

Tools for Deep Learning Scenarios



DVC Tools for Data Scientists & Analysts



Coursé lessons

- **Lesson 1.** Course Introduction
- **Lesson 2.** Practices and Tools for Efficient Collaboration in ML projects
- Lesson 3. Pipelines Automation and Configuration Management
- **Lesson 4.** Versioning Data and Models
- **Lesson 5.** Visualize Metrics & Compare Experiments with DVC and Studio
- **Lesson 6.** Experiments Management and Collaboration
- **Lesson 7.** Tools for Deep Learning Scenarios
- **Lesson 8.** Review Advanced Topics and Use Cases





- DVCLive and checkpoints in deep learning scenarios
- Setting up checkpoints
- Experimenting and collaboration



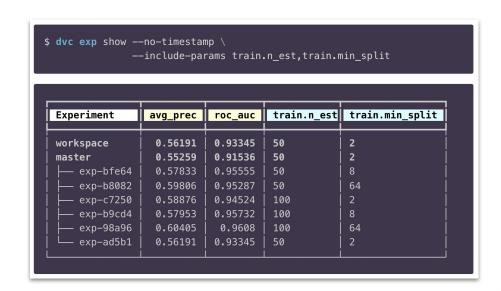


DVCLive and Checkpoints in Deep Learning

DVC tools for Deep Learning



- Save your model weights at checkpoints
- Log metrics and results
- Track all code and data changes
 - update params
 - queueing experiments
 - comparing many experiments
 - persisting experiments
 - sharing experiments



Setting up a DVC pipeline: dvc stage add



```
Defines the checkpoint file or
                                            directory
dvc stage add --name train
     --deps data/MNIST --deps train.py \
     --checkpoints model.pt \
     --plots-no-cache predictions.json \
     --params seed, Ir, weight_decay \
     --live dvclive \
     python train.py
```

Enables **DVCLive** logger, which helps to register checkpoints from your code.

Setting up a DVC pipeline: update dvc.yaml

```
stages:
 train:
   cmd: python train.py
                                                                 Or, update dvc.yaml
   deps:
                                                                      accordingly
     data/MNIST
     - train.py
   params:
     - 1r
     - seed
     weight_decay
   outs:
     - model.pt:
         checkpoint: true
   live:
     dvclive:
       summary: true
         html: true
```

Registering checkpoints in Python code

```
import dvclive
                                                                   Log metrics and save
                                                                  model for each training
# Iterate over training epochs.
                                                                            epoch
for i in range(1, EPOCHS+1):
    # Train model
    # Save model
    torch.save(model.state_dict(), "model.pt")
    # Evaluate and checkpoint.
    metrics = evaluate(model, x_test, y_test)
    for k, v in metrics.items():
      print('Epoch %s: %s=%s'%(i, k, v))
      dvclive.log(name=k, val=v)
    dvclive.next_step()
```

Running experiments



run command

\$ dvc exp run

output

```
Epoch 1: loss=1.9428282976150513
```

Epoch 1: acc=0.5715

Generating lock file 'dvc.lock'

Updating lock file 'dvc.lock'

Checkpoint experiment iteration 'd99d81c'.

file:///Users/milecia/Repos/checkpoints-tutorial/dvclive.html

Epoch 2: loss=1.25374174118042

Epoch 2: acc=0.7738

..

DVC saves checkpoints and publishes metrics updates in dvclive.html file for each epochs

Running experiments & metrics tracking



run command

\$ dvc exp show

	!			!
Experiment	Created	step	acc	loss
workspace	-	4	0.92623	0.19567
main	01:58 PM	-	_	-
_{bf81637 [exp-a1f53]}	02:05 PM	4	0.92623	0.19567
	02:04 PM	3	0.89344	0.27423
	02:03 PM	2	0.87295	0.29018
	02:02 PM	1	0.89754	0.26993
a95260d	02:01 PM 	0	0.73361	0.5271



Live code example

Setting up DVCLive and Checkpoints

Experimenting and collaboration with Studio



Live code example

Experimenting and collaboration with Studio

What have we learned?

What have we learned?



- DVCLive and checkpoints in deep learning scenarios
- 2. Setting up checkpoints
- **3.** Experimenting and collaboration



Links



- Data Science blueprint
 https://data-science-blueprint.readthedocs.io/en/latest/presentation/schema.html
- Post: Using Experiments for Transfer Learning
 https://dvc.org/blog/transfer-learning-experiments
- Post: Adding Data to Build a More Generic Model
 https://dvc.org/blog/adding-data-to-build-a-more-generic-model
- Docs: Checkpoints
 https://dvc.org/doc/user-quide/experiment-management/checkpoints