ML Pipelines, Reproducibility and Experimentation

Alex Kim @alex000kim

We have a Jupyter notebook...



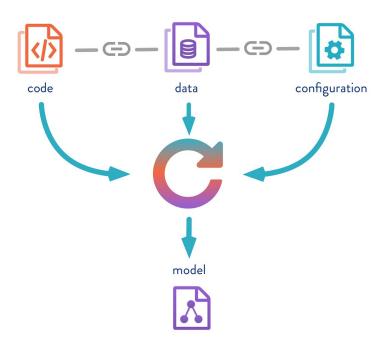
- Data loading
- Feature Engineering
- Model Training
- Model Evaluation

Can you easily answer these questions?

- What exactly was used to produce a particular model?
- Can you easily compare many ML experiments?
- Will you be able to reproduce them later?

Goal #1: Achieve best performance

- Running many experiments
- Experiment = a particular combination of
 Code & Data & Config



Goal #2: Ensure reproducibility

- Improving model performance: you can't improve what you can't reproduce
- Transparency and team collaboration: know everything your team members did to achieve certain performance

Goal #3: Minimal setup and dependency of 3rd party services

Problems:

- Vendor lock-in: instrument code with framework-specific code
- Maintenance & cost: maintain your own ML tracking server (or pay them to take care of it)
- Security concerns: send data to an external service or database

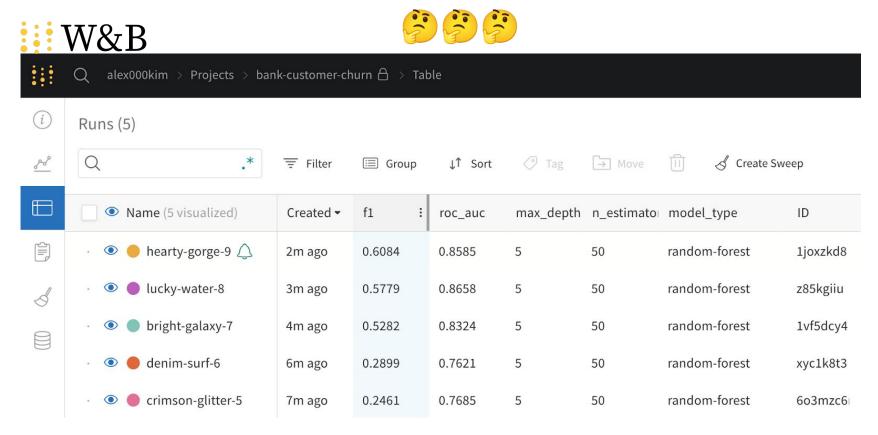
Most ML tracking solutions (MLflow, W&B, comet.ml, etc) have at least 2 of these problems

Fact: It's difficult to achieve all three goals

Can we do all of the following?

- Iterate quickly i.e. generate many experiments
- 2. Automatically track **all** changes to code, configs and data
- Avoid dependency on 3rd party services to store data, metrics and params

Same experiments, but different metrics?



Reproducibility VS. Experimentation?



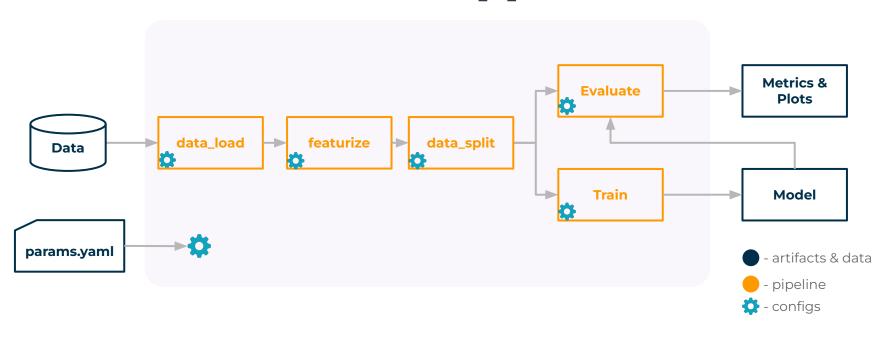
- DVC pipelines for generating many experiments
- Achieve complete reproducibility by versioning everything!
 - code and configs --> Git
 - dataset, models, other artifacts --> DVC remote storage (cloud buckets, NAS, SFTP, etc)
- VS Code as a convenient UI for experiment management
- No need to maintain (or pay for) additional services







What are DVC pipelines?

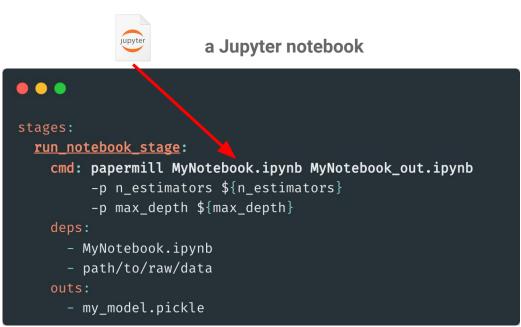


What are DVC pipelines? **Metrics & Evaluate Plots** data_load featurize data_split Data **Train** Model params.yaml - artifacts & data - pipeline - configs OR

DVC pipeline (defined in dvc.yaml) as:

a sequence of Python modules

```
data preprocessing stage:
   cmd: python process_data.py
      - process_data.py
     - path/to/raw/data
      - path/to/processed/data
      - preprocessing params
  train stage:
    cmd: python train.py
     - train.py
     - path/to/processed/data
      - my_model.pickle
      - train params
  eval stage:
    cmd: python eval.py
     - eval.pv
     - path/to/processed/data
      - my model.pickle
      - eval params
```



Run an experiment

```
$ dvc exp run -S train.params.n_estimators=120
'data/Churn_Modelling.csv.dvc' didn't change, skipping
Running stage 'run notebook':
> papermill TrainChurnModel.ipynb TrainChurnModel_out.ipynb -p n_estimators 120 -p max_depth 10 -p model_type lightgbm
Input Notebook: TrainChurnModel.ipynb
Output Notebook: TrainChurnModel_out.ipynb
Black is not installed, parameters wont be formatted
Executing: 0%|
                                                                                                                      0/27
[00:00<?, ?cell/s]Executing notebook with kernel: python3
Executing: 100%
                                                                                                                       27/27
[00:03<00:00, 7.58cell/s]
Updating lock file 'dvc.lock'
```

Track and manage many experiments

Ä Experiments ×	Me	Metrics		Parameters			
En un audien austi			tra	in	base	train	
Experiment	metri	metrics.json		params			
Experiment	↓ f	1 roc_auc	n_estimators	max_depth	random_st	model_type	
workspace	0.5639	1 0.83224	150	20	42	lightgbm	
☐ ☆ ✔ • one-stage-pi	peli 0.53850	6 0.84736	30	10	42	lightgbm	
→ dc21d5d o [exp-e9ada]	0.5639	1 0.83224	150	20	42	lightgbm	
■ ★ • 20894cb [exp-2d337]	0.5568	3 0.84207	100	15	42	lightgbm	
○ e83077e ○ [exp-03968]	0.5533	0.83767	100	20	42	lightgbm	
525985d [exp−4ec26]	0.5521	6 0.84408	50	10	42	lightgbm	
○ 9985cd0 [exp-da078]	0.5519	0.83779	150	15	42	lightgbm	
● 9dd91be [exp-d9c06]	0.5508	4 0.84411	50	15	42	lightgbm	
■ ★ • 16f1028 [exp-d4c6c]	0.5492	2 0.84241	50	20	42	lightgbm	
	0.5486	0.83717	150	10	42	lightgbm	
b9e2ff4 [exp-0db86]	0.5404	0.84086	100	10	42	lightgbm	

Practice time!

