# **Quickstart**

Setup

Get up and running with Actions! In this quickstart you will learn the basics to get up and running with Actions and Giza Platform. You will create a Giza Workspace, initialize your first Action, generate a deployment and create executions with Actions Runs

Before you begin, make sure you have all the necessary libraries installed:
Copy pipinstallgiza-actions pipxinstallgiza-cli
From your terminal, create a Giza user through our CLI in order to access the Giza Platform:
Copy gizauserscreate
After creating your user, log into Giza:
Copy gizauserslogin
Optional: you can create an API Key for your user in order to not regenerate your access token every few hours.
Copy gizauserscreate-api-key
To create Actions Runs you will need a Giza Workspace, create it executing the following command in your terminal:
Copy gizaworkspacescreate
Create your first Action
You can use our Giza CLI to initialize an Action project:
Copy gizaactionsnew
After submitting the required parameters you will get a directory with the following structure:
After submitting the required parameters you will get a directory with the following structure:
Copy gizaactionsnewmnist_example cdexample tree

This structure represents a ready-to-use Python project to work with Giza Actions based or MNIST Tutorial.

An example of how to define an action can be found inmnist example/train action.py. This script is responsible for creating your Action (workflow) for training your neural network and deploying the action on your GizaWorkspace.

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## mnist example/train action.py

importnumpyasnp importtorch importtorch.nnasnn importtorch.onnx importtorch.optimasoptim importtorchvision fromgiza actions.actionimportAction,action fromgiza actions.taskimporttask fromscipy.ndimageimportzoom fromtorch.utils.dataimportDataLoader,TensorDataset

device=torch.device("cuda"iftorch.cuda.is\_available()else"cpu")

## **Neural Network Definition**

optimizer.zero\_grad() loss.backward() optimizer.step()

```
input size=196# 14x14 hidden size=10 num classes=10 num epochs=10 batch size=256 learning rate=0.001
```

classNeuralNet(nn.Module): def\_\_init\_\_(self,input\_size,hidden\_size,num\_classes): super(NeuralNet, self).init() self.input size=input size self.11=nn.Linear(input size, hidden size) self.relu=nn.ReLU() self.12=nn.Linear(hidden size, num\_classes)

defforward(self,x): out=self.l1(x) out=self.relu(out) out=self.l2(out) returnout

```
Task Definitions
defresize images(images): returnnp.array([zoom(image[0], (0.5,0.5))forimageinimages])
@task(name="Prepare Datasets") defprepare datasets(): print("Prepare dataset...")
train dataset=torchvision.datasets.MNIST(root="./data", train=True, download=True)
test dataset=torchvision.datasets.MNIST(root="./data", train=False)
x train=resize images(train dataset) x test=resize images(test dataset)
x_train=torch.tensor(x_train.reshape(-1,14*14).astype("float32")/255) y_train=torch.tensor([labelfor_, labelintrain_dataset],
dtype=torch.long)
x_test=torch.tensor(x_test.reshape(-1,14*14).astype("float32")/255) y_test=torch.tensor([labelfor_, labelintest_dataset],
dtype=torch.long)
print(" Datasets prepared successfully")
returnx train,y train,x test,y test
@task(name="Create Loaders") defcreate data loaders(x train,y train,x test,y test): print("Create loaders...")
train loader=DataLoader( TensorDataset(x train, y train), batch size=batch size, shuffle=True ) test loader=DataLoader(
TensorDataset(x test, y test), batch size=batch size, shuffle=False)
print(" Loaders created!")
returntrain loader, test loader
@task(name="Train model") deftrain_model(train_loader): print("Train model...")
model=NeuralNet(input size, hidden size, num classes).to(device) criterion=nn.CrossEntropyLoss()
optimizer=optim.Adam(model.parameters(), lr=learning rate)
forepochinrange(num_epochs): fori,(images,labels)inenumerate(train_loader): images=images.to(device).reshape(-1,14*14)
labels=labels.to(device)
outputs=model(images) loss=criterion(outputs, labels)
```

if(i+1)%100==0: print(f"Epoch [{epoch+1}/{num\_epochs}], Step [{i+1}/{len(train\_loader)}], Loss:{loss.item():.4f}")

print(" Model trained successfully") returnmodel

@task(name="Test model") deftest\_model(model,test\_loader): print("Test model...") withtorch.no\_grad(): n\_correct=0 n\_samples=0 forimages,labelsintest\_loader: images=images.to(device).reshape(-1,14\*14) labels=labels.to(device) outputs=model(images) \_,predicted=torch.max(outputs.data,1) n\_samples+=labels.size(0) n\_correct+= (predicted==labels).sum().item()

acc=100.0\*n\_correct/n\_samples print(f"Accuracy of the network on the 10000 test images:{acc}%")

@task(name="Convert To ONNX") defconvert\_to\_onnx(model,onnx\_file\_path): dummy\_input=torch.randn(1, input\_size).to(device) torch.onnx.export( model, dummy\_input, onnx\_file\_path, export\_params=True, opset\_version=10, do constant folding=True, )

print(f"Model has been converted to ONNX and saved as{onnx\_file\_path}")

@action(name="Action: Convert To ONNX", log\_prints=True) defexecution(): x\_train,y\_train,x\_test,y\_test=prepare\_datasets() train\_loader,test\_loader=create\_data\_loaders(x\_train, y\_train, x\_test, y\_test) model=train\_model(train\_loader) test\_model(model, test\_loader)

### Convert to ONNX

```
onnx_file_path="mnist_model.onnx" convert_to_onnx(model, onnx_file_path)

if__name__=="main": action_deploy=Action(entrypoint=execution, name="pytorch-mnist-action")

action_deploy.serve(name="pytorch-mnist-deployment")

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Now you are set to create your first Action deployment! Execute the Action file to deploy it.

...
```

Copy python mnist\_example/train\_action.py

...

Your deployment is ready for executions! You can create Actions Runs directly from the GizaWorkspace. To learn more aboutcreating an Action Run check our linked guided overview.

To deploy the other actions in themnist\_example directory in your workspace, you must first create a model on the Giza platform. To better understand the process, please refer to our step-by-step tutorial.

What's next?

Congrats! Now that you've completed the Actions quickstart, check out our guides and learn how to do more specific things in our<u>how-to-guides</u> and<u>tutorials</u>. If you're interested in learning more about Actions core concepts, grab a cup of coffee and take a look at our<u>Conceptual Guides</u>!

Previous Installation Next Contribution Guidelines

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