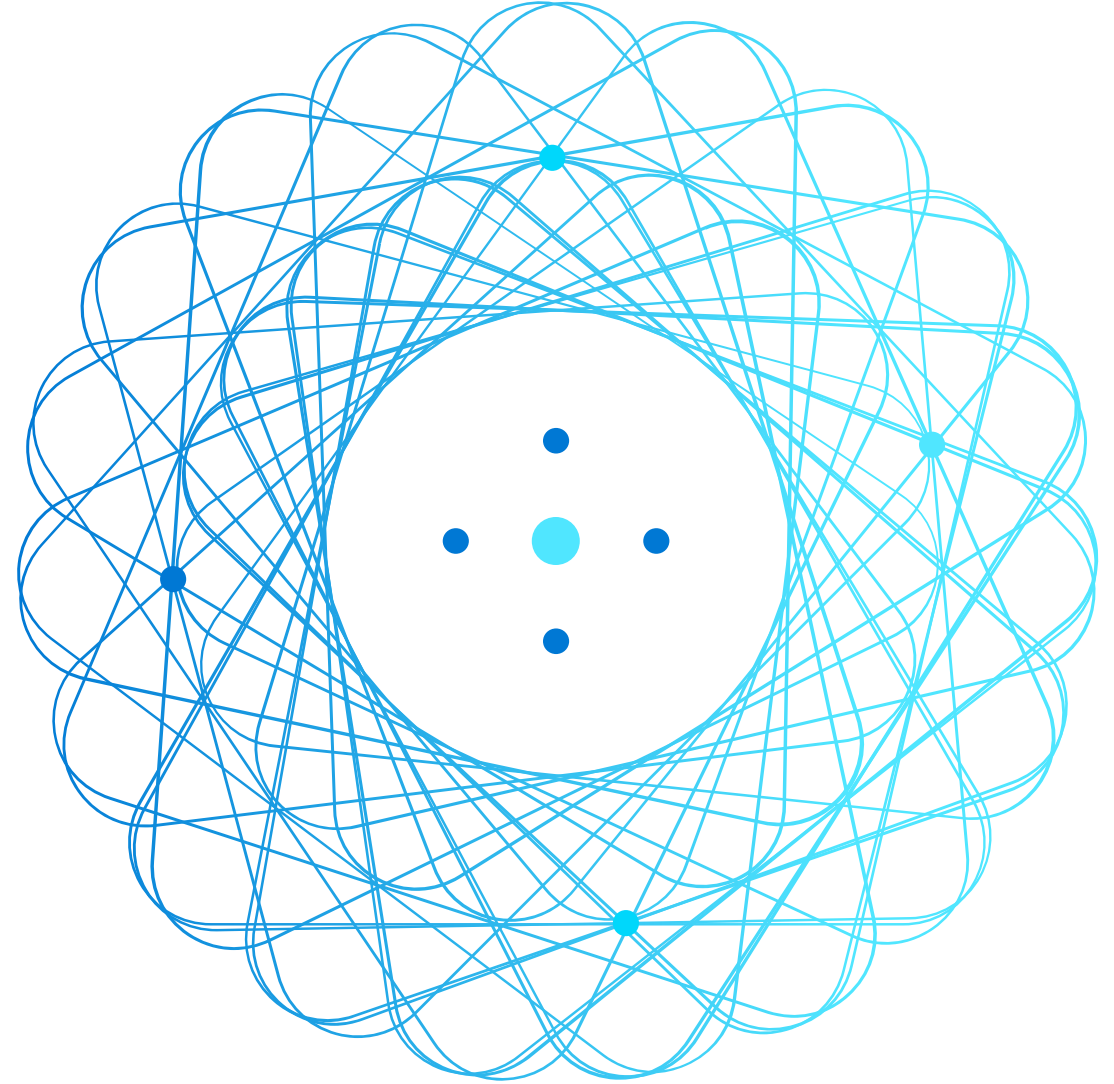


# Module 5: Working with Compute

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# Agenda



Environments



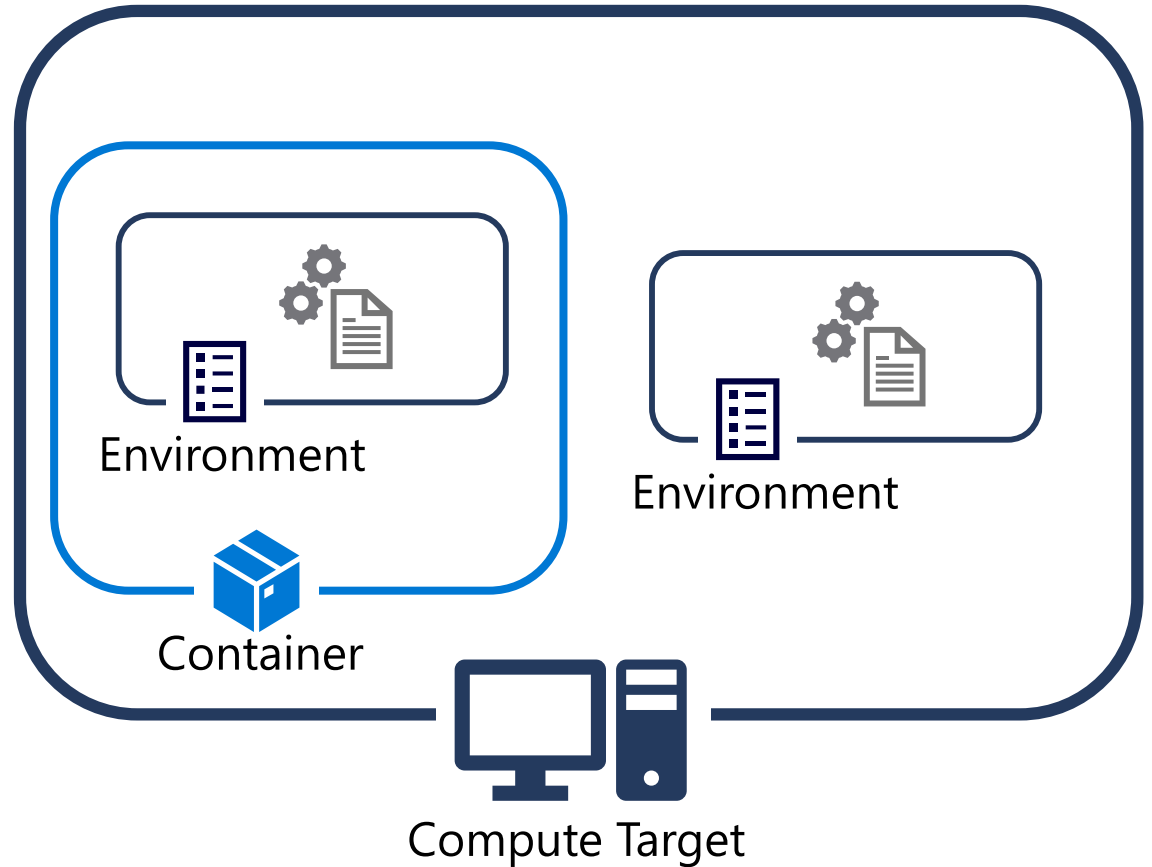
Compute Targets

# Environments



# Run Contexts for Experiments

- Python scripts run in a virtual *environment* that defines the Python version and installed packages
- The environment is usually (but not always) in a *container*
- The container (or environment) is hosted on a *compute target*
  - The default in most cases is the *local* compute (where the control code is run)



# Explicitly Creating Environments

## Create from specification file

```
env = Environment.from_conda_specification(name='training_environment',  
                                          file_path='./conda.yml')
```

File in standard YAML  
format for Conda  
environments

## Create from existing conda environment

```
env = Environment.from_existing_conda_environment(name='training_environment',  
                                                  conda_environment_name='py36')
```

Existing conda  
environment on  
local compute

## Create with specified packages

```
env = Environment('training_environment')  
deps = CondaDependencies.create(conda_packages=['scikit-learn', 'pandas', 'pip'],  
                               pip_packages=['azureml-defaults'])  
env.python.conda_dependencies = deps
```

Conda package installation is  
generally more efficient, so  
use it when possible

Most experiments  
require azureml-defaults

Use conda to install pip if  
you plan to also install  
pip packages

# Configuring Environment Containers

## Experiments run in Docker by default

```
docker_config = DockerConfiguration(use_docker=True)
script_config = ScriptRunConfig(..., docker_runtime_config=docker_config)
```

Use a DockerConfiguration class to explicitly use Docker (or not!)

## Use the *docker* section of the environment to use custom container images

```
env.docker.base_image='my-base-image'
env.docker.base_image_registry='myregistry.azurecr.io/myimage'
```

Override the default base image with your own prebuilt container image...

```
env.docker.base_image = None
env.docker.base_dockerfile = './Dockerfile'
```

...or create one from a dockerfile

## Override managed Python configuration

```
env.python.user_managed_dependencies=True
env.python.interpreter_path = '/opt/miniconda/bin/python'
```

If your image already includes Python and packages, manage dependencies yourself

# Registering and Reusing Environments

## Register an environment in the workspace

```
env.register(workspace=ws)
```

Saves a definition of the environment in the workspace for later use

## View Registered Environments

```
env_names = Environment.list(workspace=ws)
for env_name in env_names:
    print('Name:', env_name)
```

Azure Machine Learning provides a set of useful *curated* environments with names that begin "AzureML..."

## Retrieve and use an environment

```
training_env = Environment.get(workspace=ws, name='training_environment')

script_config = ScriptRunConfig(source_directory='my_dir',
                                script='script.py',
                                environment=training_env)
```

Enables you to reuse the environment on any compute target

Environment will be created if not already on compute target

# Compute Targets





# Compute Options for Experiment Runs



## Local Compute

- Compute where the control code for the experiment is running
- Often a development workstation or Azure Machine Learning compute instance



## Compute Cluster

- Cloud-based cluster managed in an Azure Machine Learning workspace
- Starts, stops, and scales on-demand



## Attached Compute

- Azure compute resource outside of a workspace
- For example:
  - Virtual Machine
  - Azure Databricks
  - Azure HDInsight

# Creating a Compute Cluster

Create in Azure Machine Learning studio

or

Use the SDK

```
from azureml.core.compute import ComputeTarget, AmlCompute

compute_name = 'aml-cluster'
compute_config = AmlCompute.provisioning_configuration(vm_size='STANDARD_DS11_V2',
max_nodes=4,
vm_priority='lowpriority')

aml_compute = ComputeTarget.create(ws, compute_name, compute_config)
aml_compute.wait_for_completion(show_output=True)
```

Specify a suitable Azure VM image  
(consider cores, memory, disk, GPU)

Cluster will  
scale up to  
this size as  
required

Low-priority or dedicated  
(low-priority can be pre-empted, causing  
runs to restart; dedicated is more  
expensive)

Additional options for virtual network and  
managed identity for access to other Azure  
resources

# Attaching Azure Databricks Compute

Create in Azure Machine Learning studio

or

Use the SDK

```
from azureml.core import Workspace
from azureml.core.compute import ComputeTarget, DatabricksCompute

compute_name = 'db_cluster'

db_workspace_name = 'db_workspace'
db_resource_group = 'db_resource_group'
db_access_token = '1234-abc-5678-defg-90...'
db_config = DatabricksCompute.attach_configuration(resource_group=db_resource_group,
                                                    workspace_name=db_workspace_name,
                                                    access_token=db_access_token)

databricks_compute = ComputeTarget.attach(ws, compute_name, db_config)
databricks_compute.wait_for_completion(True)
```


An existing Azure Databricks workspace in the same Azure subscription as the workspace

Generate a token in the Azure Databricks workspace and specify it here

# Using Compute Targets

Specify the compute target for an experiment

```
script_config = ScriptRunConfig(source_directory='my_dir',  
                                script='script.py',  
                                environment=env,  
                                compute_target=compute_name)
```



Specify the compute target name or object

# Lab: Work with Compute



1. View the lab instructions at <https://aka.ms/mslearn-dp100>
2. Complete the **Work with compute** exercise

# Knowledge check



You need to create an environment from a Conda configuration (.yaml) file.  
Which method of the *Environment* class should you use?

- ☐ create
  - ☒ create\_from\_conda\_specification
  - ☐ create\_from\_existing\_conda\_environment
- 



You need to run a training script on compute that scales on-demand from 0 to 3 GPU-based nodes.  
Which kind of compute target should you create?

- ☐ Compute Instance
  - ☒ Compute Cluster
  - ☐ Inference Cluster
- 



Which ScriptRunConfig parameter causes the script to run on a compute cluster named *train-cluster*?

- ☐ arguments=['--AmlCluster', 'train-cluster']
- ☐ environment='train-cluster'
- ☒ compute\_target='train-cluster'

# References

**Microsoft Learn: Work with Compute in Azure Machine Learning**

<https://docs.microsoft.com/learn/modules/use-compute-contexts-in-aml/>

**Azure Machine Learning environments documentation**

<https://docs.microsoft.com/azure/machine-learning/concept-environments>

**Azure Machine Learning compute targets documentation**

<https://docs.microsoft.com/azure/machine-learning/concept-compute-target>

**Microsoft Learn: Perform data science with Azure Databricks**

<https://docs.microsoft.com/learn/paths/perform-data-science-azure-databricks/>

