Cloudera Al Top Tasks (on premises) Map

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Creating and deploying a model

Using Cloudera AI, you can create any function within a script and deploy it to a REST API. In a Cloudera AI project, this is typically a predict function that accepts an input and returns a prediction based on the model's parameters.

As an example, we create a function that adds two numbers, and deploy it as a model that returns the sum of the numbers. This function will accept two numbers in JSON format as input and return the sum.



Note: In case of PBJ (Powered by Jupyter) Runtime, a specific decorator, the cml_model decorator is needed to create a model. This decorator allows a function to work as a model in a PBJ Runtime. The decorator can also be used to enable gathering of model metrics. For more details, see Example models with PBJ Runtimes.

For Cloudera AI UI

- 1. Create a new project. Note that models are always created within the context of a project.
- 2. Click New Session and launch a new Python 3 session.
- **3.** Create a new file within the project called add_numbers.py. This is the file where we define the function that will be called when the model is run. For example:

add_numbers.py

```
import cml.models_v1 as models
@models.cml_model
def add(args):
    result = args["a"] + args["b"]
    return result
```



Note: In practice, do not assume that users calling the model will provide input in the correct format or enter good values. Always perform input validation.

4. Before deploying the model, test it by running the add_numbers.py script, and then by calling the add function directly from the interactive workbench session. For example:

```
add({"a": 3, "b": 5})

File Edit View Navigate Run

Project >_Terminal access  Clear  interrupt  Stop Sessions  IIII

# Function to add two numbers

def add(args):
    result = args["a"] + args["b"]
    return result

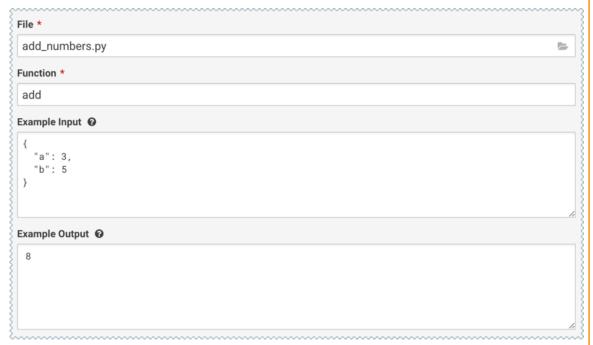
Untitled Session  Python 3 Session − 1 vCPU / 2 GiB Memory − 50 minutes ago

add({"a": 4, "b": 7})

8

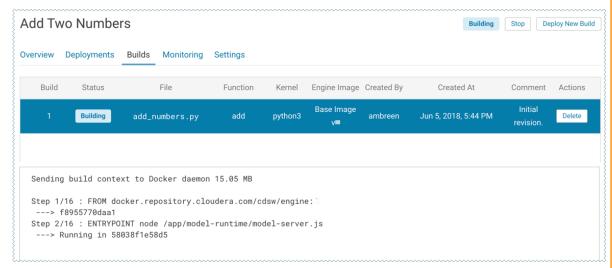
add({"a": 4, "b": 7})
```

- **5.** Deploy the add function to a REST endpoint.
 - a. Go to the project Overview page.
 - b. Click Model Deployments New Model.
 - c. Give the model a Name and Description.
 - **d.** In the **General** section of the page, under the Deploy Model as option, you can select the Service Account option, if the model is to be deployed in a service account, and choose the account from the dropdown menu.
 - e. Enter the details about the model that you want to build, for example:
 - File: add_numbers.py
 - · Function: add
 - Example Input: {"a": 3, "b": 5}
 - Example Output: 8



- **f.** Select the resources needed to run this model, including any replicas for load balancing. To specify the maximum number of replicas in a model deployment, go to Site Administration Settings Models. The default is 9 replicas, and up to 199 can be set.
- g. Click Deploy Model.

6. Select the model to go on its Overview page. Click Builds to track realtime progress as the model is built and deployed. This process essentially creates a Docker container where the model will live and serve requests.



7. Once the model has been deployed, go back to the model Overview page and use the Test Model widget to make sure the model works as expected.

If you entered example input when creating the model, the Input field will be pre-populated with those values. Click Test. The result returned includes the output response from the model, as well as the ID of the replica that served the request.

Model response times depend largely on your model code. That is, how long it takes the model function to perform the computation needed to return a prediction. Note that model replicas can only process one request at a time. Concurrent requests will be queued until the model can process them.

For Cloudera AI APIv2

Create and deploy a model using the Models API as instructed in this example:

This example demonstrates the use of the Models API. Run this example:

- 1. Create a project with the Python template and a legacy engine.
- 2. Start a session.
- 3. Run !pip3 install sklearn.
- **4.** Run fit.py.

The example script first obtains the project ID, then creates and deploys a model.

```
projects = client.list_projects(search_filter=json.dumps({"name": "<your</pre>
 project name>"}))
project = projects.projects[0] # assuming only one project is returned by
 the above query
model_body = cmlapi.CreateModelRequest(project_id=project.id, name="Demo
 Model", description="A simple model")
model = client.create_model(model_body, project.id)
model_build_body = cmlapi.CreateModelBuildRequest(project_id=project.id,
 model_id=model.id, file_path="predict.py", function_name="predict", ker
nel="python3")
model build = client.create model build(model build body, project.id, mod
el.id)
while model_build.status not in ["built", "build failed"]:
print("waiting for model to build...")
time.sleep(10)
 model_build = client.get_model_build(project.id, model.id, model_build
if model_build.status == "build failed":
```

```
print("model build failed, see UI for more information")
 sys.exit(1)
print("model built successfully!")
model_deployment_body = cmlapi.CreateModelDeploymentRequest(project_id=p
roject.id, model_id=model.id, build_id=model_build.id)
model_deployment = client.create_model_deployment(model_deployment_body,
project.id, model.id, build.id)
while model_deployment.status not in ["stopped", "failed", "deployed"]:
print("waiting for model to deploy...")
 time.sleep(10)
model_deployment = client.get_model_deployment(project.id, model.id, m
odel_build.id, model_deployment.id)
if model_deployment.status != "deployed":
print("model deployment failed, see UI for more information")
 sys.exit(1)
print("model deployed successfully!")
```

Registering a model using the Al Registry user interface

You can register a model using the **AI Registry** user interface or the MLFlow SDK.

Using the Al Registry user interface to register a model

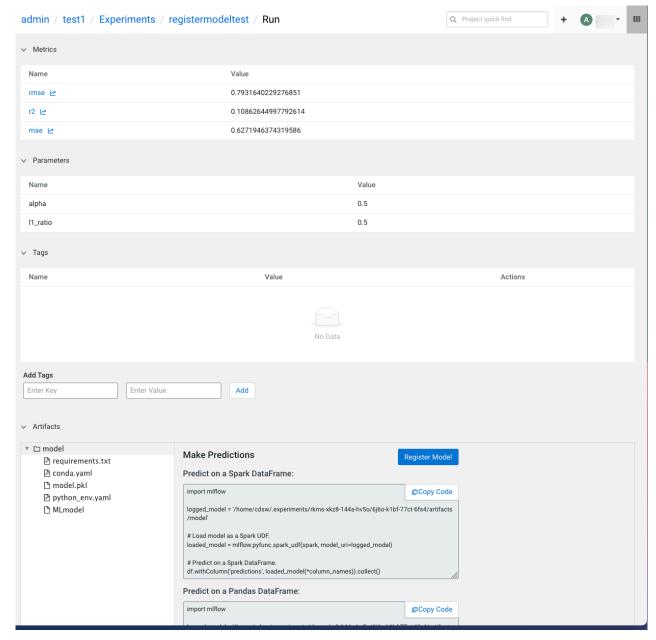
Registering a model enables you to track your model and upload and share the model. Registering a model stores the model archives in the Cloudera AI Registry with a version tag. The first time you register a model, **AI Registry** automatically creates a model repository with the first version of the model.

Before you begin

You must have permission to access a project in which the model is created before you can register the model.

Procedure

- 1. Click Projects in the left navigation pane to display the Projects page.
- 2. Select the project that contains the model that you want to register.
 - **AI Registry** displays all of the models under the specific project along with their source, deployment status, replicas, memory and a drop-down function for actions that can be made pertaining to that model for deployment.
- 3. Click the Experiments tab in the left navigation pane and select the experiment that contains the model you want to register.
- **4.** Select the model you want to register.
 - Cloudera AI displays the Experiment Run Details page.



- 5. Select the run that contains the model you want to register.
- **6.** Select Register Model to begin the registration process.
 - **AI Registry** displays the Registry Model dialog box.
- **7.** Enter the name of your registered model.

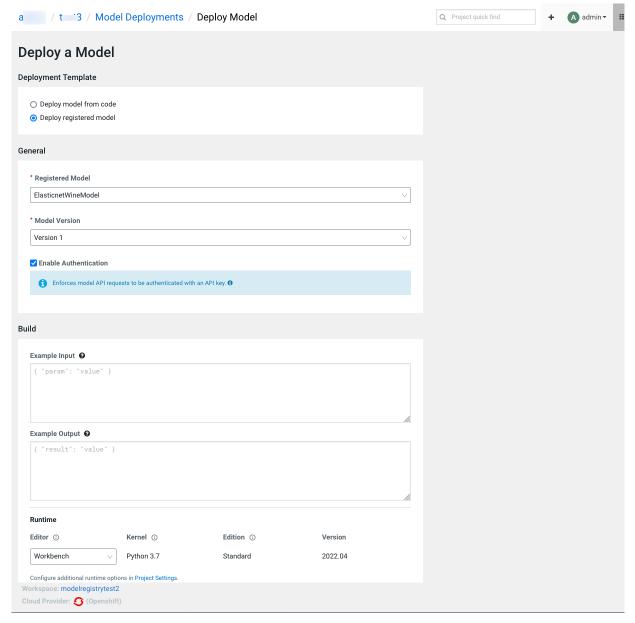
 You can also enter optional information for the description, version notes, and version tags.
- **8.** Click OK to complete the registration.

Deploying a model from the AI Registry page

You can deploy a model once or more times to create different versions of the model. You can also deploy a model you created in one workbench to a different workbench.

Procedure

- 1. Select AI Registry from the left navigation pane.
- **2.** Select the model you want to deploy.
 - AI Registry displays the Model Version List page.
- **3.** Select the model version you want to deploy.
 - AI Registry displays a side window that lists the version information. Dismiss this window to proceed.
- 4. Under the Actions menu, click Deploy.
- 5. Select the project you want to deploy to in the dialog box and click Go.
 - You can select either the project the model is located in or another project to deploy the model to.
 - AI Registry displays the Deploy a Model page with the detailed model information auto populated.



- **6.** If you enable authentication and you have deployed the model to a shared project, you will need to enter an API key to be able to access and use the model.
- 7. Click OK.