Data Versioning Motivation

INTRODUCTION TO DATA VERSIONING WITH DVC



Ravi Bhadauria Machine Learning Engineer



What is Data Versioning?

- Definition
 - Monitors data changes over time
 - Snapshots data over iterations
 - Similar to code versioning
- Benefits
 - Retrieval and scruting
 - Data consistency, accountability, and lineage
- Applications
 - Data Science and Machine Learning
 - Data Engineering
 - Financial Analysis, Auditing and Compliance

Data vs Code Versioning

Code Versioning

- Well known in software development
- Uses tools like Git to do decentralized version control
- Easier to manage as codebases are small

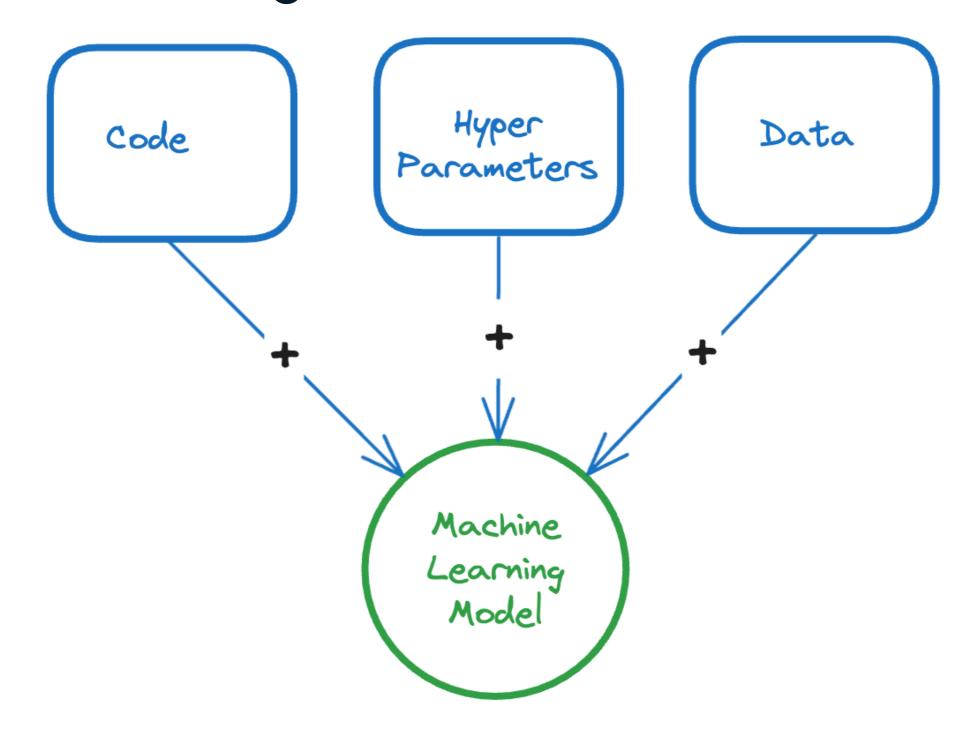
Data Versioning

- Relatively new (SciDB proposed in 2012)
- Toolchains like DVC are used in conjunction with Git
- Relatively difficult to manage due to large dataset size

¹ doi: 10.1109/ICDE.2012.102



Why Data Versioning in ML?



Dataset influence

Dataset A

	Booking_ID	number of adults	number of children	number of weekend nights	•••	booking status
0	INN32220	2	0	1		0
1	INN19707	2	0	0		0
2	INN36276	2	0	1		0
3	INN02246	2	0	0		1
4	INN27306	1	0	0		0

Dataset B

	Booking_ID	number of adults	number of children	number of weekend nights	•••	booking status
0	INN33507	2	0	0		0
1	INN29646	1	0	0		0
2	INN34621	2	0	2		0
3	INN19236	2	0	0		0
4	INN02991	2	0	0		0

Dataset influence

Hyperparameters are kept consistent, dataset changed

Metric	Dataset A	Dataset B
Precision	0.78	0.79
Recall	0.54	0.57
F1 Score	0.64	0.66
Accuracy	0.80	0.81

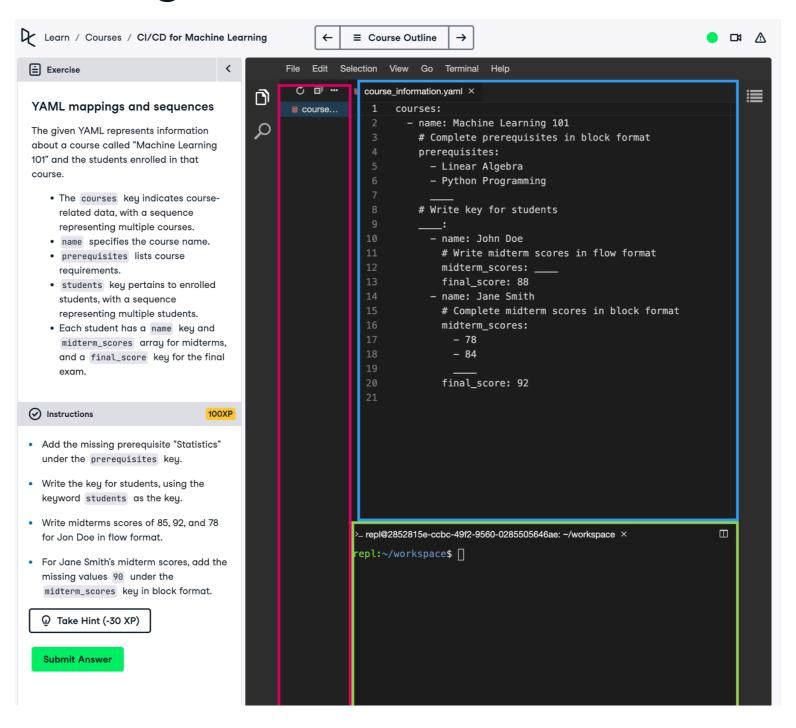


Hyperparameters influence

Dataset kept consistent, hyperparameters changed

Metric	n_estimators=5	n_estimators=10
Precision	0.78	0.85
Recall	0.54	0.52
F1 Score	0.64	0.65
Accuracy	0.80	0.81

Editor Exercises Layout





Let's practice!

INTRODUCTION TO DATA VERSIONING WITH DVC



Introduction to DVC

INTRODUCTION TO DATA VERSIONING WITH DVC

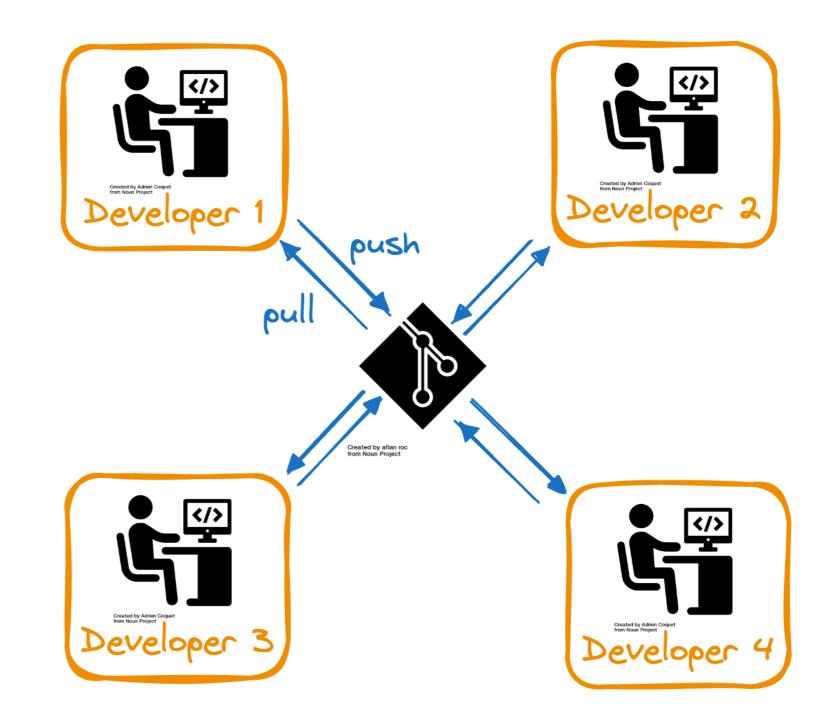


Ravi Bhadauria Machine Learning Engineer



Git as Version Control

- Code version control system
- Independent local development
 - Branch and merge
 - Version history management
- Enables collaboration



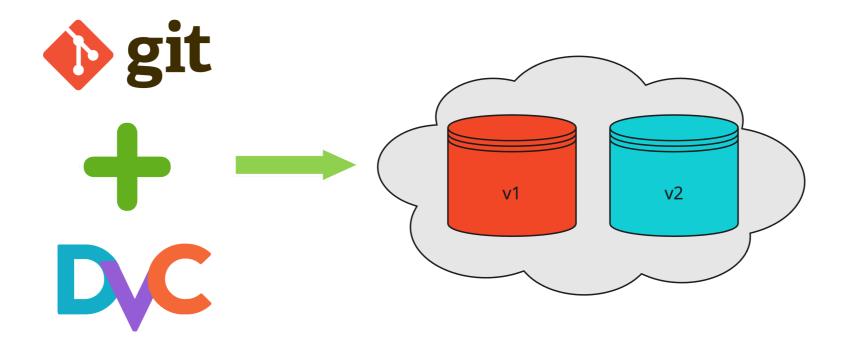
Git as Version Control

- CLI based interaction
- Run on *terminal*, aka *shell*
- Git tracks contents via a repository
 - Actual files/folders to be tracked
 - Git metadata (in .git folder)

```
Git Repository
$ tree -aL 2
    .git
        HEAD
        config
        description
        hooks
        info
        objects
      - refs
    code.py
    data
    └─ mydata.csv
```

Data Version Control (DVC)

- DVC: Data Version Control tool
 - Manages data and experiments
 - Similar to Git



• Git tracks metadata, DVC handles data versioning

Git vs DVC CLI

Git

Initialize repository in working folder

```
$ git init
```

Adding files to repository (staging changes)

```
$ git add code.py
```

Commit changes (in version history)

```
$ git commit -m "adding first file"
```

DVC

Initialize DVC repository in working folder

```
$ dvc init
```

Adding data files to DVC

```
$ dvc add data/mydata.csv
```

Updating all tracked data files

```
$ dvc commit
```

Git vs DVC CLI

Git

Push code changes to remote server

```
$ git push
```

Pulling changes from remote

```
$ git pull
```

 Cloning an existing repository from remote (Github)

```
$ git clone \
https://github.com/username/repository-name.git
```

DVC

Push data changes to remote data server

```
$ dvc push
```

Synchronizing your DVC project

```
$ dvc pull
```

 Download a file or directory tracked by DVC

```
$ dvc get \
https://github.com/username/repo-name model.pkl
```

Let's practice!

INTRODUCTION TO DATA VERSIONING WITH DVC



DVC features and use cases

INTRODUCTION TO DATA VERSIONING WITH DVC



Ravi Bhadauria Machine Learning Engineer



DVC features and use cases

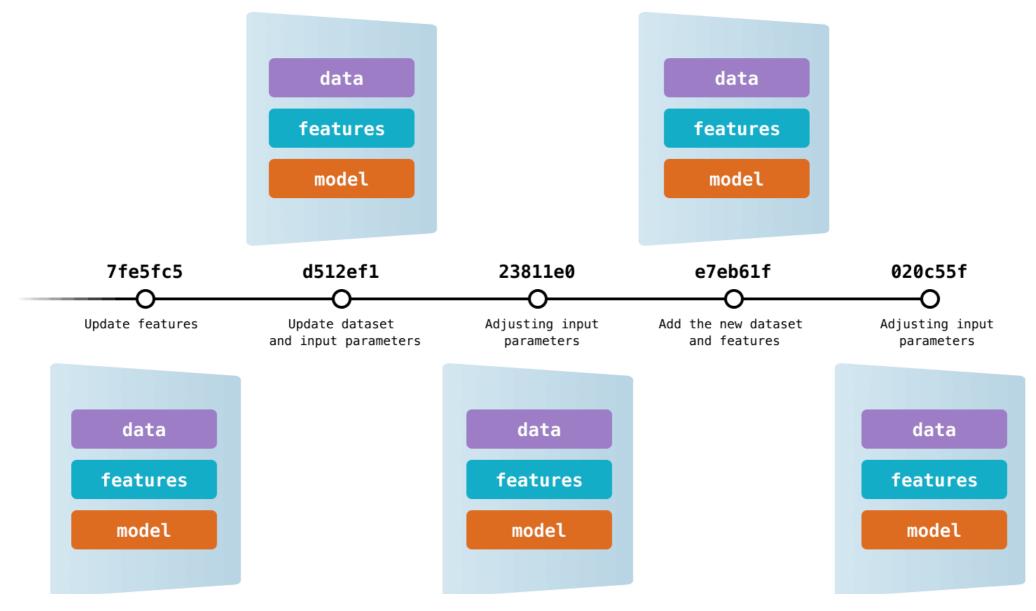
Covered topics

- Versioning data and models
- DVC Pipelines
- Metrics and plots tracking

Advanced topics (not covered)

- Experiment tracking
- CI/CD for machine learning
- Data registry

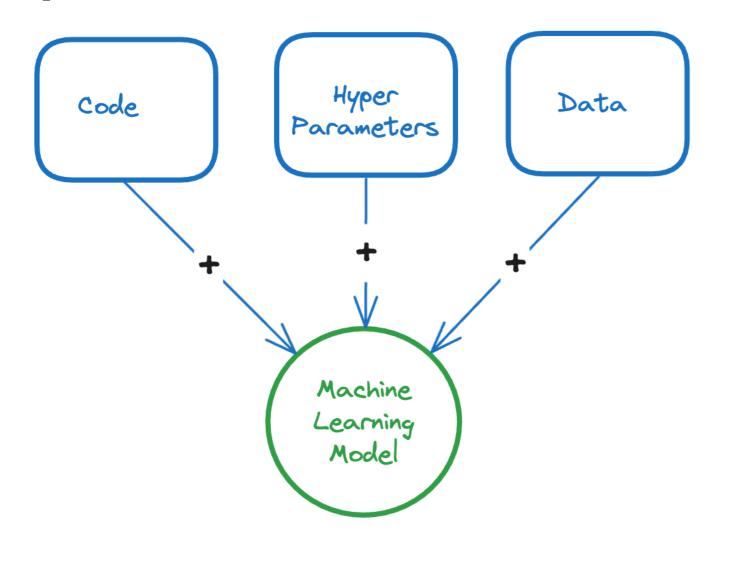
Versioning data and models



¹ https://dvc.org/doc/use-cases/versioning-data-and-models



Pipelines



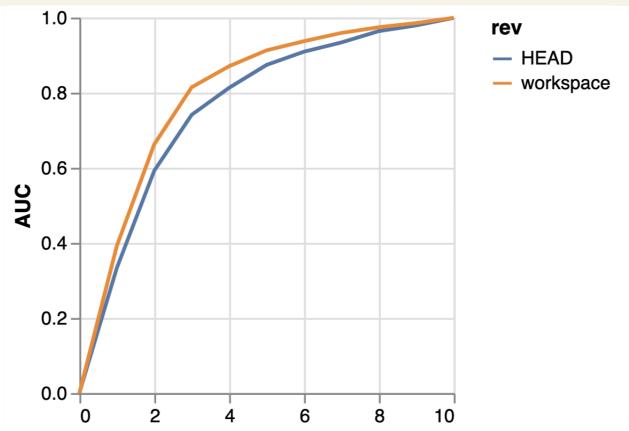
• Define pipeline in dvc.yaml

```
stages:
  train:
    cmd: python train.py
    deps:
      - code/train.py
      - data/input_data.csv
      - params/params.json
    outs:
      - model_output/model.pkl
```

• Run with dvc repro.

Tracking metrics and plots

```
$ dvc metrics diff
Path
                       Metric
                                 HEAD
                                          workspace
                                                        Change
dvclive/metrics.json
                      AUC
                                 0.78912
                                          0.18114
                                                        -0.60798
dvclive/metrics.json
                       TP
                                 215
                                          768
                                                        553
```



¹ https://dvc.org/doc/command-reference/plots/diff



Experiment tracking

- Run experiment and log metrics
 - o dvc repro
 - o dvc exp save
- Alternatively, combine two steps dvc exp run
- Experiments are custom Git references
 - Prevent bloating up Git commits
 - Explicit saves can be made with dvc exp save
- Visualize using dvc exp show

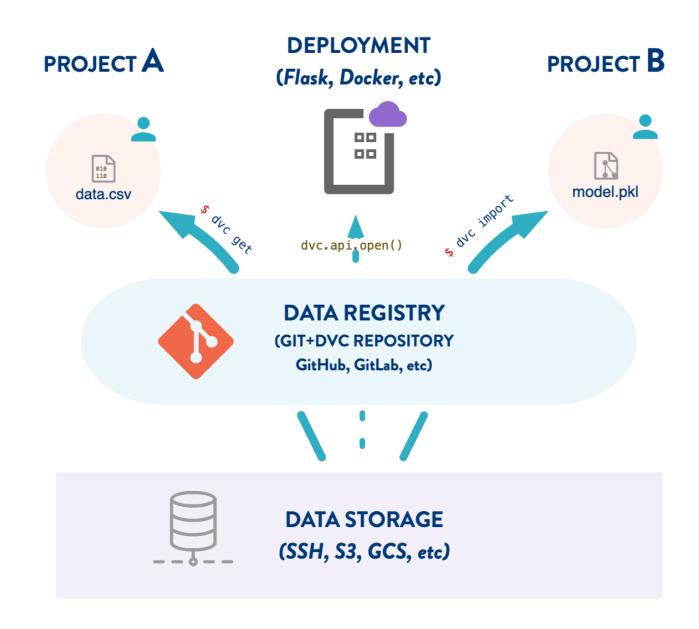
CI/CD for Machine Learning



¹ Picture credits: https://dvc.org/doc/use-cases/ci-cd-for-machine-learning



Data registry



¹ Picture credits: https://dvc.org/doc/use-cases/data-registry



Let's practice!

INTRODUCTION TO DATA VERSIONING WITH DVC

