### The Model Review

improving transparency, reproducibility & knowledge sharing with MLflow



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ML Engineer at Cash App

#### **About Me**

- Snowboarder 1st 🏂
- Originally from Alaska, currently living in Salt Lake City
- PhD in Physics 🔭
- Machine Learning Engineer (Modeler) at Cash App, Block
- Working on Natural Language Understanding for Customer Support

#### About this talk

- Why did my team decide to adopt a process for Model Review?
- What does Model Review mean?
- Our primary tool for review: MLflow
- Intro to MLflow
- How we are using MLflow to solve some of our problems

#### **Motivation**

Our team had recently doubled in size and was deploying lots of models.

But we had no good record keeping on what models exactly were in production, how they were trained, or what tricks or approaches were working well...

### **Motivation**

What if a model needed to be retrained?

A new team member wants to build off the work that came before them?

What was the precision on that model supposed to be anyway?

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A new team member wants to build off the work that came before them?

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uhhh, let me see if I can find that notebook...

#### Goals for a new Model Review Process

- 1. Transparency and record keeping of what exactly is being deployed
- 2. **Reproducibility** of past experiments and ease of building off of them
- 3. **Knowledge Sharing** so we can learn from each other and new teammembers can get up to speed

#### Goals for a new Model Review Process

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ALSO: We need to **automate** as much of this as possible!

# **Comparison to Code Review**

## Why do we Review Code?

- More •• on code to spot bugs and potential issues
- Pull Requests create a record of changes/commits and also (ideally) documentation of changes, design decisions, trade-offs, etc.
- Knowledge sharing between teammates

## Code vs Model Review Similarities

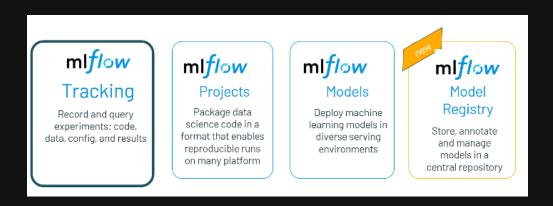
There is code involved.

### Code vs Model Review Differences

- There is a lot of context beyond the Code Used to train a model
  - data, including any transformations
  - performance
  - entire ML process, including all your failed experiments
- We can't really review a model just by reading the final code
- Where to record async review comments? (GitHub not really a good fit for this)

#### What is MLflow?

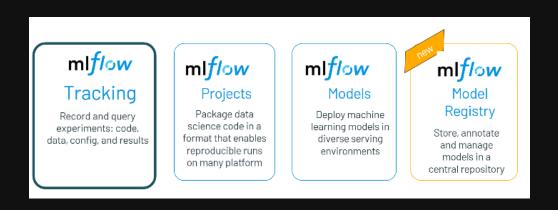
"Open source platform for managing the end-to-end machine learning lifecycle"



Language & library agnostic: includes APIs for Python, R, Java, but everything is accessible throuh a REST API & CLI, so it's very flexible for difference use cases.

#### What is MLflow?

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Language & library agnostic: includes APIs for Python, R, Java, but everything is accessible throuh a REST API & CLI, so it's very flexible for difference use cases.

\$ pip install mlflow

## **MLflow Tracking**

Easily log almost anything you want to keep track of:

- parameters
- metrics
- arbitrary files ("artifacts" in mlflow)
  - such as plots, output files, Jupyter notebooks...
- code version

## First Example

```
In [1]:
    import mlflow

mlflow.log_param('my_parameter', 4) # run starts automatically
    mlflow.log_metric('score', 100)
    mlflow.end_run()
```

#### **MLflow Tracking**



## **MLflow Tracking UI**

```
$ mlflow ui

[2021-11-26 10:15:48 -0700] [19408] [INFO] Starting gunicorn 20.1.0

[2021-11-26 10:15:48 -0700] [19408] [INFO] Listening at: http://127.0.0.1:5000
(19408)

[2021-11-26 10:15:48 -0700] [19408] [INFO] Using worker: sync

[2021-11-26 10:15:48 -0700] [19411] [INFO] Booting worker with pid: 19411
```

## MLflow Tracking UI

```
$ mlflow ui

[2021-11-26 10:15:48 -0700] [19408] [INFO] Starting gunicorn 20.1.0

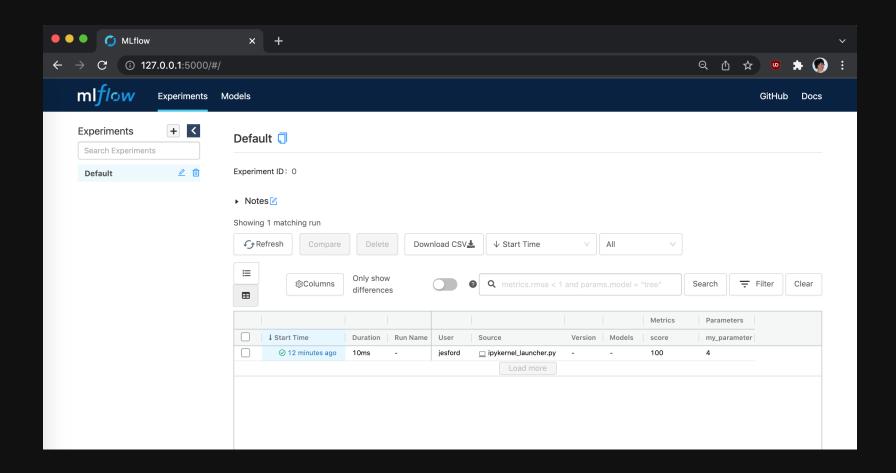
[2021-11-26 10:15:48 -0700] [19408] [INFO] Listening at: http://127.0.0.1:5000

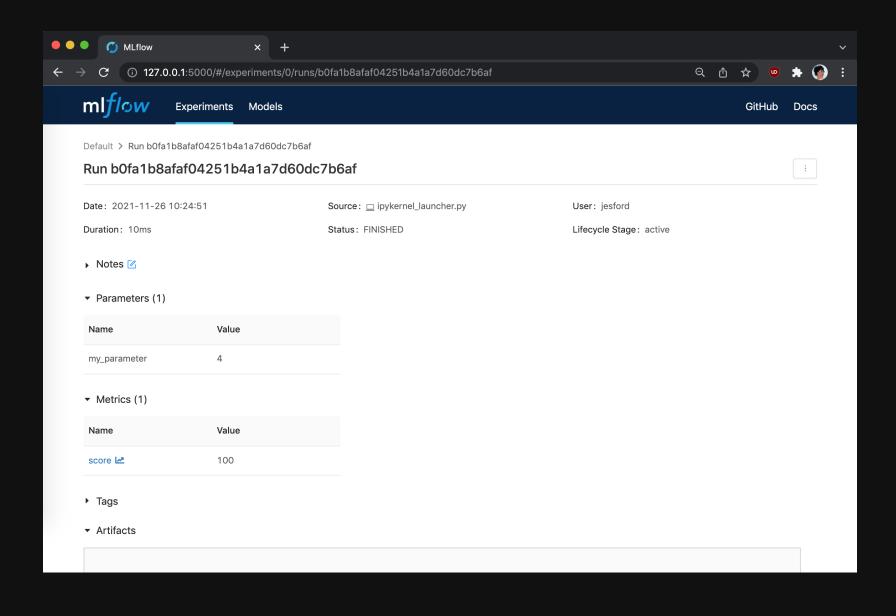
(19408)

[2021-11-26 10:15:48 -0700] [19408] [INFO] Using worker: sync

[2021-11-26 10:15:48 -0700] [19411] [INFO] Booting worker with pid: 19411
```

 $\rightarrow$  Go to http://127.0.0.1:5000 in your browser...





## **Logging Artifacts**

```
In [2]:
    import mlflow

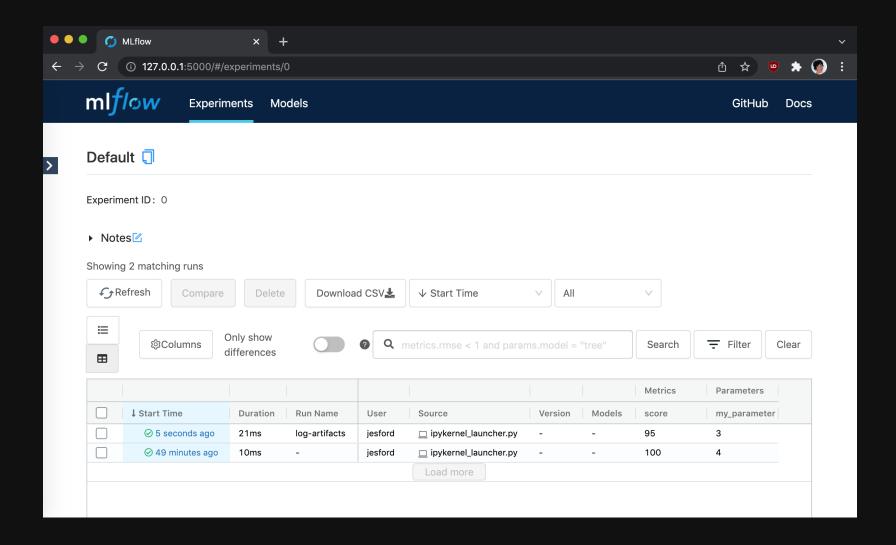
# explicitly start the run to give it a nice name
    mlflow.start_run(run_name='log-artifacts')

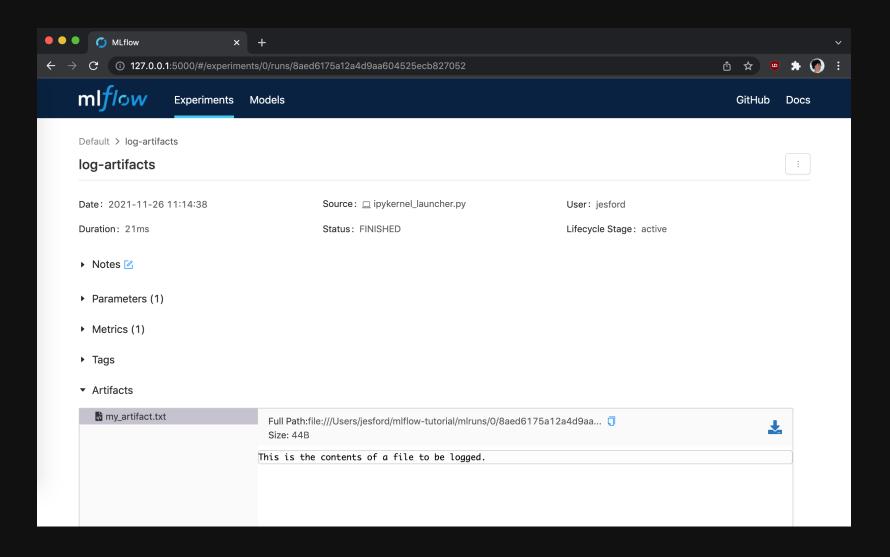
mlflow.log_param('my_parameter', 3)
    mlflow.log_metric('score', 95)

with open('my_artifact.txt', 'w') as f:
    f.write('This is the contents of a file to be logged.')

mlflow.log_artifact('my_artifact.txt')

mlflow.end_run()
```





### Where MLflow data is stored

By default your runs are recorded in files in a local mlruns/ folder that gets created in your current working directory.

Lots of other options for local and remote tracking (the latter is best for teams / sharing results) - see MLflow docs for possibilities.

## Not just for ML!

Notice that nothing we've done so far has been ML specific!

You could use MLFlow Tracking for kind of any analyses or projects where you find yourself manually recording values.

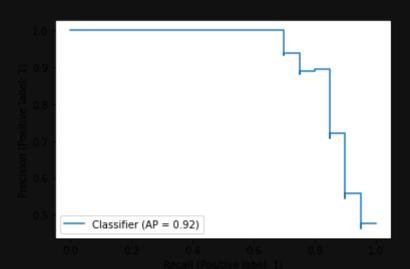
## Tracking your ML Model

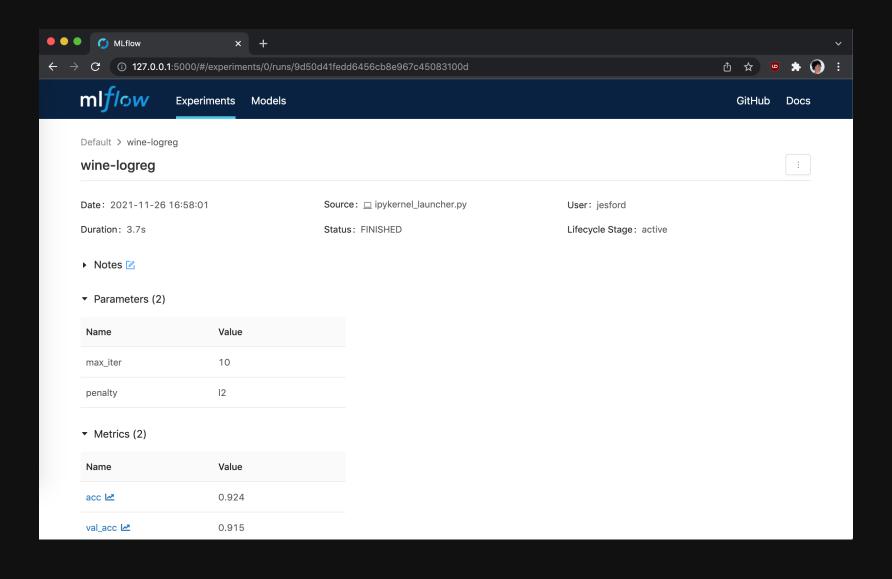
```
In [3]:
           import mlflow
           import matplotlib.pyplot as plt
           from sklearn.linear model import LogisticRegression
           from sklearn.metrics import accuracy score, PrecisionRecallDisplay
           from sklearn.model selection import train test split
           from sklearn.datasets import load wine
           # load and prepare data
           data = load wine()
           binary target = (data.target == 0).astype(int)
           X train, X test, y train, y test = train test split(
               data.data, binary target, test size=0.33, shuffle=True, random state=0,
           with mlflow.start run(run name='wine-logreg'): # use context manager instead of start/en
               penalty = '12'
               max iter = 10
               clf = LogisticRegression(penalty=penalty, max iter=max iter)
               clf.fit(X train, y train)
               # get predictions
               y train pred = clf.predict(X train)
               y test pred = clf.predict(X test)
               y test predprob = clf.predict proba(X test)[:, 1]
               fig, ax = plt.subplots(1, 1)
               PrecisionRecallDisplay.from predictions(y test, y test predprob, ax=ax)
               fig.savefig('PR.png')
```

```
mlflow.log_params({'penalty': penalty, 'max_iter': max_iter})
mlflow.log_metric('acc', accuracy_score(y_train, y_train_pred))
mlflow.log_metric('val_acc', accuracy_score(y_test, y_test_pred))
mlflow.log_artifact('PR.png')
mlflow.log_artifact('tutorial.ipynb')
mlflow.sklearn.log_model(clf, 'model')
```

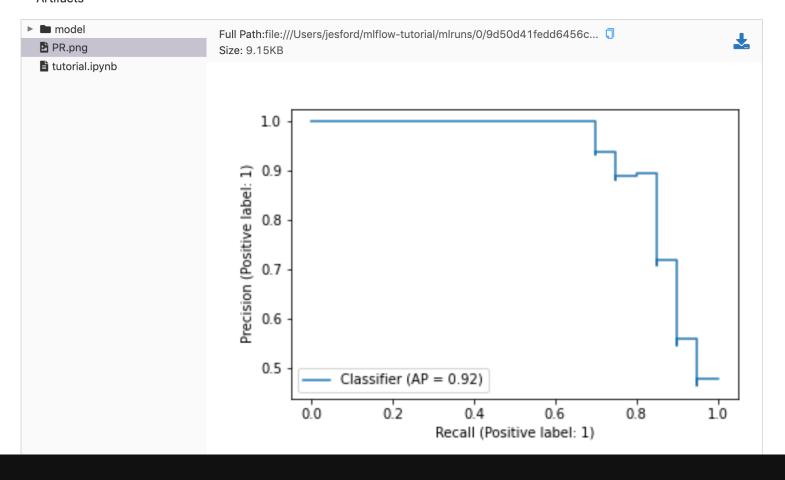
```
/Users/jesford/anaconda3/envs/mlflow-demo/lib/python3.7/site-packages/skle
arn/linear_model/_logistic.py:818: ConvergenceWarning: lbfgs failed to con
verge (status=1):
STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.

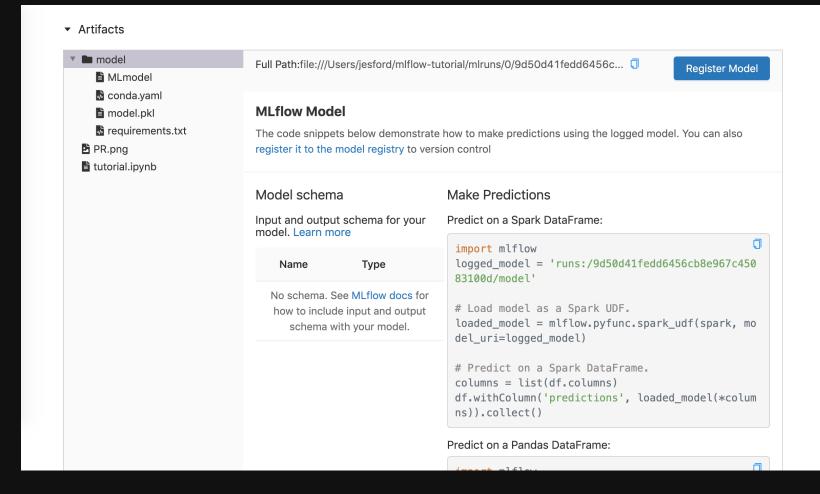
Increase the number of iterations (max_iter) or scale the data as shown i
n:
    https://scikit-learn.org/stable/modules/preprocessing.html
Please also refer to the documentation for alternative solver options:
    https://scikit-learn.org/stable/modules/linear_model.html#logistic-reg
ression
    extra_warning_msg=_LOGISTIC_SOLVER_CONVERGENCE_MSG,
```





#### ▼ Artifacts





MLflow provides code snippets for loading & using the trained model via the mlflow API

GitHub

Docs

Default > wine-logreg

#### wine-logreg

Date: 2021-11-26 16:58:01

Status: FINISHED

Source: \_\_ ipykernel\_launcher.py

User: jesford

Lifecycle Stage: active

▼ Notes <a></a>

Duration: 3.7s

#### **Wine Predictions**

Details about my LogisticRegression model to predict wines...

#### ▼ Parameters (2)

Name	Value
max_iter	10
penalty	12

▼ Metrics (2)

# **Autologging Model Training**

## **Autologging Model Training**

A single line of code will automatically tracks lots of useful things about your model, metrics, plots.

Supported for scikit-learn, tensorflow & keras, pytorch, xgboost, and more.

*Note:* this is an "experimental" feature in MLflow!

### **Load & Prepare a Real World Dataset**

"20 Newsgroups" dataset: text documents containing discussions of 20 different topics.

We'll just use 2 of these topics and build a binary classifier to distinguish between text that is about baseball vs hockey.



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"20 Newsgroups" dataset: text documents containing discussions of 20 different topics.

We'll just use 2 of these topics and build a binary classifier to distinguish between text that is about baseball vs hockey.

```
In [4]:
```

```
from sklearn.datasets import fetch_20newsgroups
from sklearn.feature_extraction.text import TfidfVectorizer

# load train and test data for 2 categories
categories_to_classify = ['rec.sport.baseball', 'rec.sport.hockey']
X_train_raw, y_train = fetch_20newsgroups(
    subset='train', categories=categories_to_classify, return_X_y=True
)
X_test_raw, y_test = fetch_20newsgroups(
    subset='test', categories=categories_to_classify, return_X_y=True
)
# turn the text into numbers
vectorizer = TfidfVectorizer(max_features=5000)
X_train = vectorizer.fit_transform(X_train_raw).todense()
X_test = vectorizer.transform(X_test_raw).todense()
print('Training set size after TF-IDF transform: {}'.format(X_train.shape))
print('\nExample document:\n\n{}'.format(X_train_raw[0]))
```

Training set size after TF-IDF transform: (1197, 5000)

Example document:

From: dougb@comm.mot.com (Doug Bank)

Subject: Re: Info needed for Cleveland tickets

Reply-To: dougb@ecs.comm.mot.com

Organization: Motorola Land Mobile Products Sector

Distribution: usa

Nntp-Posting-Host: 145.1.146.35

Lines: 17

In article <1993Apr1.234031.4950@leland.Stanford.EDU>, bohnert@leland.Stanford.EDU (matthew bohnert) writes:

- > I'm going to be in Cleveland Thursday, April 15 to Sunday, April 18.
- > Does anybody know if the Tribe will be in town on those dates, and
- > if so, who're they playing and if tickets are available?

The tribe will be in town from April 16 to the 19th.

There are ALWAYS tickets available! (Though they are playing Toronto, and many Toronto fans make the trip to Cleveland as it is easier to get tickets in Cleveland than in Toronto. Either way, I seriously doubt they will sell out until the end of the season.)

Doug Bank
dougb@ecs.comm.mot.com
dougb@nwu.edu
dougb@casbah.acns.nwu.edu

Private Systems Division
Motorola Communications Sector
Schaumburg, Illinois
708-576-8207

### Autologging Example 1: sklearn

In [5]:

```
from sklearn.ensemble import RandomForestClassifier

mlflow.sklearn.autolog() # just one line added!

rfc = RandomForestClassifier(n_estimators=10)

rfc.fit(X_train, y_train)
```

2021/11/27 15:57:24 WARNING mlflow.utils.autologging\_utils: You are using an unsupported version of sklearn. If you encounter errors during autologging, try upgrading / downgrading sklearn to a supported version, or try upgrading MLflow.

2021/11/27 15:57:24 INFO mlflow.utils.autologging\_utils: Created MLflow au tologging run with ID '0767d058f4c34495901426d10fd4d8b0', which will track hyperparameters, performance metrics, model artifacts, and lineage information for the current sklearn workflow

/Users/jesford/anaconda3/envs/mlflow-demo/lib/python3.7/site-packages/skle arn/utils/validation.py:590: FutureWarning: np.matrix usage is deprecated in 1.0 and will raise a TypeError in 1.2. Please convert to a numpy array with np.asarray. For more information see: https://numpy.org/doc/stable/reference/generated/numpy.matrix.html

FutureWarning,

/Users/jesford/anaconda3/envs/mlflow-demo/lib/python3.7/site-packages/skle arn/utils/validation.py:590: FutureWarning: np.matrix usage is deprecated in 1.0 and will raise a TypeError in 1.2. Please convert to a numpy array with np.asarray. For more information see: https://numpy.org/doc/stable/reference/generated/numpy.matrix.html

FutureWarning,

/Users/jesford/anaconda3/envs/mlflow-demo/lib/python3.7/site-packages/skle

arn/utils/validation.py:590: FutureWarning: np.matrix usage is deprecated in 1.0 and will raise a TypeError in 1.2. Please convert to a numpy array with np.asarray. For more information see: https://numpy.org/doc/stable/reference/generated/numpy.matrix.html

FutureWarning,

2021/11/27 15:57:25 WARNING mlflow.utils.autologging\_utils: MLflow autolog ging encountered a warning: "/Users/jesford/anaconda3/envs/mlflow-demo/lib/python3.7/site-packages/sklearn/utils/deprecation.py:87: FutureWarning: Function plot\_confusion\_matrix is deprecated; Function `plot\_confusion\_matrix` is deprecated in 1.0 and will be removed in 1.2. Use one of the class methods: ConfusionMatrixDisplay.from\_predictions or ConfusionMatrixDisplay.from estimator."

/Users/jesford/anaconda3/envs/mlflow-demo/lib/python3.7/site-packages/skle arn/utils/validation.py:590: FutureWarning: np.matrix usage is deprecated in 1.0 and will raise a TypeError in 1.2. Please convert to a numpy array with np.asarray. For more information see: https://numpy.org/doc/stable/reference/generated/numpy.matrix.html

FutureWarning,

2021/11/27 15:57:25 WARNING mlflow.utils.autologging\_utils: MLflow autolog ging encountered a warning: "/Users/jesford/anaconda3/envs/mlflow-demo/lib/python3.7/site-packages/sklearn/utils/deprecation.py:87: FutureWarning: Function plot\_roc\_curve is deprecated; Function `plot\_roc\_curve` is deprecated in 1.0 and will be removed in 1.2. Use one of the class methods: RocCurveDisplay.from\_predictions or RocCurveDisplay.from\_estimator."
/Users/jesford/anaconda3/envs/mlflow-demo/lib/python3.7/site-packages/sklearn/utils/validation.py:590: FutureWarning: np.matrix usage is deprecated in 1.0 and will raise a TypeError in 1.2. Please convert to a numpy array with np.asarray. For more information see: https://numpy.org/doc/stable/reference/generated/numpy.matrix.html

FutureWarning,

2021/11/27 15:57:25 WARNING mlflow.utils.autologging\_utils: MLflow autolog ging encountered a warning: "/Users/jesford/anaconda3/envs/mlflow-demo/lib/python3.7/site-packages/sklearn/utils/deprecation.py:87: FutureWarning:

Function plot\_precision\_recall\_curve is deprecated; Function `plot\_precisi on\_recall\_curve` is deprecated in 1.0 and will be removed in 1.2. Use one of the class methods: PrecisionRecallDisplay.from\_predictions or Precision RecallDisplay.from estimator."

/Users/jesford/anaconda3/envs/mlflow-demo/lib/python3.7/site-packages/skle arn/utils/validation.py:590: FutureWarning: np.matrix usage is deprecated in 1.0 and will raise a TypeError in 1.2. Please convert to a numpy array with np.asarray. For more information see: https://numpy.org/doc/stable/reference/generated/numpy.matrix.html

FutureWarning,

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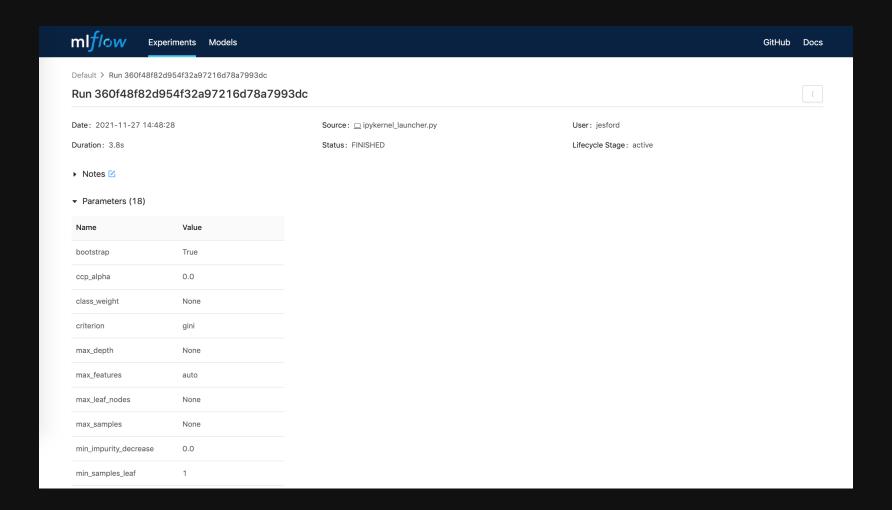
FutureWarning,

/Users/jesford/anaconda3/envs/mlflow-demo/lib/python3.7/site-packages/skle arn/utils/validation.py:590: FutureWarning: np.matrix usage is deprecated in 1.0 and will raise a TypeError in 1.2. Please convert to a numpy array with np.asarray. For more information see: https://numpy.org/doc/stable/reference/generated/numpy.matrix.html

FutureWarning,

#### Out[5]: RandomForestClassifier(n\_estimators=10)

### **Back in the MLflow UI...**



min_samples_split	2
min_weight_fraction_leaf	0.0
n_estimators	10
n_jobs	None
oob_score	False
random_state	None
verbose	0
warm_start  • Metrics (7)	False
warm_start  Metrics (7)  Name	False Value
Metrics (7)	Value
▼ Metrics (7)	Value
▼ Metrics (7)  Name  training_accuracy_score	<b>Value</b> 0.997
▼ Metrics (7)  Name  training_accuracy_score   training_f1_score	<b>Value</b> 0.997 0.997
Metrics (7)  Name  training_accuracy_score  training_f1_score  training_log_loss  training_log_loss	Value 0.997 0.997 0.081
Metrics (7)  Name  training_accuracy_score Lateral ining_f1_score Lateral ining_log_loss Lateral ining_precision_score Lateral	Value 0.997 0.997 0.081 0.997
Metrics (7)  Name  training_accuracy_score Lateral ining_f1_score Lateral ining_log_loss Lateral ining_precision_score Lateral ining_recall_score Lateral in	Value 0.997 0.997 0.081 0.997 0.997

# We see all RF hyperparameters (even those we didn't set explicitly) are recorded...

... as well as some default training metrics (would have to log validation metrics manually since fit doesn't know about them)

... and we get some really useful plots out of the box!

▼ Artifacts ▼ **■** model Full Path:file:///Users/jesford/mlflow-tutorial/mlruns/0/360f48f82d954f32a97216d78a7993dc/artifacts/training\_confusion\_matrix.png 🗍 MLmodel Size: 8.6KB 👪 conda.yaml model.pkl Normalized confusion matrix training\_confusion\_matrix.png training\_precision\_recall\_curve.png training\_roc\_curve.png - 0.8 0 0 -- 0.6 True label 0.4 0.0067 0.99 1 0.2

# Autologging Example 2: tensorflow keras

In [6]: import tensorflow as tf from tensorflow.keras import Input, Model, layers with mlflow.start run(run name='hockey-vs-baseball-NN'): mlflow.tensorflow.autolog() # just one line added! # parameters to vary hidden layer size = 16 learning rate = 0.1 # a simple NN with one hidden layer inputs = Input(shape=(X train.shape[1],)) x = layers.Dense(hidden layer size, activation='relu')(inputs) outputs = layers.Dense(1, activation='sigmoid')(x) model = Model(inputs, outputs) loss = 'binary crossentropy' optimizer = tf.optimizers.SGD(learning rate=learning rate) metrics = [ tf.metrics.BinaryAccuracy(), tf.metrics.AUC(curve='ROC', name='AUROC'), model.compile(optimizer=optimizer, loss=loss, metrics=metrics) model.fit(X train, y train, epochs=40, shuffle=True, validation data=(X test, y test)

2021-11-27 15:57:29.940299: I tensorflow/core/platform/cpu\_feature\_guard.cc:145] This TensorFlow binary is optimized with Intel(R) MKL-DNN to use the

```
SE4.2 AVX AVX2 FMA
To enable them in non-MKL-DNN operations, rebuild TensorFlow with the appr
opriate compiler flags.
2021-11-27 15:57:29.940539: I tensorflow/core/common runtime/process util.
cc:115] Creating new thread pool with default inter op setting: 4. Tune us
ing inter op parallelism threads for best performance.
Train on 1197 samples, validate on 796 samples
Epoch 1/40
ry accuracy: 0.5000 - AUROC: 0.4967
2021-11-27 15:57:32.775138: I tensorflow/core/profiler/lib/profiler sessio
n.cc:184] Profiler session started.
- binary accuracy: 0.5138 - AUROC: 0.5178 - val loss: 0.6923 - val binary
accuracy: 0.5038 - val AUROC: 0.5648
Epoch 2/40
5 - binary accuracy: 0.5556 - AUROC: 0.6514 - val loss: 0.6881 - val binar
y accuracy: 0.6583 - val AUROC: 0.7370
Epoch 3/40
3 - binary accuracy: 0.6942 - AUROC: 0.8035 - val loss: 0.6834 - val binar
y accuracy: 0.6118 - val AUROC: 0.8632
Epoch 4/40
2 - binary accuracy: 0.7853 - AUROC: 0.8988 - val loss: 0.6772 - val binar
y accuracy: 0.6093 - val AUROC: 0.9249
Epoch 5/40
7 - binary accuracy: 0.8755 - AUROC: 0.9448 - val loss: 0.6685 - val binar
```

e following CPU instructions in performance critical operations: SSE4.1 S

```
y accuracy: 0.8894 - val AUROC: 0.9536
Epoch 6/40
4 - binary accuracy: 0.8881 - AUROC: 0.9638 - val loss: 0.6623 - val binar
y accuracy: 0.5239 - val AUROC: 0.9693
Epoch 7/40
1 - binary accuracy: 0.8688 - AUROC: 0.9742 - val loss: 0.6469 - val binar
y accuracy: 0.9146 - val AUROC: 0.9752
Epoch 8/40
5 - binary accuracy: 0.9206 - AUROC: 0.9831 - val loss: 0.6327 - val binar
y accuracy: 0.9058 - val AUROC: 0.9774
Epoch 9/40
7 - binary accuracy: 0.9532 - AUROC: 0.9926 - val loss: 0.6161 - val binar
y accuracy: 0.8555 - val AUROC: 0.9809
Epoch 10/40
7 - binary accuracy: 0.9674 - AUROC: 0.9948 - val loss: 0.5978 - val binar
y accuracy: 0.8028 - val AUROC: 0.9823
Epoch 11/40
2 - binary accuracy: 0.9524 - AUROC: 0.9939 - val loss: 0.5704 - val binar
y accuracy: 0.9384 - val AUROC: 0.9830
Epoch 12/40
6 - binary accuracy: 0.9657 - AUROC: 0.9961 - val loss: 0.5457 - val binar
y accuracy: 0.9196 - val AUROC: 0.9838
Epoch 13/40
1 - binary accuracy: 0.9716 - AUROC: 0.9959 - val loss: 0.5161 - val binar
y accuracy: 0.9121 - val AUROC: 0.9851
```

```
Epoch 14/40
3 - binary accuracy: 0.9758 - AUROC: 0.9974 - val loss: 0.4851 - val binar
y accuracy: 0.9460 - val AUROC: 0.9856
Epoch 15/40
7 - binary accuracy: 0.9733 - AUROC: 0.9979 - val loss: 0.4630 - val binar
y accuracy: 0.9146 - val AUROC: 0.9860
Epoch 16/40
8 - binary accuracy: 0.9816 - AUROC: 0.9980 - val loss: 0.4260 - val binar
y accuracy: 0.9472 - val AUROC: 0.9869
Epoch 17/40
0 - binary accuracy: 0.9766 - AUROC: 0.9985 - val loss: 0.4081 - val binar
y accuracy: 0.9008 - val AUROC: 0.9875
Epoch 18/40
3 - binary accuracy: 0.9749 - AUROC: 0.9985 - val loss: 0.3729 - val binar
y accuracy: 0.9485 - val AUROC: 0.9880
Epoch 19/40
2 - binary accuracy: 0.9841 - AUROC: 0.9990 - val loss: 0.3501 - val binar
y accuracy: 0.9497 - val AUROC: 0.9886
Epoch 20/40
7 - binary accuracy: 0.9891 - AUROC: 0.9993 - val loss: 0.3304 - val binar
y accuracy: 0.9485 - val AUROC: 0.9891
Epoch 21/40
5 - binary accuracy: 0.9891 - AUROC: 0.9991 - val loss: 0.3101 - val binar
y accuracy: 0.9447 - val AUROC: 0.9896
Epoch 22/40
```

```
8 - binary accuracy: 0.9866 - AUROC: 0.9994 - val loss: 0.3046 - val binar
y accuracy: 0.9234 - val AUROC: 0.9900
Epoch 23/40
8 - binary accuracy: 0.9908 - AUROC: 0.9995 - val loss: 0.2832 - val binar
y accuracy: 0.9560 - val AUROC: 0.9903
Epoch 24/40
9 - binary accuracy: 0.9925 - AUROC: 0.9997 - val loss: 0.2640 - val binar
y accuracy: 0.9523 - val AUROC: 0.9906
Epoch 25/40
6 - binary accuracy: 0.9925 - AUROC: 0.9998 - val loss: 0.2530 - val binar
y accuracy: 0.9447 - val AUROC: 0.9908
Epoch 26/40
5 - binary accuracy: 0.9925 - AUROC: 0.9997 - val loss: 0.2417 - val binar
y accuracy: 0.9485 - val AUROC: 0.9911
Epoch 27/40
3 - binary accuracy: 0.9958 - AUROC: 0.9998 - val loss: 0.2332 - val binar
y accuracy: 0.9598 - val AUROC: 0.9914
Epoch 28/40
8 - binary accuracy: 0.9958 - AUROC: 0.9999 - val loss: 0.2232 - val binar
y accuracy: 0.9485 - val AUROC: 0.9917
Epoch 29/40
2 - binary accuracy: 0.9958 - AUROC: 0.9999 - val loss: 0.2139 - val binar
y accuracy: 0.9548 - val AUROC: 0.9919
Epoch 30/40
```

```
7 - binary accuracy: 0.9950 - AUROC: 0.9999 - val loss: 0.2087 - val binar
y accuracy: 0.9472 - val AUROC: 0.9921
Epoch 31/40
8 - binary accuracy: 0.9958 - AUROC: 0.9999 - val loss: 0.2000 - val binar
y accuracy: 0.9560 - val AUROC: 0.9923
Epoch 32/40
3 - binary accuracy: 0.9975 - AUROC: 0.9999 - val loss: 0.1935 - val binar
y accuracy: 0.9560 - val AUROC: 0.9926
Epoch 33/40
7 - binary accuracy: 0.9958 - AUROC: 1.0000 - val loss: 0.1883 - val binar
y accuracy: 0.9560 - val AUROC: 0.9928
Epoch 34/40
6 - binary accuracy: 0.9983 - AUROC: 1.0000 - val loss: 0.1827 - val binar
y accuracy: 0.9585 - val AUROC: 0.9930
Epoch 35/40
8 - binary accuracy: 0.9992 - AUROC: 1.0000 - val loss: 0.1781 - val binar
y accuracy: 0.9548 - val AUROC: 0.9931
Epoch 36/40
9 - binary accuracy: 0.9992 - AUROC: 1.0000 - val loss: 0.1735 - val binar
y accuracy: 0.9585 - val AUROC: 0.9933
Epoch 37/40
9 - binary accuracy: 0.9992 - AUROC: 1.0000 - val loss: 0.1700 - val binar
y accuracy: 0.9648 - val AUROC: 0.9934
Epoch 38/40
9 - binary accuracy: 0.9992 - AUROC: 1.0000 - val loss: 0.1662 - val binar
```

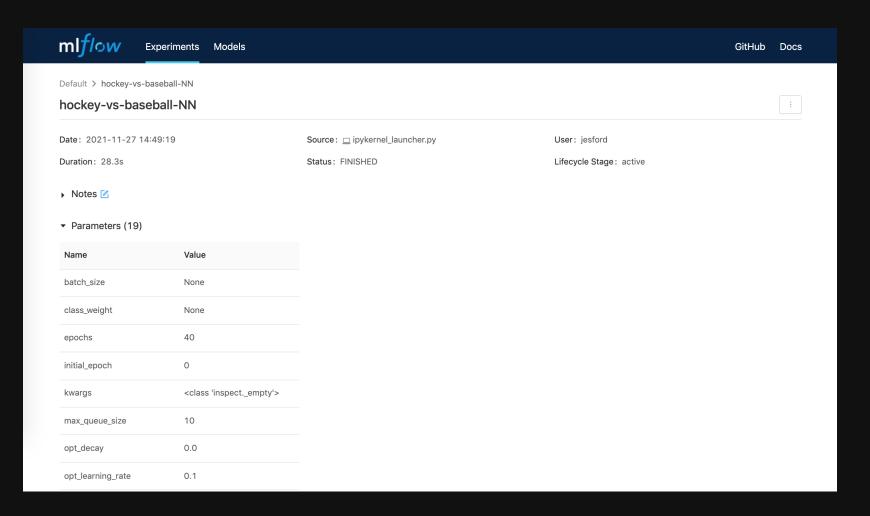
y accuracy: 0.9648 - val AUROC: 0.9935 Epoch 39/40 6 - binary accuracy: 0.9992 - AUROC: 1.0000 - val loss: 0.1638 - val binar y accuracy: 0.9636 - val AUROC: 0.9936 Epoch 40/40 1 - binary accuracy: 0.9992 - AUROC: 1.0000 - val loss: 0.1598 - val binar y accuracy: 0.9661 - val AUROC: 0.9938 WARNING: tensorflow: From /Users/jesford/anaconda3/envs/mlflow-demo/lib/pyth on3.7/site-packages/tensorflow core/python/ops/resource variable ops.py:17 81: calling BaseResourceVariable. init (from tensorflow.python.ops.reso urce variable ops) with constraint is deprecated and will be removed in a future version. Instructions for updating:

If using Keras pass \* constraint arguments to layers.

2021-11-27 15:57:54.053267: W tensorflow/python/util/util.cc:299] Sets are not currently considered sequences, but this may change in the future, so consider avoiding using them.

INFO:tensorflow:Assets written to: /var/folders/xv/48y52h7j27g9jp0vs5n4s8r c0000gn/T/tmph4l1rgnb/model/data/model/assets

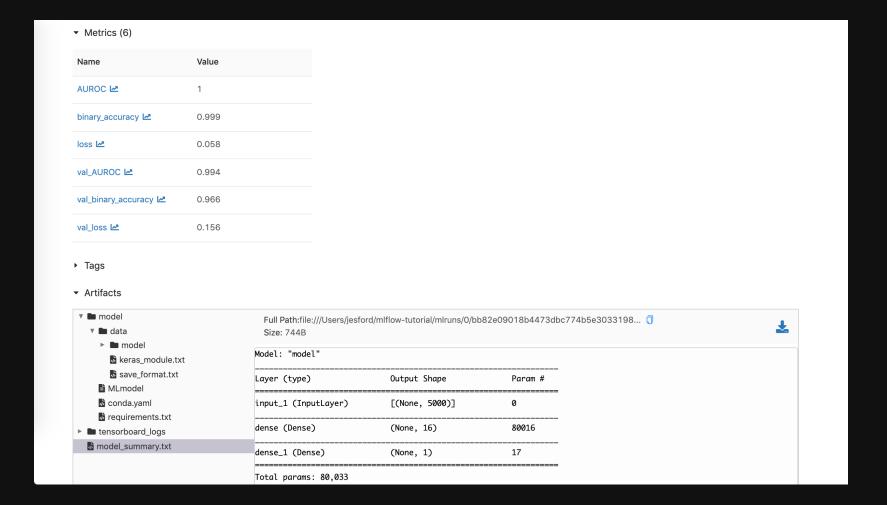
# Again, all the parameters & metrics relevant to the run are recorded automatically



opt_momentum	0.0
opt_name	SGD
opt_nesterov	False
sample_weight	None
shuffle	True
steps_per_epoch	None
use_multiprocessing	False
validation_freq	1
validation_split	0.0
validation_steps	None
workers	1

#### ▼ Metrics (6)

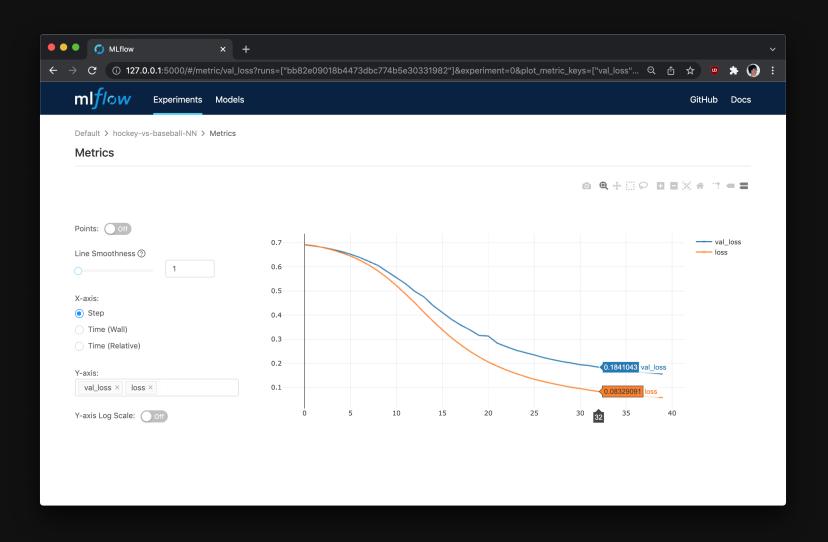
Name	Value
AUROC <u>▶</u>	1
binary_accuracy 🗠	0.999
loss 🗠	0.058



## Notice model summary, tensorboard logs, and the TF saved model

Note: our team uses this saved model artifact (the inner model folder) directly for deployment, since our infrastructure isn't yet set up to deploy from MLflow directly.

# Clicking into one of the metrics, we can view training curves





# Powerful features for comparing lots of models

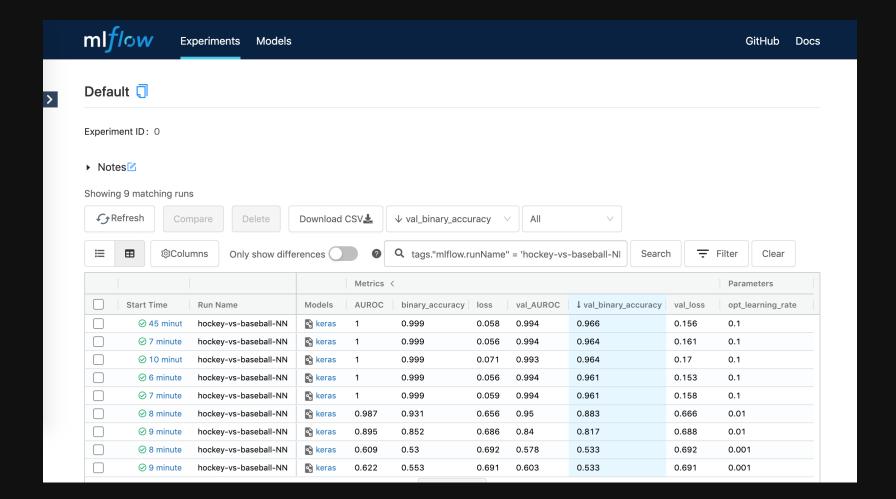
Suppose you had run many iterations of this small NN with different hyperparameters...



# Powerful features for comparing lots of models

Suppose you had run many iterations of this small NN with different hyperparameters...

In the UI you can sort by metrics

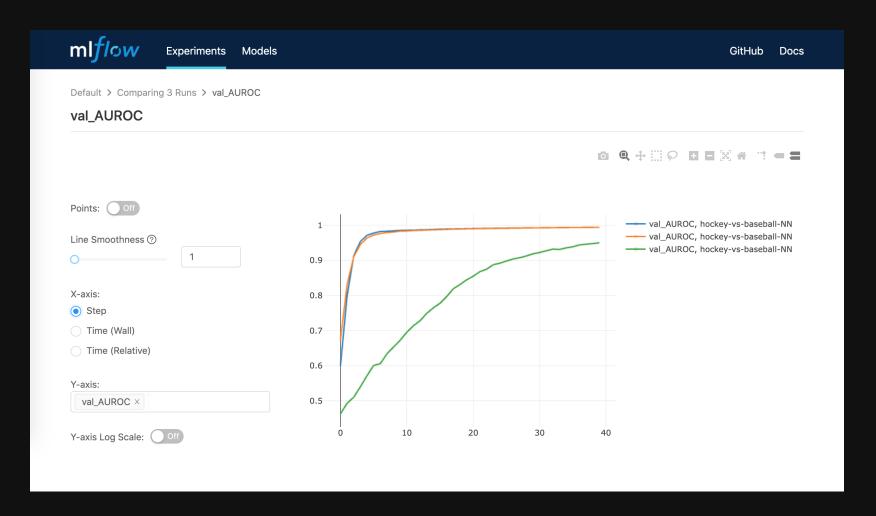


#### ... you also view run details side-by-side

By selecting a few checkboxes and clicking the "Compare" button. Notice parameter differences are highlighted!

#### ... and plot training curves from different runs

Similar to TensorBoard.



# Incorporating MLflow into our Model Review Process

## **Process Part 1: Training Infrastructure**

Create/maintain a reasonably-flexibly **common training infrastructure** that works for *most* of our team's problems.

### **Process Part 2: MLflow Tracking**

**MLflow Tracking** is embedded in this infrastructure, so experiment record-keeping happens automatically:

- all training parameters & settings
- train and test metrics
- environment: the docker image, code version, the training script or notebook itself
- model (and related artifacts needed for deployment)
- common plots and results of analyses that help us understand performance

#### **Process Part 3: Use Notes for Context**

**MLflow UI Markdown Notes** section used for all the context that can't be recorded automatically:

- business requirements, e.g. links to any project docs
- things we tried that didn't work
- describe any non-standard treatment of the data
- decisions or trade-offs made along the way
- if shadow deployed, any observations on live data
- anything you'd want to know if you were picking up the project from scratch

#### **Process Part 4: Model Review**

Prior to Live Launch of a new model, trigger a **Model Review**:

- make sure all "required" fields were tracked and fill out the Notes section
- request a primary and secondary reviewer and send them the link to the MLflow run
- reviewers expected to review the MLflow run & related code async
- schedule a 30min synchronous review, other team members invited as optional
- meeting is to discuss, ask questions, make suggestions (rarely to block deployment)

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Note: this part of our process is evolving! It's not perfect (particularly around collecting reviewer feedback) and I'd love to learn what others are doing.

#### **Back to our Goals for Model Review**

- 1. **Transparency** is acheived by having shareable links to all the training details.
- 2. **Reproducibility** is improved by recording everything needed to re-create a training run.
- 3. **Knowledge Sharing** happens through the review process or simply by sharing MLflow examples with colleagues.

MLflow is a helpful tool to achieve these goals because we can view so many details about the model in one place.

### **Summary**

- We improved model deployment record keeping by integrating MLflow tracking into our shared training infrastructure.
- MLflow is a lightweight and powerful way to track your ML experiments
  - adding one line of code (autologging) can get you a long way!
  - we only talked about Tracking here (see also: MLflow Projects, Models, Model Registry)
- Our review "process" is a WIP! I'd really love to talk to anyone who has different processes in place for things like this!

## **Questions?**

This notebook presentation is on GitHub: https://github.com/jesford/model-review