carrot": "Box(low=0, high=2304, shape=())",
    "cobblestone": "Box(low=0, high=2304, shape=())",
    "fence": "Box(low=0, high=2304, shape=())",
    "fence_gate": "Box(low=0, high=2304, shape=())",
    "snowball": "Box(low=0, high=2304, shape=())",
    "stone_pickaxe": "Box(low=0, high=2304, shape=())",
    "stone_shovel": "Box(low=0, high=2304, shape=())",
    "water_bucket": "Box(low=0, high=2304, shape=())",
    "wheat": "Box(low=0, high=2304, shape=())",
    "wheat_seeds": "Box(low=0, high=2304, shape=())"
  },
  "pov": "Box(low=0, high=255, shape=(64, 64, 3))"
})
```

## Action Space

```
Dict({
  "attack": "Discrete(2)",
  "back": "Discrete(2)",
  "camera": "Box(low=-180.0, high=180.0, shape=(2,))",
  "equip": "Enum(air,bucket,carrot,cobblestone,fence,fence_gate,none,other,snowball,
↪stone_pickaxe,stone_shovel,water_bucket,wheat,wheat_seeds)",
  "forward": "Discrete(2)",
  "jump": "Discrete(2)",
  "left": "Discrete(2)",
  "right": "Discrete(2)",
  "sneak": "Discrete(2)",
  "sprint": "Discrete(2)",
  "use": "Discrete(2)"
})
```



## Starting Inventory

```
Dict({
  "snowball": 1
})
```

## Usage

```
import gym
import minerl

# Run a random agent through the environment
env = gym.make("MineRLBasaltFindCave-v0") # A MineRLBasaltFindCave-v0 env

obs = env.reset()
done = False

while not done:
    # Take a no-op through the environment.
    obs, rew, done, _ = env.step(env.action_space.noop())
    # Do something

#####

# Sample some data from the dataset!
data = minerl.data.make("MineRLBasaltFindCave-v0")

# Iterate through a single epoch using sequences of at most 32 steps
for obs, rew, done, act in data.batch_iter(num_epochs=1, batch_size=32):
    # Do something
```

### 1.11.2 MineRLBasaltMakeWaterfall-v0

After spawning in an extreme hills biome, use your waterbucket to make an beautiful waterfall. Then take an aesthetic “picture” of it by moving to a good location, positioning player’s camera to have a nice view of the waterfall, and throwing a snowball. Throwing the snowball ends the episode.

## Observation Space

```
Dict({
  "equipped_items": {
    "mainhand": {
      "damage": "Box(low=-1, high=1562, shape=())",
      "maxDamage": "Box(low=-1, high=1562, shape=())",
      "type": "Enum(air,bucket,carrot,cobblestone,fence,fence_gate,none,
↪other,snowball,stone_pickaxe,stone_shovel,water_bucket,wheat,wheat_seeds)"
    }
  },
  "inventory": {
    "bucket": "Box(low=0, high=2304, shape=())",
    "carrot": "Box(low=0, high=2304, shape=())",
    "cobblestone": "Box(low=0, high=2304, shape=())",
    "fence": "Box(low=0, high=2304, shape=())",
    "fence_gate": "Box(low=0, high=2304, shape=())",
    "snowball": "Box(low=0, high=2304, shape=())",
    "stone_pickaxe": "Box(low=0, high=2304, shape=())",
    "stone_shovel": "Box(low=0, high=2304, shape=())",
    "water_bucket": "Box(low=0, high=2304, shape=())",
    "wheat": "Box(low=0, high=2304, shape=())",
    "wheat_seeds": "Box(low=0, high=2304, shape=())"
  },
  "pov": "Box(low=0, high=255, shape=(64, 64, 3))"
})
```

## Action Space

```
Dict({
  "attack": "Discrete(2)",
  "back": "Discrete(2)",
  "camera": "Box(low=-180.0, high=180.0, shape=(2,))",
  "equip": "Enum(air,bucket,carrot,cobblestone,fence,fence_gate,none,other,snowball,
↪stone_pickaxe,stone_shovel,water_bucket,wheat,wheat_seeds)",
  "forward": "Discrete(2)",
  "jump": "Discrete(2)",
  "left": "Discrete(2)",
  "right": "Discrete(2)",
  "sneak": "Discrete(2)",
  "sprint": "Discrete(2)",
  "use": "Discrete(2)"
})
```

## Starting Inventory

```
Dict({
  "cobblestone": 20,
  "snowball": 1,
  "stone_pickaxe": 1,
  "stone_shovel": 1,
  "water_bucket": 1
})
```

## Usage

```
import gym
import minerl

# Run a random agent through the environment
env = gym.make("MineRLBasaltMakeWaterfall-v0") # A MineRLBasaltMakeWaterfall-v0 env

obs = env.reset()
done = False

while not done:
    # Take a no-op through the environment.
    obs, rew, done, _ = env.step(env.action_space.noop())
    # Do something

#####

# Sample some data from the dataset!
data = minerl.data.make("MineRLBasaltMakeWaterfall-v0")

# Iterate through a single epoch using sequences of at most 32 steps
for obs, rew, done, act in data.batch_iter(num_epochs=1, batch_size=32):
    # Do something
```

### 1.11.3 MineRLBasaltCreateVillageAnimalPen-v0

After spawning in a plains village, surround two or more animals of the same type in a fenced area (a pen), constructed near the house. You can't have more than one type of animal in your enclosed area. Allowed animals are chickens, sheep, cows, and pigs.

Do not harm villagers or existing village structures in the process.

Throw a snowball to end the episode.

## Observation Space

```
Dict({
  "equipped_items": {
    "mainhand": {
      "damage": "Box(low=-1, high=1562, shape=())",
      "maxDamage": "Box(low=-1, high=1562, shape=())",
      "type": "Enum(air,bucket,carrot,cobblestone,fence,fence_gate,none,
↪other,snowball,stone_pickaxe,stone_shovel,water_bucket,wheat,wheat_seeds)"
    }
  },
  "inventory": {
    "bucket": "Box(low=0, high=2304, shape=())",
    "carrot": "Box(low=0, high=2304, shape=())",
    "cobblestone": "Box(low=0, high=2304, shape=())",
    "fence": "Box(low=0, high=2304, shape=())",
    "fence_gate": "Box(low=0, high=2304, shape=())",
    "snowball": "Box(low=0, high=2304, shape=())",
    "stone_pickaxe": "Box(low=0, high=2304, shape=())",
    "stone_shovel": "Box(low=0, high=2304, shape=())",
    "water_bucket": "Box(low=0, high=2304, shape=())",
    "wheat": "Box(low=0, high=2304, shape=())",
    "wheat_seeds": "Box(low=0, high=2304, shape=())"
  },
  "pov": "Box(low=0, high=255, shape=(64, 64, 3))"
})
```

## Action Space

```
Dict({
  "attack": "Discrete(2)",
  "back": "Discrete(2)",
  "camera": "Box(low=-180.0, high=180.0, shape=(2,))",
  "equip": "Enum(air,bucket,carrot,cobblestone,fence,fence_gate,none,other,snowball,
↪stone_pickaxe,stone_shovel,water_bucket,wheat,wheat_seeds)",
  "forward": "Discrete(2)",
  "jump": "Discrete(2)",
  "left": "Discrete(2)",
  "right": "Discrete(2)",
  "sneak": "Discrete(2)",
  "sprint": "Discrete(2)",
  "use": "Discrete(2)"
})
```

## Starting Inventory

```
Dict({
  "carrot": 1,
  "fence": 64,
  "fence_gate": 64,
  "snowball": 1,
  "wheat": 1,
  "wheat_seeds": 1
})
```

## Usage

```
import gym
import minerl

# Run a random agent through the environment
env = gym.make("MineRLBasaltCreateVillageAnimalPen-v0") # A_
↳ MineRLBasaltCreateVillageAnimalPen-v0 env

obs = env.reset()
done = False

while not done:
    # Take a no-op through the environment.
    obs, rew, done, _ = env.step(env.action_space.noop())
    # Do something

#####

# Sample some data from the dataset!
data = minerl.data.make("MineRLBasaltCreateVillageAnimalPen-v0")

# Iterate through a single epoch using sequences of at most 32 steps
for obs, rew, done, act in data.batch_iter(num_epochs=1, batch_size=32):
    # Do something
```

### 1.11.4 MineRLBasaltBuildVillageHouse-v0

Build a house in the style of the village without damaging the village. Give a tour of the house and then throw a snowball to end the episode.

**Note:** In the observation and action spaces, the following (internal Minecraft) item IDs can be interpreted as follows:

- log#0 is oak logs.

- log#1 is spruce logs.
  - log2#0 is acacia logs.
  - planks#0 is oak planks.
  - planks#1 is spruce planks.
  - planks#4 is acacia planks.
  - sandstone#0 is cracked sandstone.
  - sandstone#2 is smooth sandstone.
- 

**Tip:** You can find detailed information on which materials are used in each biome-specific village (plains, savannah, taiga, desert) here: [https://minecraft.fandom.com/wiki/Village/Structure\\_\(old\)/Blueprints#Village\\_generation](https://minecraft.fandom.com/wiki/Village/Structure_(old)/Blueprints#Village_generation)

---

## Observation Space

```
Dict({
  "equipped_items": {
    "mainhand": {
      "damage": "Box(low=-1, high=1562, shape=())",
      "maxDamage": "Box(low=-1, high=1562, shape=())",
      "type": "Enum(acacia_door,acacia_fence,cactus,cobblestone,dirt,fence,
↪flower_pot,glass,ladder,log#0,log#1,log2#0,none,other,planks#0,planks#1,planks#4,red_
↪flower,sand,sandstone#0,sandstone#2,sandstone_stairs,snowball,spruce_door,spruce_fence,
↪stone_axe,stone_pickaxe,stone_stairs,torch,wooden_door,wooden_pressure_plate)"
    }
  },
  "inventory": {
    "acacia_door": "Box(low=0, high=2304, shape=())",
    "acacia_fence": "Box(low=0, high=2304, shape=())",
    "cactus": "Box(low=0, high=2304, shape=())",
    "cobblestone": "Box(low=0, high=2304, shape=())",
    "dirt": "Box(low=0, high=2304, shape=())",
    "fence": "Box(low=0, high=2304, shape=())",
    "flower_pot": "Box(low=0, high=2304, shape=())",
    "glass": "Box(low=0, high=2304, shape=())",
    "ladder": "Box(low=0, high=2304, shape=())",
    "log#0": "Box(low=0, high=2304, shape=())",
    "log#1": "Box(low=0, high=2304, shape=())",
    "log2#0": "Box(low=0, high=2304, shape=())",
    "planks#0": "Box(low=0, high=2304, shape=())",
    "planks#1": "Box(low=0, high=2304, shape=())",
    "planks#4": "Box(low=0, high=2304, shape=())",
    "red_flower": "Box(low=0, high=2304, shape=())",
    "sand": "Box(low=0, high=2304, shape=())",
    "sandstone#0": "Box(low=0, high=2304, shape=())",
    "sandstone#2": "Box(low=0, high=2304, shape=())",
    "sandstone_stairs": "Box(low=0, high=2304, shape=())",
    "snowball": "Box(low=0, high=2304, shape=())",
    "spruce_door": "Box(low=0, high=2304, shape=())",
    "spruce_fence": "Box(low=0, high=2304, shape=())",
  }
})
```

```

    "stone_axe": "Box(low=0, high=2304, shape=())",
    "stone_pickaxe": "Box(low=0, high=2304, shape=())",
    "stone_stairs": "Box(low=0, high=2304, shape=())",
    "torch": "Box(low=0, high=2304, shape=())",
    "wooden_door": "Box(low=0, high=2304, shape=())",
    "wooden_pressure_plate": "Box(low=0, high=2304, shape=())"
  },
  "pov": "Box(low=0, high=255, shape=(64, 64, 3))"
})

```

## Action Space

```

Dict({
  "attack": "Discrete(2)",
  "back": "Discrete(2)",
  "camera": "Box(low=-180.0, high=180.0, shape=(2,))",
  "equip": "Enum(acacia_door,acacia_fence,cactus,cobblestone,dirt,fence,flower_pot,
↪glass,ladder,log#0,log#1,log2#0,none,other,planks#0,planks#1,planks#4,red_flower,sand,
↪sandstone#0,sandstone#2,sandstone_stairs,snowball,spruce_door,spruce_fence,stone_axe,
↪stone_pickaxe,stone_stairs,torch,wooden_door,wooden_pressure_plate)",
  "forward": "Discrete(2)",
  "jump": "Discrete(2)",
  "left": "Discrete(2)",
  "right": "Discrete(2)",
  "sneak": "Discrete(2)",
  "sprint": "Discrete(2)",
  "use": "Discrete(2)"
})

```

## Starting Inventory

```

Dict({
  "acacia_door": 64,
  "acacia_fence": 64,
  "cactus": 3,
  "cobblestone": 64,
  "dirt": 64,
  "fence": 64,
  "flower_pot": 3,
  "glass": 64,
  "ladder": 64,
  "log#0": 64,
  "log#1": 64,
  "log2#0": 64,
  "planks#0": 64,
  "planks#1": 64,
  "planks#4": 64,
  "red_flower": 3,
  "sand": 64,
  "sandstone#0": 64,
  "sandstone#2": 64,
  "sandstone_stairs": 64,
  "snowball": 1,

```

```
"spruce_door": 64,
"spruce_fence": 64,
"stone_axe": 1,
"stone_pickaxe": 1,
"stone_stairs": 64,
"torch": 64,
"wooden_door": 64,
"wooden_pressure_plate": 64
})
```

## Usage

```
import gym
import minerl

# Run a random agent through the environment
env = gym.make("MineRLBasaltBuildVillageHouse-v0") # A MineRLBasaltBuildVillageHouse-v0_
↪env

obs = env.reset()
done = False

while not done:
    # Take a no-op through the environment.
    obs, rew, done, _ = env.step(env.action_space.noop())
    # Do something

#####

# Sample some data from the dataset!
data = minerl.data.make("MineRLBasaltBuildVillageHouse-v0")

# Iterate through a single epoch using sequences of at most 32 steps
for obs, rew, done, act in data.batch_iter(num_epochs=1, batch_size=32):
    # Do something
```

## 1.12 Performance tips

### 1.12.1 Slowdown in obfuscated environments

Obfuscated environments, like `MineRLObtainDiamondVectorObf-v0` make extensive use of `np.dot` function, which by default is parallelized over multiple threads. Since the vectors/matrices are small, the overhead from this outweighs benefits, and the environment appears much slower than it really is.

To speed up obfuscated environments, try setting environment variable `OMP_NUM_THREADS=1` to restrict Numpy to only use one thread.



## 1.13 Links to papers and projects

Here you can find useful links to the presentations, code and papers of the finalists in previous MineRL competitions, as well as other publications and projects that use MineRL.

To see all papers that cite MineRL, check [Google Scholar](#). You can also create alerts there to get notified whenever a new citation appears.

If you want to add your paper/project here, do not hesitate to create a pull request in the [main repository](#)!

### 1.13.1 Presentations

- [MineRL 2019 - Finalists presentations at NeurIPS 2019](#)
- [MineRL 2019 - 1st place winners presentation, longer one \(slides in English, talk in Russian\)](#)
- [MineRL 2020 - Round 1 finalists presentations at NeurIPS 2020](#)
- [MineRL 2020 - Round 2 finalists presentations at Microsoft AI and Gaming Research Summit 2021](#)

### 1.13.2 MineRL papers

- [MineRL: A Large-Scale Dataset of Minecraft Demonstrations](#)
- [The MineRL 2019 Competition on Sample Efficient Reinforcement Learning using Human Priors](#)
- [Retrospective Analysis of the 2019 MineRL Competition on Sample Efficient Reinforcement Learning](#)
- [The MineRL 2020 Competition on Sample Efficient Reinforcement Learning using Human Priors](#)
- [Towards robust and domain agnostic reinforcement learning competitions: MineRL 2020](#)

### 1.13.3 2019 competitor code/papers

- 1st place: [paper](#).
- 2nd place: [paper](#), [code](#).
- 3rd place: [paper](#), [code](#).
- 4th place: [code](#).
- 5th place: [paper](#), [code](#).

### 1.13.4 2020 competitor code/papers

- 1st place: [paper](#).
- 3rd place: [code](#).
- 4th place: [code](#).

### 1.13.5 Other papers that use the MineRL environment

- PiCoEDL: Discovery and Learning of Minecraft Navigation Goals from Pixels and Coordinates (CVPR Embodied AI Workshop, 2021)
- Universal Value Iteration Networks: When Spatially-Invariant Is Not Universal (AAAI, 2020)
- Multi-task curriculum learning in a complex, visual, hard-exploration domain: Minecraft
- Follow up paper from the #1 team in 2019 (obtains diamond): [paper](#), [code](#).
- Align-RUDDER: Learning From Few Demonstrations by Reward Redistribution (obtains diamond): [paper](#), [code](#).

### 1.13.6 Other

- Data analysis for vector obfuscation/kmeans
- Malmo and MineRL tutorial

## 1.14 Windows FAQ

This note serves as a collection of fixes for errors which may occur on the Windows platform.

### 1.14.1 The The system cannot find the path specified error (installing)

If during installation you get errors regarding missing files or unspecified paths, followed by a long path string, you might be limited by the `MAX_PATH` setting on windows. Try removing this limitation with [these instructions](#).

### 1.14.2 The freeze\_support error (multiprocessing)

RuntimeError:

```
An attempt has been made to start a new process before the
current process has finished its bootstrapping phase.
```

This probably means that you are not using `fork` to start your child processes and you have forgotten to use the proper idiom in the main module:

```
if __name__ == '__main__':
    freeze_support()
    ...
```

The "`freeze_support()`" line can be omitted if the program is not going to be frozen to produce an executable.

The implementation of multiprocessing is different on Windows, which uses `spawn` instead of `fork`. So we have to wrap the code with an if-clause to protect the code from executing multiple times. Refactor your code into the following structure.

```

import minerl
import gym

def main()
    # do your main minerl code
    env = gym.make('MineRLTreechop-v0')

if __name__ == '__main__':
    main()

```

## 1.15 minerl.env

The `minerl.env` package provides an optimized python software layer over *MineRLEnv*, a fork of the popular Minecraft simulator Malmö which enables **synchronous**, **stable**, and **fast** samples from the Minecraft environment.

### 1.15.1 MineRLEnv

### 1.15.2 InstanceManager

```

class minerl.env.malmo.CustomAsyncRemoteMethod(proxy, name, max_retries)
    Bases: Pyro4.core._AsyncRemoteMethod

```

```

class minerl.env.malmo.InstanceManager
    Bases: object

```

The Minecraft instance manager library. The instance manager can be used to allocate and safely terminate existing Malmö instances for training agents.

Note: This object never needs to be explicitly invoked by the user of the MineRL library as the creation of one of the several MineRL environments will automatically query the InstanceManager to create a new instance.

Note: In future versions of MineRL the instance manager will become its own daemon process which provides instance allocation capability using remote procedure calls.

```

DEFAULT_IP = 'localhost'

```

```

KEEP_ALIVE_PYRO_FREQUENCY = 5

```

```

MAXINSTANCES = None

```

```

MINECRAFT_DIR = '/home/docs/checkouts/readthedocs.org/user_builds/minerl/envs/stable/lib/python3.7/site-packages/minerl/env/./Malmö/Minecraft'

```

```

REMOTE = False

```

```

SCHEMAS_DIR = '/home/docs/checkouts/readthedocs.org/user_builds/minerl/envs/stable/lib/python3.7/site-packages/minerl/env/./Malmö/Schemas'

```

```

STATUS_DIR = '/home/docs/checkouts/readthedocs.org/user_builds/minerl/envs/stable/lib/python3.7/site-packages/sphinx/performance'

```

```

X11_DIR = '/tmp/.X11-unix'

```

```

classmethod add_existing_instance(port)

```

```

classmethod add_keep_alive(_pid, _callback)

```

**classmethod** `allocate_pool(num)`

**classmethod** `configure_malmo_base_port(malmo_base_port)`

Configure the lowest or base port for Malmo

**classmethod** `get_instance(pid, instance_id=None)`

Gets an instance from the instance manager. This method is a context manager and therefore when the context is entered the method yields a `InstanceManager.Instance` object which contains the allocated port and host for the given instance that was created.

**Yields** The allocated `InstanceManager.Instance` object.

**Raises**

- **RuntimeError** – No available instances or the maximum number of allocated instances reached.
- **RuntimeError** – No available instances and automatic allocation of instances is off.

`headless = False`

**classmethod** `is_remote()`

`managed = True`

`ninstances = 0`

**classmethod** `set_valid_jdwp_port_for_instance(instance) → None`

Find a valid port for JDWP (Java Debug Wire Protocol), so that the instance can be debugged with an attached debugger. The port is set in the instance, so that other instances can check whether the port is reserved. :param instance: Instance to find and port for, and where we will set the jdwp port.

**classmethod** `shutdown()`

**class** `minerl.env.malmo.MinecraftInstance(port=None, existing=False, status_dir=None, seed=None, instance_id=None)`

Bases: `object`

A subprocess wrapper which maintains a reference to a minecraft subprocess and also allows for stable closing and launching of such subprocesses across different platforms.

The Minecraft instance class works by launching two subprocesses: the Malmo subprocess, and a watcher subprocess with access to the process IDs of both the parent process and the Malmo subprocess. If the parent process dies, it will kill the subprocess, and then itself.

This scheme has a single failure point of the process dying before the watcher process is launched.

`MAX_PIPE_LENGTH = 500`

**property** `actor_name`

**property** `client_socket`

`client_socket_close()`

`client_socket_recv_message()`

`client_socket_send_message(msg)`

`client_socket_shutdown(param)`

`close()`

Closes the object.

`create_multiagent_instance_socket(socktime)`

`get_output()`

```

property had_to_clean
has_client_socket()
property host
property jdpw_port
    JDWP (Java Debug Wire Protocol) port, if any, so the instance can be debugged with an attached debugger.
kill()
    Kills the process (if it has been launched.)
launch(daemonize=False, replaceable=True)
property port
release_lock()
property status_dir
class minerl.env.malmo.SeedType(value)
    Bases: enum.IntEnum
    The seed type for an instance manager.
Values: 0 - NONE: No seeding whatsoever. 1 - CONSTANT: All environments have the same seed (the one
    specified
        to the instance manager) (or alist of seeds , separated)

    2 - GENERATED: All environments have different seeds generated from a single random generator
        with the seed specified to the InstanceManager.

    3 - SPECIFIED: Each instance is given a list of seeds. Specify this like
        1,2,3,4;848,432,643;888,888,888 Each instance's seed list is separated by ; and each seed is
        separated by ,

    CONSTANT = 1
    GENERATED = 2
    NONE = 0
    SPECIFIED = 3
    classmethod get_index(type)
minerl.env.malmo.launch_instance_manager()
    Defines the entry point for the remote procedure call server.
minerl.env.malmo.launch_queue_logger_thread(output_producer, should_end)

```

## 1.16 minerl.data

The `minerl.data` package provides a unified interface for sampling data from the *MineRL-v0 Dataset*. Data is accessed by making a dataset from one of the `minerl` environments and iterating over it using one of the iterators provided by the `minerl.data.DataPipeline`

The following is a description of the various methods included within the package as well as some basic usage examples. To see more detailed [descriptions and tutorials](#) on how to use the data API, please [take a look at our numerous getting started manuals](#).

### 1.16.1 MineRLv0

**class** `minerl.data.DataPipeline`(*data\_directory*: <module 'posixpath' from  
*/home/docs/checkouts/readthedocs.org/user\_builds/minerl/envs/stable/lib/python3.7/posixpath*  
*environment*: str, *num\_workers*: int, *worker\_batch\_size*: int,  
*min\_size\_to\_dequeue*: int, *random\_seed*=42)

Bases: object

Creates a data pipeline object used to iterate through the MineRL-v0 dataset

**property** `action_space`

action space of current MineRL environment

**Type Returns**

**batch\_iter**(*batch\_size*: int, *seq\_len*: int, *num\_epochs*: int = - 1, *preload\_buffer\_size*: int = 2, *seed*:  
*Optional[int]* = None)

Returns batches of sequences length SEQ\_LEN of the data of size BATCH\_SIZE. The iterator produces batches sequentially. If an element of a batch reaches the end of its episode, it will be appended with a new episode.

If you wish to obtain metadata of the episodes, consider using `load_data` instead.

**Parameters**

- **batch\_size** (*int*) – The batch size.
- **seq\_len** (*int*) – The size of sequences to produce.
- **num\_epochs** (*int*, *optional*) – The number of epochs to iterate over the data. Defaults to -1.
- **preload\_buffer\_size** (*int*, *optional*) – Increase to IMPROVE PERFORMANCE. The data iterator uses a queue to prevent blocking, the queue size is the number of trajectories to load into the buffer. Adjust based on memory constraints. Defaults to 32.
- **seed** (*int*, *optional*) – [int]. NOT IMPLEMENTED Defaults to None.

**Returns** A generator that yields (sarsd) batches

**Return type** Generator

**get\_trajectory\_names**()

Gets all the trajectory names

**Returns** [description]

**Return type** A list of experiment names

**load\_data**(*stream\_name*: str, *skip\_interval*=0, *include\_metadata*=False, *include\_monitor\_data*=False)

Iterates over an individual trajectory named stream\_name.

**Parameters**

- **stream\_name** (*str*) – The stream name desired to be iterated through.
- **skip\_interval** (*int*, *optional*) – How many sices should be skipped.. Defaults to 0.
- **include\_metadata** (*bool*, *optional*) – Whether or not meta data about the loaded trajectory should be included.. Defaults to False.
- **include\_monitor\_data** (*bool*, *optional*) – Whether to include all of the monitor data from the environment. Defaults to False.

**Yields** A tuple of (state, player\_action, reward\_from\_action, next\_state, is\_next\_state\_terminal). These are tuples are yielded in order of the episode.

**property observation\_space**

action space of current MineRL environment

**Type** Returns

**static read\_frame**(*cap*)

**sarsd\_iter**(*num\_epochs=-1, max\_sequence\_len=32, queue\_size=None, seed=None, include\_metadata=False*)

Returns a generator for iterating through (state, action, reward, next\_state, is\_terminal) tuples in the dataset. Loads num\_workers files at once as defined in minerl.data.make() and return up to max\_sequence\_len consecutive samples wrapped in a dict observation space

**Parameters**

- **num\_epochs** (*int, optional*) – number of epochs to iterate over or -1 to loop forever. Defaults to -1
- **max\_sequence\_len** (*int, optional*) – maximum number of consecutive samples - may be less. Defaults to 32
- **seed** (*int, optional*) – seed for random directory walk - note, specifying seed as well as a finite num\_epochs will cause the ordering of examples to be the same after every call to seq\_iter
- **queue\_size** (*int, optional*) – maximum number of elements to buffer at a time, each worker may hold an additional item while waiting to enqueue. Defaults to 16\*self.number\_of\_workers or 2\* self.number\_of\_workers if max\_sequence\_len == -1
- **include\_metadata** (*bool, optional*) – adds an additional member to the tuple containing metadata about the stream the data was loaded from. Defaults to False

**Yields** A tuple of (state, player\_action, reward\_from\_action, next\_state, is\_next\_state\_terminal, (metadata)). Each element is in the format of the environment action/state/reward space and contains as many samples are requested.

**seq\_iter**(*num\_epochs=-1, max\_sequence\_len=32, queue\_size=None, seed=None, include\_metadata=False*)

DEPRECATED METHOD FOR SAMPLING DATA FROM THE MINERL DATASET.

This function is now `DataPipeline.batch_iter()`

**property spec:** `minerl.herobrine.env_spec.EnvSpec`

`minerl.data.download`(*directory: Optional[str] = None, environment: Optional[str] = None, competition: Optional[str] = None, resolution: str = 'low', texture\_pack: int = 0, update\_environment\_variables: bool = True, disable\_cache: bool = False*) → None

Low-level interface for downloading MineRL dataset.

Using the `python -m minerl.data.download` CLI script is preferred because it performs more input validation and hides internal-use arguments.

Run this command with `environment=None` and `competition=None` to download a minimal dataset with 2 demonstrations from each environment. Provide the `environment` or `competition` arguments to download a full dataset for a particular environment or competition.

**Parameters**

- **directory** – Destination folder for downloading MineRL datasets. If None, then use the `MINERL_DATA_ROOT` environment variable, or error if this environment variable is not set.

- **environment** – The name of a MineRL environment or None. If this argument is the name of a MineRL environment and `competition` is None, then this function downloads the full dataset for the specifies MineRL environment.

If both `environment=None` and `competition=None`, then this function downloads a minimal dataset.

- **competition** – The name of a MineRL competition (“diamond” or “basalt”) or None. If this argument is the name of a MineRL environment and `competition` is None, then this function downloads the full dataset for the specified MineRL competition.

If both `environment=None` and `competition=None`, then this function downloads a minimal dataset.

- **resolution** – For internal use only. One of [‘low’, ‘high’] corresponding to video resolutions of [64x64,1024x1024] respectively (note: high resolution is not currently supported).
- **texture\_pack** – For internal use only. 0: default Minecraft texture pack, 1: flat semi-realistic texture pack.
- **update\_environment\_variables** – For internal use only. If True, then export of `MINERL_DATA_ROOT` environment variable (note: for some os this is only for the current shell).
- **disable\_cache** – If False (default), then the tar download and other temporary download files are saved inside `directory`.

If `disable_cache` is False on a future call to this function and temporary download files are detected, then the download is resumed from previous download progress. If `disable_cache` is False on a future call to this function and the completed tar file is detected, then the download is skipped entirely and we immediately extract the tar to `directory`.

`minerl.data.make(environment=None, data_dir=None, num_workers=4, worker_batch_size=32, minimum_size_to_dequeue=32, force_download=False)`

Initializes the data loader with the chosen environment

#### Parameters

- **environment** (*string*) – desired MineRL environment
- **data\_dir** (*string, optional*) – specify alternative dataset location. Defaults to None.
- **num\_workers** (*int, optional*) – number of files to load at once. Defaults to 4.
- **force\_download** (*bool, optional*) – specifies whether or not the data should be downloaded if missing. Defaults to False.

**Returns** initialized data pipeline

**Return type** *DataPipeline*



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- modindex
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