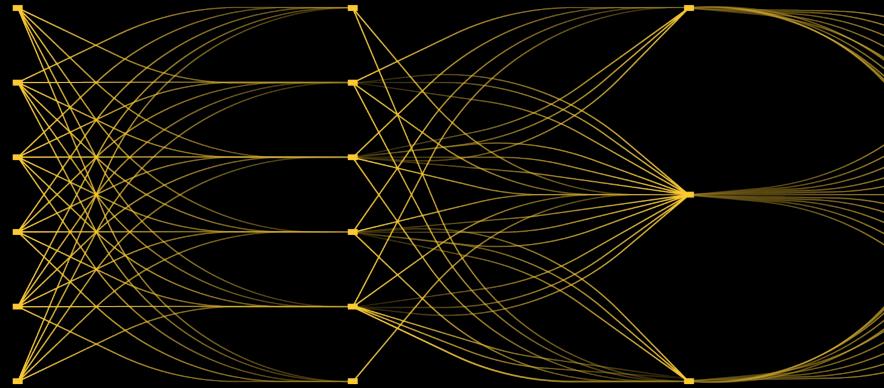




Iterative, Multiplayer Research



January 2023



Accounts Needed for Pt. 2, Competition

kaggle.com

wandb.ai

colab.research.google.com



The W&B Course

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The Goal

Table 3: Detection results on **PASCAL VOC 2007 test set**. The detector is Fast R-CNN and VGG-16. Training data: “07”: VOC 2007 trainval, “07+12”: union set of VOC 2007 trainval and VOC 2012 trainval. For RPN, the train-time proposals for Fast R-CNN are 2000. \dagger : this number was reported in [2]; using the repository provided by this paper, this result is higher (68.1).

method	# proposals	data	mAP (%)
SS	2000	07	66.9 \dagger
SS	2000	07+12	70.0
RPN+VGG, unshared	300	07	68.5
RPN+VGG, shared	300	07	69.9
RPN+VGG, shared	300	07+12	73.2
RPN+VGG, shared	300	COCO+07+12	78.8

Table 4: Detection results on **PASCAL VOC 2012 test set**. The detector is Fast R-CNN and VGG-16. Training data: “07”: VOC 2007 trainval, “07++12”: union set of VOC 2007 trainval+test and VOC 2012 trainval. For RPN, the train-time proposals for Fast R-CNN are 2000. \dagger : <http://host.robots.ox.ac.uk:8080/anonymous/HZJTQA.html>. \ddagger : <http://host.robots.ox.ac.uk:8080/anonymous/YNPLXB.html>. \S : <http://host.robots.ox.ac.uk:8080/anonymous/XEDH10.html>.

method	# proposals	data	mAP (%)
SS	2000	12	65.7
SS	2000	07++12	68.4
RPN+VGG, shared \dagger	300	12	67.0
RPN+VGG, shared \ddagger	300	07++12	70.4
RPN+VGG, shared \S	300	COCO+07++12	75.9

Table 5: **Timing** (ms) on a K40 GPU, except SS proposal is evaluated in a CPU. “Region-wise” includes NMS, pooling, fully-connected, and softmax layers. See our released code for the profiling of running time.

model	system	conv	proposal	region-wise	total	rate
VGG	SS + Fast R-CNN	146	1510	174	1830	0.5 fps
VGG	RPN + Fast R-CNN	141	10	47	198	5 fps
ZF	RPN + Fast R-CNN	31	3	25	59	17 fps

	backbone	AP	AP ₅₀	AP ₇₅	AP _S	AP _M	AP _L
<i>Two-stage methods</i>							
Faster R-CNN+++ [5]	ResNet-101-C4	34.9	55.7	37.4	15.6	38.7	50.9
Faster R-CNN w FPN [8]	ResNet-101-FPN	36.2	59.1	39.0	18.2	39.0	48.2
Faster R-CNN by G-RMI [6]	Inception-ResNet-v2 [21]	34.7	55.5	36.7	13.5	38.1	52.0
Faster R-CNN w TDM [20]	Inception-ResNet-v2-TDM	36.8	57.7	39.2	16.2	39.8	52.1
<i>One-stage methods</i>							
YOLOv2 [15]	DarkNet-19 [15]	21.6	44.0	19.2	5.0	22.4	35.5
SSD513 [11, 3]	ResNet-101-SSD	31.2	50.4	33.3	10.2	34.5	49.8
DSSD513 [3]	ResNet-101-DSSD	33.2	53.3	35.2	13.0	35.4	51.1
RetinaNet [9]	ResNet-101-FPN	39.1	59.1	42.3	21.8	42.7	50.2
RetinaNet [9]	ResNeXt-101-FPN	40.8	61.1	44.1	24.1	44.2	51.2
YOLOv3 608 × 608	Darknet-53	33.0	57.9	34.4	18.3	35.4	41.9

```
0000 p1 DRZ 3.11 - not super eval
mag_threshold 0.21
2L 300N BLSTM (BasicLSTM)
20D
sigmoid
AdamOptimizer
100 frames
dropout 1.0
zero input and label
log(x+1.0)
103300 training, 2000 CV
model:
weights20170224-005946_v10.1419
(p1, loss .1419, epoch 40 [task0])
MEAN IBM SDR GAIN: 2.324 -
with 0.15 thresh during cluster
STD IBM SDR GAIN: 2.276
MEAN IBM SDR GAIN: 2.110 -
with 0.32 threshold during cluster
STD IBM SDR GAIN: 2.254

0001 p2 DRZ 3.23
mag_threshold 0.12
2L 300N BLSTM_clean (LSTM & many
reworks) - note, this was the
massive model rewrite
20D
sigmoid
AdamOptimizer
100 frames
dropout 1.0
zero input and label
log(x+1.0)
10330 training, 2000 cv
model:
weights20170224-032054_v10.1418
(p2, loss .1418, epoch 40 [task0])
MEAN IBM SDR GAIN: 2.056 -
with 0.15 thresh during cluster
STD IBM SDR GAIN: 2.214
MEAN IBM SDR GAIN: 2.068 -
with 0.32 threshold during cluster
STD IBM SDR GAIN: 2.205
MEAN IBM SDR GAIN: 2.108 - with
fancy best SDR of the two system
(eval_sdr2.py)
STD IBM SDR GAIN: 4.114
MEAN IBM SDR GAIN: 5.915 -
using EXACT script measuring both
voices gain (SUPER_EVAL)
STD IBM SDR GAIN: 4.349
```

```
0000 p1 DRZ 3.11 - not super eval
```

```
mag_threshold 0.21
2L 300N BLSTM (BasicLSTM)
20D
sigmoid
AdamOptimizer
100 frames
dropout 1.0
zero input and label
log(x+1.0)
103300 training, 2000 CV
model:
```

```
weights20170224-005946_v10
(p1, loss .1419, epoch 40
MEAN IBM SDR GAIN: 2.
with 0.15 thresh during cl
STD IBM SDR GAIN: 2.
MEAN IBM SDR GAIN: 2.
with 0.32 threshold during
STD IBM SDR GAIN: 2.
```

```
0001 p2 DRZ 3.23
```

```
mag threshold 0.12
2L 300N BLSTM_clean (LSTM & many
reworks) - note, this was the
massive model rewrite
20D
sigmoid
AdamOptimizer
100 frames
dropout 1.0
zero input and label
log(x+1.0)
```



A	B	◀ ▶	L	M	N	O	P
Experiment Name	Created		train_loss	valid_loss	acc	traffic_acc	road_acc
best car acc (50% data)	2021-04-14		0.5375041962	0.442730248	0.8823291659	0.8663836718	0.9399003386
best traffic acc (50% data))	2021-04-14		0.4919361174	0.4202951491	0.8879730701	0.8718349934	0.9439761043
best overall IOU (20% data)	2021-04-14		0.5095784068	0.4658596516	0.8725891709	0.8592621684	0.9359762073
major-sweep-196	2021-01-31		0.5705417991	0.4875227213	0.8698127866	0.8570468426	0.9454026222
swept-sweep-164	2021-01-31		0.5535062551	0.4849829972	0.8701210618	0.8567070365	0.9204238057
silver-sweep-139	2021-01-31		0.563354373	0.5251165628	0.871628046	0.846842885	0.9262287617
laced-sweep-115	2021-01-31		0.5277443528	0.5124291778	0.8705932498	0.8521561027	0.9389513731
eager-sweep-97	2021-01-31		0.5488699675	0.5005864501	0.8738754392	0.8612990975	0.913561523
rich-sweep-88	2021-01-31		0.5587444901	0.5211353302	0.8785927892	0.8512274623	0.9295567274
hopeful-sweep-33	2021-01-31		0.503461957	0.4650281966	0.8706912994	0.8560319543	0.9387732744
autumn-sweep-24	2021-01-31		0.5777919888	0.500880897	0.8755427003	0.8561192751	0.9181208611
decent-sweep-21	2021-01-31		0.5714729428	0.4979581237	0.8745227456	0.8490597606	0.9463140965
vague-sweep-5	2021-01-31		0.6230331063	0.473508656	0.8732874393	0.8601382971	0.9297611117
Second best acc	2021-01-31		0.4194990396	0.4509823024	0.8873019218	0.8705806732	0.9445936084

```
0000 p1 DRZ 3.11 - not super eval
```

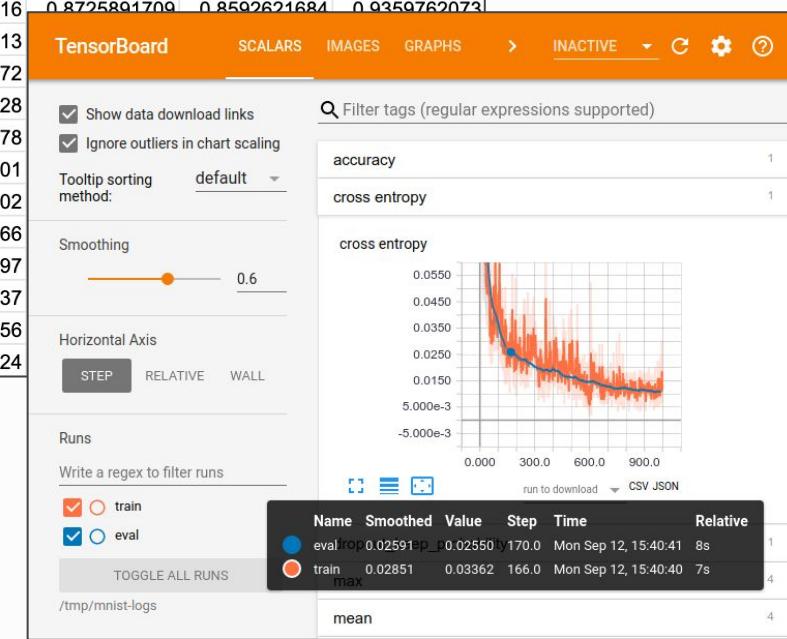
```
mag_threshold 0.21
2L 300N BLSTM (BasicLSTM)
20D
sigmoid
AdamOptimizer
100 frames
dropout 1.0
zero input and label
log(x+1.0)
103300 training, 2000 CV
model:
```

```
weights20170224-005946_v10
(p1, loss .1419, epoch 40
MEAN IBM SDR GAIN: 2.
with 0.15 thresh during cl
STD IBM SDR GAIN: 2.
MEAN IBM SDR GAIN: 2.
with 0.32 threshold during
STD IBM SDR GAIN: 2.
```

```
0001 p2 DRZ 3.23
```

```
mag threshold 0.12
2L 300N BLSTM_clean (LSTM & many
reworks) - note, this was the
massive model rewrite
20D
sigmoid
AdamOptimizer
100 frames
dropout 1.0
zero input and label
log(x+1.0)
```

A	B	L	M	N	O	P
Experiment Name	Created	train_loss	valid_loss	acc	traffic_acc	road_acc
best car acc (50% data)	2021-04-14	0.5375041962	0.442730248	0.8823291659	0.8663836718	0.9399003386
best traffic acc (50% data))	2021-04-14	0.4919361174	0.4202951491	0.8879730701	0.8718349934	0.9439761043
best overall IOU (20% data)	2021-04-14	0.5095784068	0.4658596516	0.8725891709	0.8502621684	0.939762073
major-sweep-196	2021-01-31	0.5705417991	0.4875227213			
swept-sweep-164	2021-01-31	0.5535062551	0.4849829972			
silver-sweep-139	2021-01-31	0.563354373	0.5251165628			
laced-sweep-115	2021-01-31	0.5277443528	0.5124291778			
eager-sweep-97	2021-01-31	0.5488699675	0.5005864501			
rich-sweep-88	2021-01-31	0.5587444901	0.5211353302			
hopeful-sweep-33	2021-01-31	0.503461957	0.4650281966			
autumn-sweep-24	2021-01-31	0.5777919888	0.500880897			
decent-sweep-21	2021-01-31	0.5714729428	0.4979581237			
vague-sweep-5	2021-01-31	0.6230331063	0.473508656			
Second best acc	2021-01-31	0.4194990396	0.4509823024			

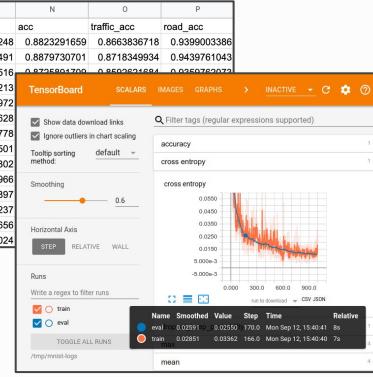


```

0000 p1 DRZ 3.11 - not super eval
model = Sequential([
    ...
    Dense(10, activation='relu'),
    Dense(10, activation='relu'),
    Dense(10, activation='softmax')
])
optimizer = Adam()
loss_fn = SparseCategoricalCrossentropy()
metrics = [accuracy]
log_dir = 'logs/fit'
tensorboard_callback = TensorBoard(log_dir=log_dir, histogram_freq=1)
early_stopping = EarlyStopping(monitor='val_accuracy', patience=3, restore_best_weights=True)
reduce_lr = ReduceLROnPlateau(monitor='val_accuracy', factor=0.2, patience=2, min_lr=1e-05)
# Train the model
model.compile(optimizer=optimizer, loss=loss_fn, metrics=metrics)
history = model.fit(
    train_data,
    train_labels,
    validation_data=(val_data, val_labels),
    epochs=40,
    batch_size=32,
    callbacks=[early_stopping, reduce_lr],
    verbose=1
)
# Evaluate the model
test_loss, test_accuracy = model.evaluate(test_data, test_labels)
print(f'Test accuracy: {test_accuracy * 100:.2f}%')

```

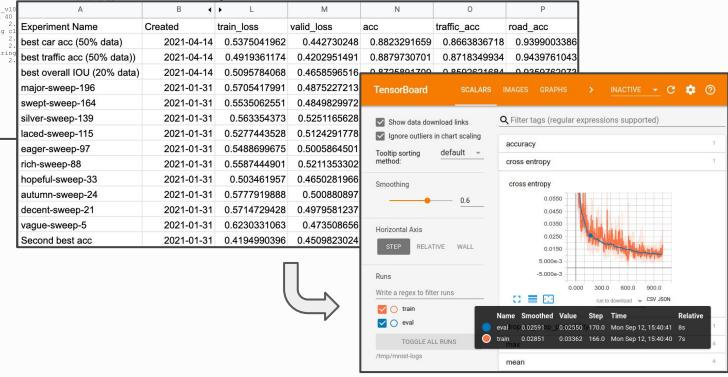
Experiment Name	Created	train_loss	valid_loss	acc	traffic_acc	road_acc
best car acc (50% data)	2021-04-14	0.5375041962	0.442730248	0.8823291659	0.866336718	0.9399003386
best traffic acc (50% data)	2021-04-14	0.4919361174	0.4202951491	0.8879730701	0.8718349934	0.9439761043
best overall IOU (20% data)	2021-04-14	0.5095784068	0.4658596516	0.8725691769	0.8692631684	0.9350762072
major-sweep-196	2021-01-31	0.5705417991	0.4875227213			
swept-sweep-164	2021-01-31	0.5535062551	0.4849829972			
silver-sweep-139	2021-01-31	0.563354373	0.5251165628			
laced-sweep-115	2021-01-31	0.5277443528	0.5124291778			
eager-sweep-97	2021-01-31	0.5488690675	0.5005864501			
rich-sweep-88	2021-01-31	0.5867444901	0.5211353302			
hopeful-sweep-33	2021-01-31	0.503461957	0.4650281966			
autumn-sweep-24	2021-01-31	0.5777919888	0.500880897			
decent-sweep-21	2021-01-31	0.5714729428	0.4979581237			
vague-sweep-5	2021-01-31	0.6230331063	0.473508656			
Second best acc	2021-01-31	0.4194990396	0.4509823024			



```

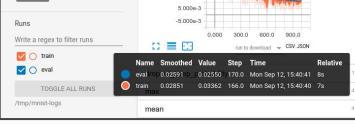
0000 p1 DRZ 3.11 - not super eval
model = Sequential([
    ...
    Dense(10, activation='relu'),
    Dense(10, activation='relu'),
    Dense(10, activation='softmax')
])
optimizer = Adam()
loss_fn = SparseCategoricalCrossentropy()
metrics = [accuracy]
log_dir = 'logs/fit'
tensorboard_callback = TensorBoard(log_dir=log_dir, histogram_freq=1)
early_stopping = EarlyStopping(monitor='val_accuracy', patience=3, restore_best_weights=True)
reduce_lr = ReduceLROnPlateau(monitor='val_accuracy', factor=0.2, patience=2, min_lr=1e-05)
# Train the model
model.compile(optimizer=optimizer, loss=loss_fn, metrics=metrics)
history = model.fit(
    train_data,
    train_labels,
    validation_data=(val_data, val_labels),
    epochs=40,
    batch_size=32,
    callbacks=[early_stopping, reduce_lr],
    verbose=1
)
# Evaluate the model
test_loss, test_accuracy = model.evaluate(test_data, test_labels)
print(f'Test accuracy: {test_accuracy * 100:.2f}%')

```



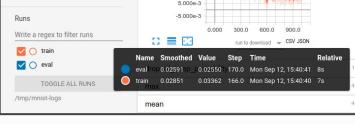
0000 p1 DRX 3.11 - not super eval
 weights0107224-005946.vtk
 best threshold 0.121
 2L_BORN_BLMST(basicLSTM)
 100 frames
 sigmoid
 AdamOptimizer
 100 frames
 zero input and label
 log(max,1.0)
 103300 training, 2000 CV
 model:
 weights0107224-005946.vtk
 (p1, loss= 1415, epoch: 40
 max iterations: 1000
 with 0.15% threads during CV
 2L_BORN_BLMST(basicLSTM)
 MEAN IBM SIMD GAIN: 2
 best traffic acc (50% data)
 2021-04-14 0.50957840688 0.4658596516
 with 0.15% threads during CV
 IBM SIMD GAIN: 2

best overall IOU (20% data)
 major-sweep-16
 swept-sweep-164
 silver-sweep-139
 laced-sweep-115
 eager-sweep-97
 rich-sweep-88
 hopeful-sweep-33
 autumn-sweep-24
 decent-sweep-21
 vague-sweep-5
 Second best acc



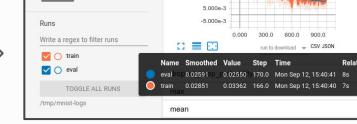
0000 p1 DRX 3.11 - not super eval
 best threshold 0.121
 2L_BORN_BLMST(basicLSTM)
 100 frames
 sigmoid
 AdamOptimizer
 100 frames
 zero input and label
 log(max,1.0)
 103300 training, 2000 CV
 model:
 weights0107224-005946.vtk
 (p1, loss= 1415, epoch: 40
 max iterations: 1000
 with 0.15% threads during CV
 2L_BORN_BLMST(basicLSTM)
 MEAN IBM SIMD GAIN: 2
 best traffic acc (50% data)
 2021-04-14 0.4919381174 0.4202951491
 with 0.15% threads during CV
 IBM SIMD GAIN: 2

best overall IOU (20% data)
 major-sweep-16
 swept-sweep-164
 silver-sweep-139
 laced-sweep-115
 eager-sweep-97
 rich-sweep-88
 hopeful-sweep-33
 autumn-sweep-24
 decent-sweep-21
 vague-sweep-5
 Second best acc



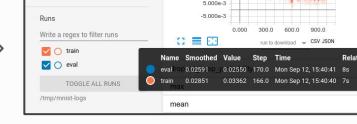
0000 p1 DRX 3.11 - not super eval
 weights0107224-005946.vtk
 best threshold 0.121
 2L_BORN_BLMST(basicLSTM)
 100 frames
 sigmoid
 AdamOptimizer
 100 frames
 zero input and label
 log(max,1.0)
 103300 training, 2000 CV
 model:
 weights0107224-005946.vtk
 (p1, loss= 1415, epoch: 40
 max iterations: 1000
 with 0.15% threads during CV
 2L_BORN_BLMST(basicLSTM)
 MEAN IBM SIMD GAIN: 2
 best traffic acc (50% data)
 2021-04-14 0.8879730701 0.8718349934
 with 0.15% threads during CV
 IBM SIMD GAIN: 2

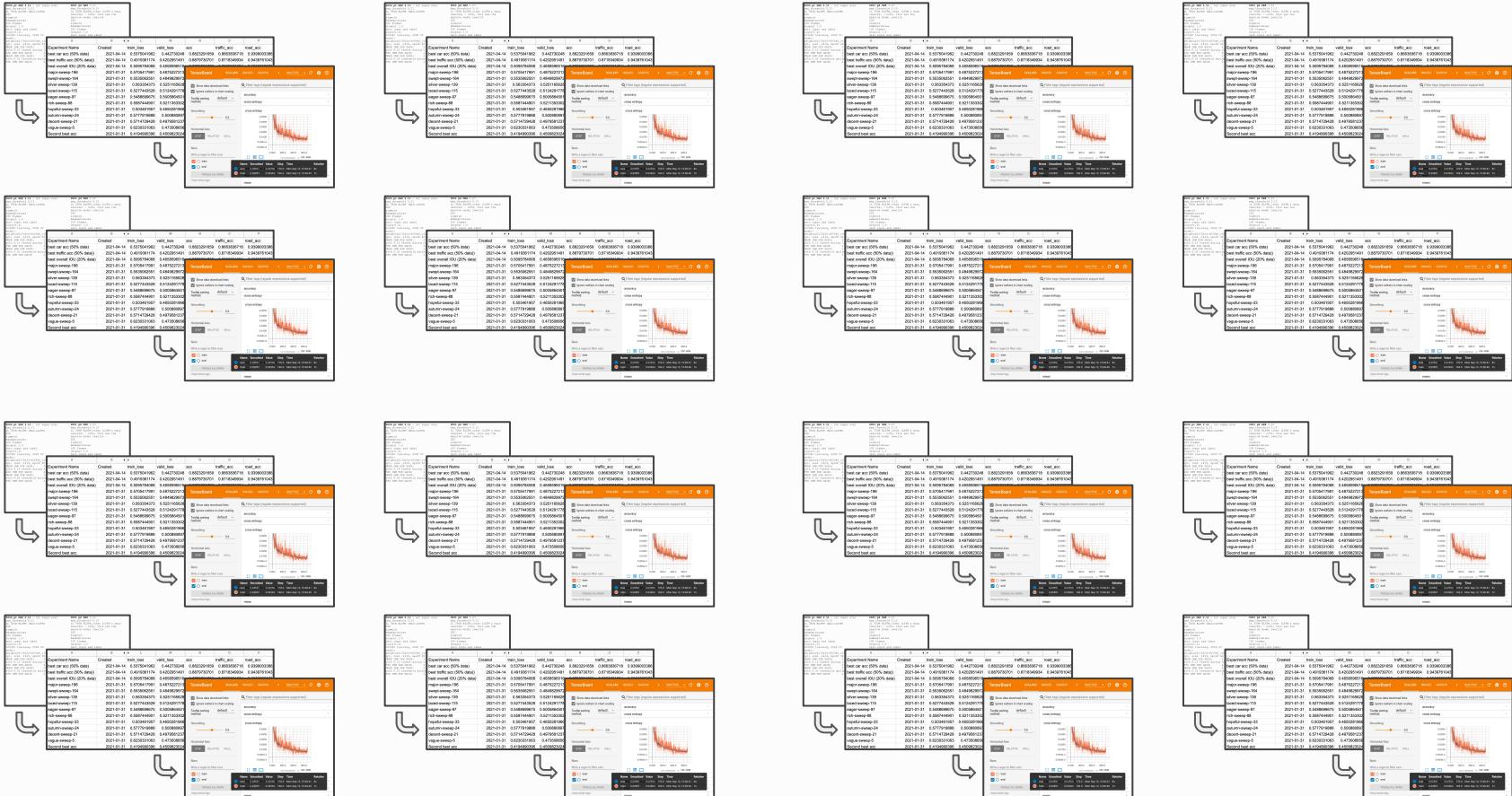
best overall IOU (20% data)
 major-sweep-16
 swept-sweep-164
 silver-sweep-139
 laced-sweep-115
 eager-sweep-97
 rich-sweep-88
 hopeful-sweep-33
 autumn-sweep-24
 decent-sweep-21
 vague-sweep-5
 Second best acc

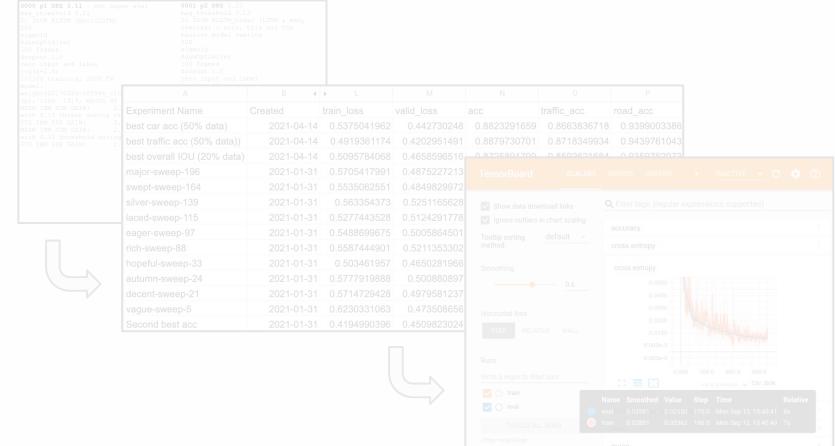
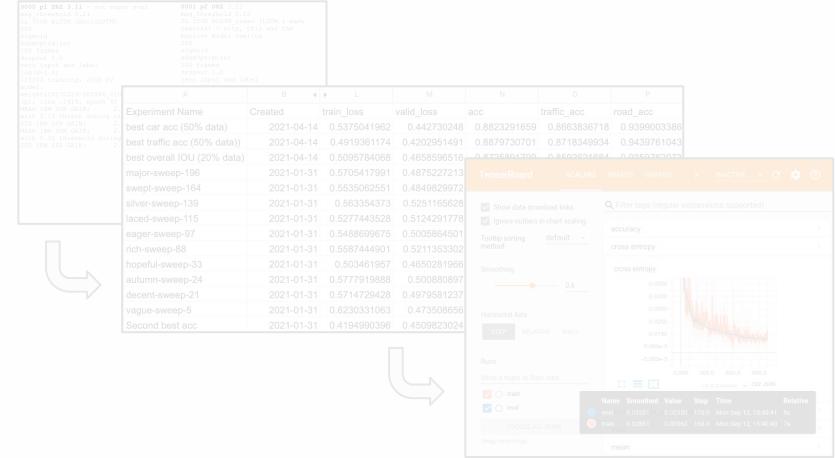
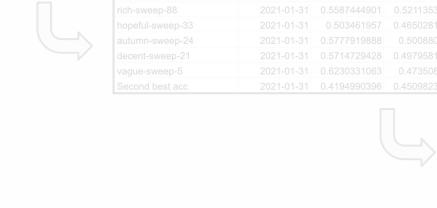
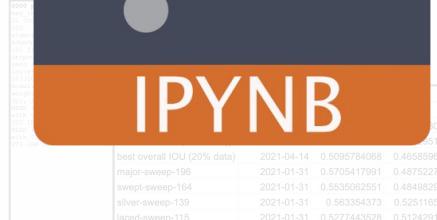
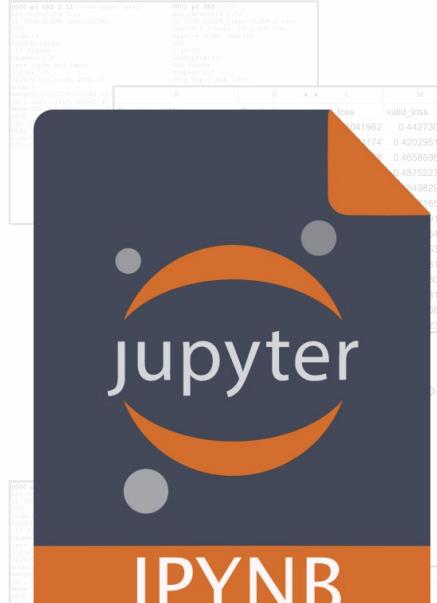


0000 p1 DRX 3.11 - not super eval
 best threshold 0.121
 2L_BORN_BLMST(basicLSTM)
 100 frames
 sigmoid
 AdamOptimizer
 100 frames
 zero input and label
 log(max,1.0)
 103300 training, 2000 CV
 model:
 weights0107224-005946.vtk
 (p1, loss= 1415, epoch: 40
 max iterations: 1000
 with 0.15% threads during CV
 2L_BORN_BLMST(basicLSTM)
 MEAN IBM SIMD GAIN: 2
 best traffic acc (50% data)
 2021-04-14 0.4919381174 0.4202951491
 with 0.15% threads during CV
 IBM SIMD GAIN: 2

best overall IOU (20% data)
 major-sweep-16
 swept-sweep-164
 silver-sweep-139
 laced-sweep-115
 eager-sweep-97
 rich-sweep-88
 hopeful-sweep-33
 autumn-sweep-24
 decent-sweep-21
 vague-sweep-5
 Second best acc

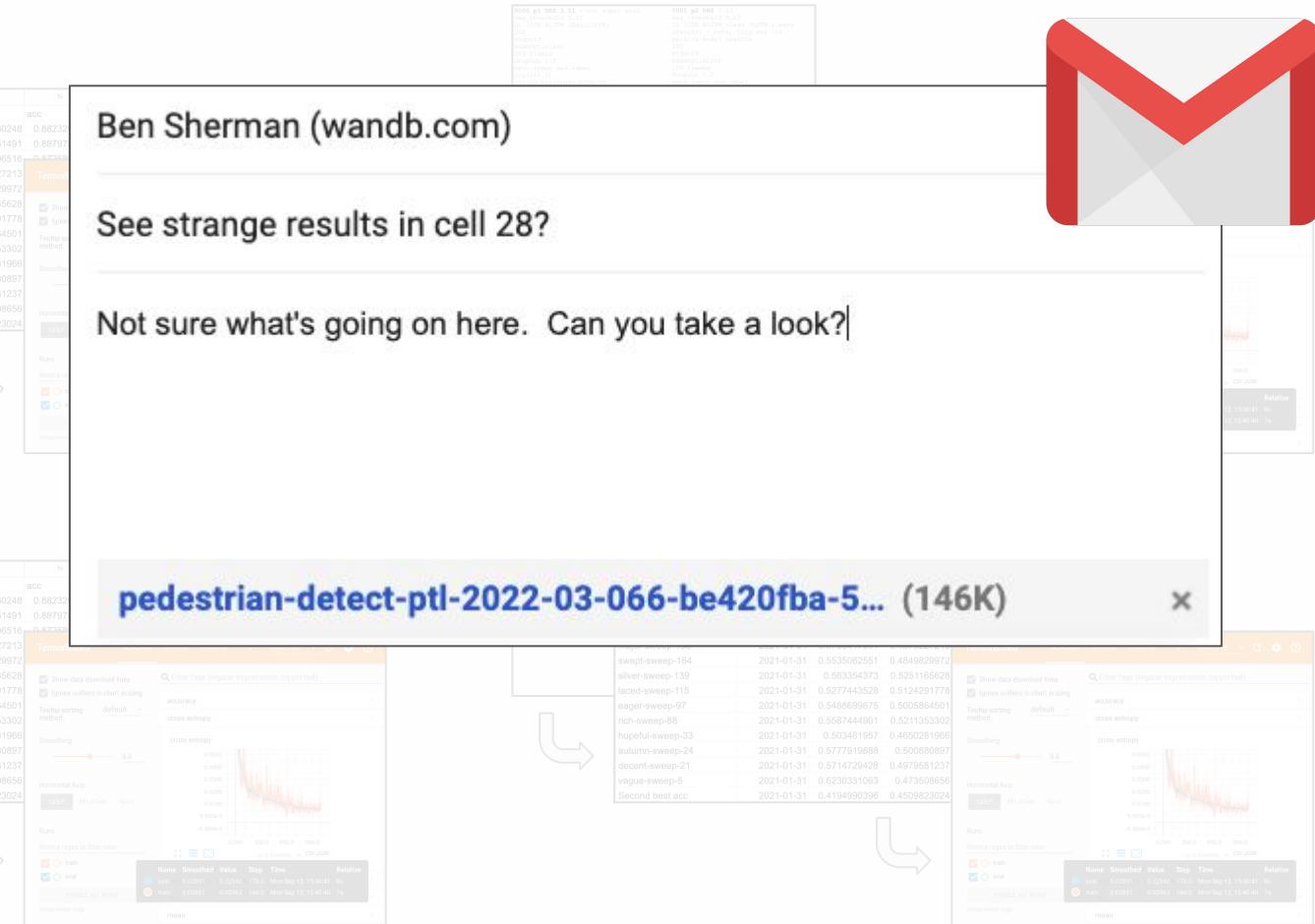


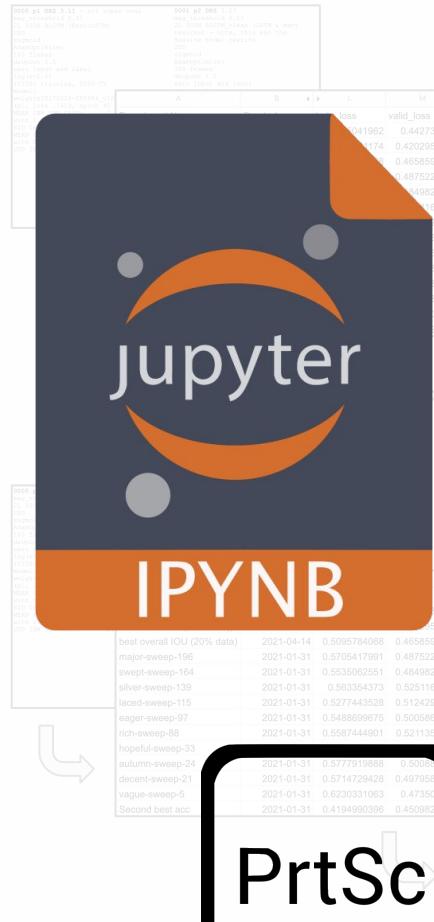






best overall IOU (20% data)	2021-04-14	0.5095784068	0.4658956516
major-sweep-196	2021-01-31	0.5705417981	0.4875227213
swept-sweep-164	2021-01-31	0.5535062591	0.4849829972
silver-sweep-139	2021-01-31	0.563534373	0.5251165628
laced-sweep-115	2021-01-31	0.5277443528	0.5124291778
eager-sweep-97	2021-01-31	0.5486696975	0.5005864501
rich-sweep-88	2021-01-31	0.5587444901	0.5211353302
hopeful-sweep-33	2021-01-31	0.5034619567	0.4650281966
autumn-sweep-24	2021-01-31	0.5777918888	0.500880897
descent-sweep-21	2021-01-31	0.5714729428	0.4979581237
vague-sweep-5	2021-01-31	0.6230331063	0.4735086656
Second best acc	2021-01-31	0.4194993936	0.4509823024





PrtSc

Ben Sherman (wandb.com)

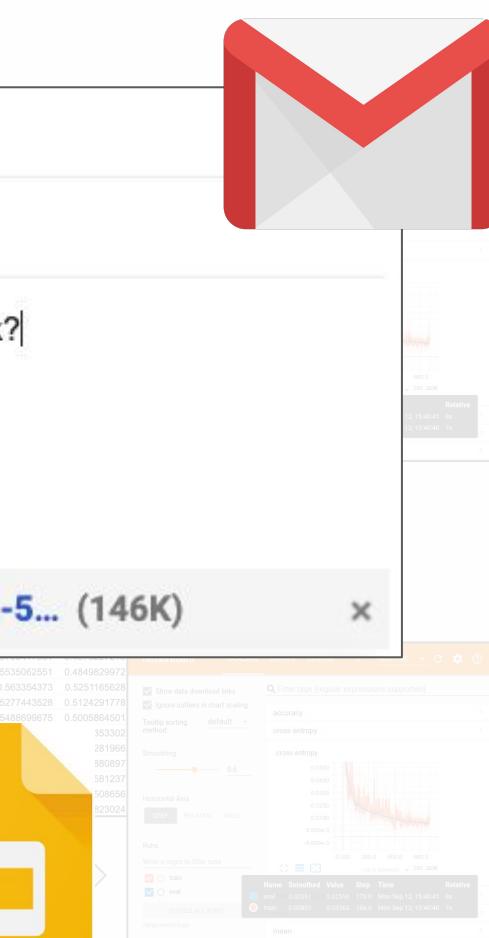
See strange results in cell 28?

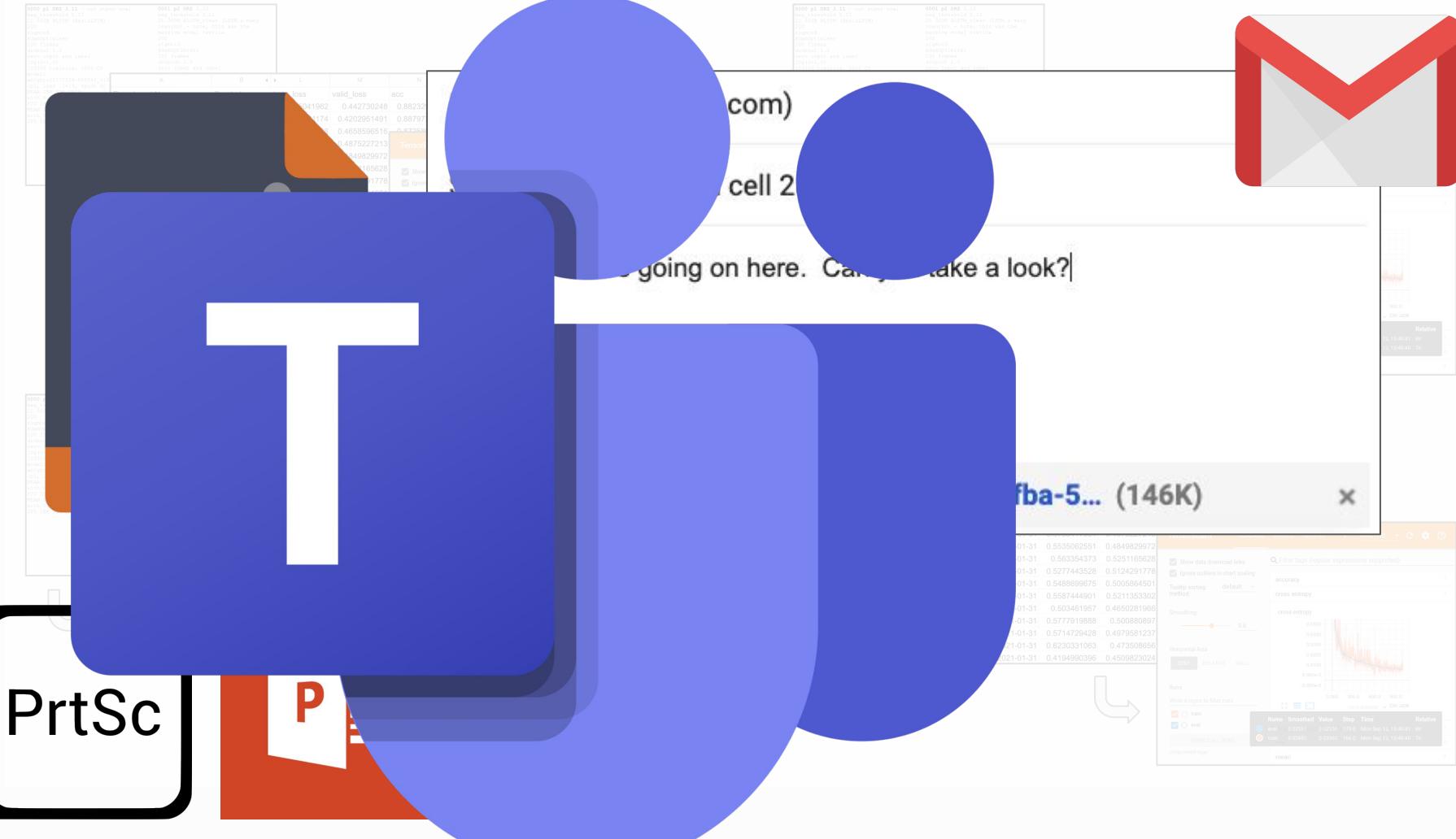
Not sure what's going on here. Can you take a look?

[pedestrian-detect-ptl-2022-03-066-be420fba-5... \(146K\)](#)

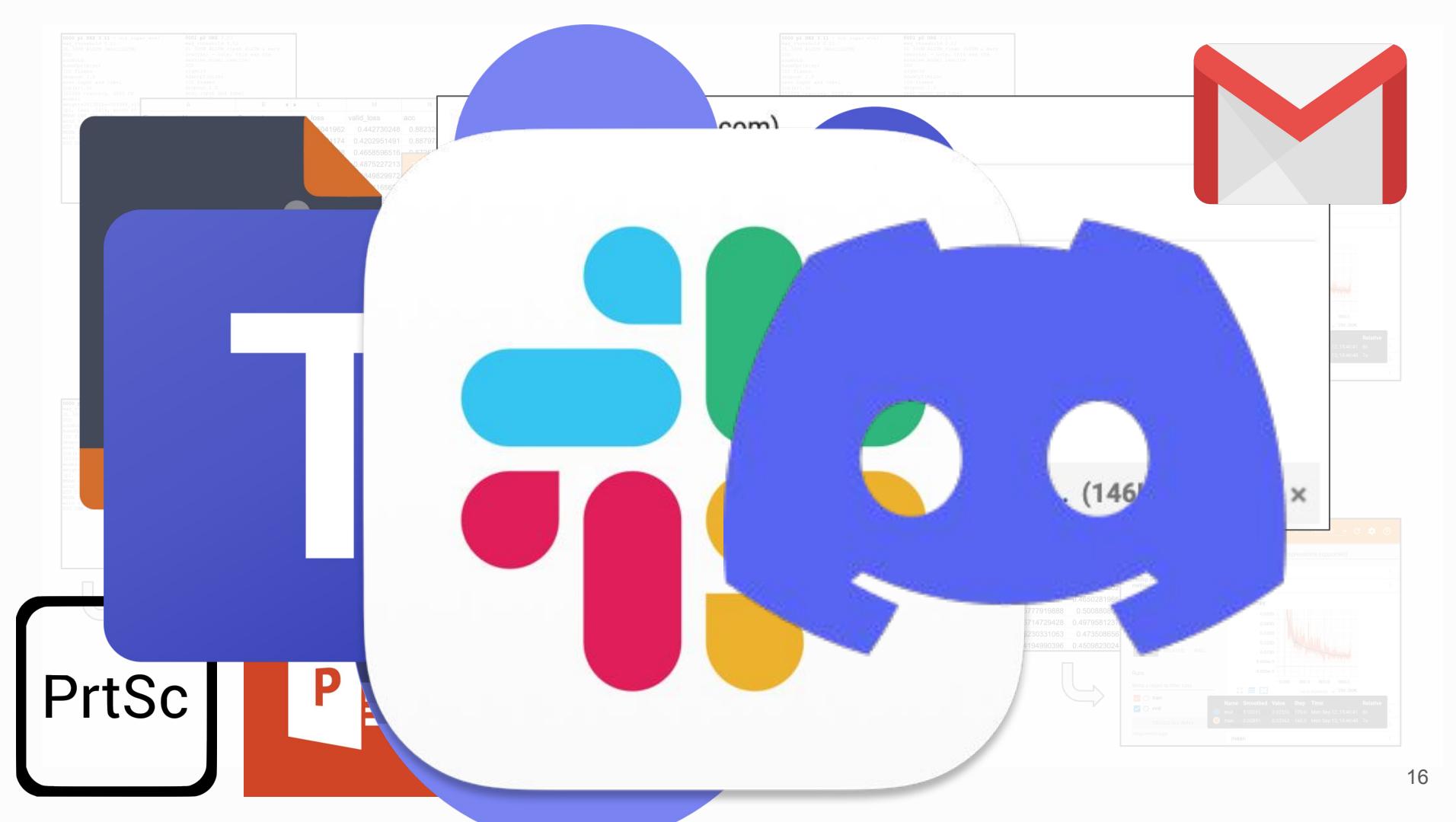


0000 pt 002 3-11	0001 pt 004 3-11	0002 pt 002 3-11	0003 pt 004 3-11
0000 pt 002 3-11	0001 pt 004 3-11	0002 pt 002 3-11	0003 pt 004 3-11
0000 pt 002 3-11	0001 pt 004 3-11	0002 pt 002 3-11	0003 pt 004 3-11
0000 pt 002 3-11	0001 pt 004 3-11	0002 pt 002 3-11	0003 pt 004 3-11
0000 pt 002 3-11	0001 pt 004 3-11	0002 pt 002 3-11	0003 pt 004 3-11









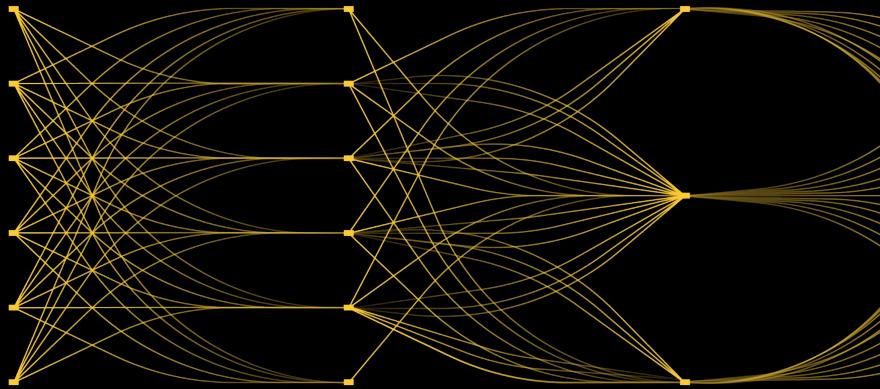


PrtScn



Weights & Biases

Just Stop



October 2022



Three principles of an ideal ML workflow



Rapidly iterate

to continuously refine
and optimize models



Reproduce

to reduce key-person
dependencies



Collaborate

to ensure knowledge transfer
across the organization



Three principles of an ideal ML workflow



Rapidly iterate

to continuously refine
and optimize models



Reproduce

to reduce key-person
dependencies



Collaborate

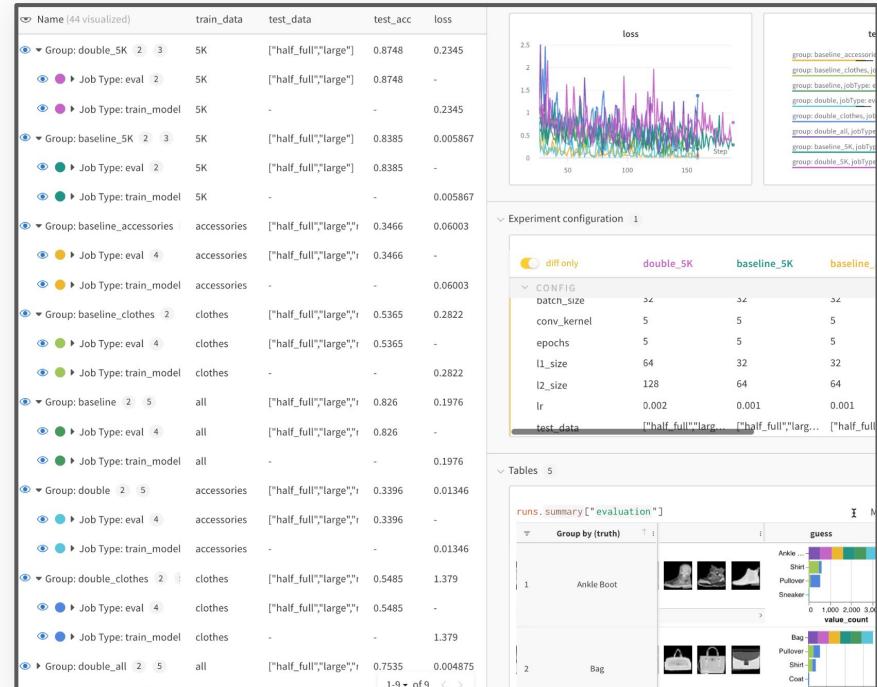
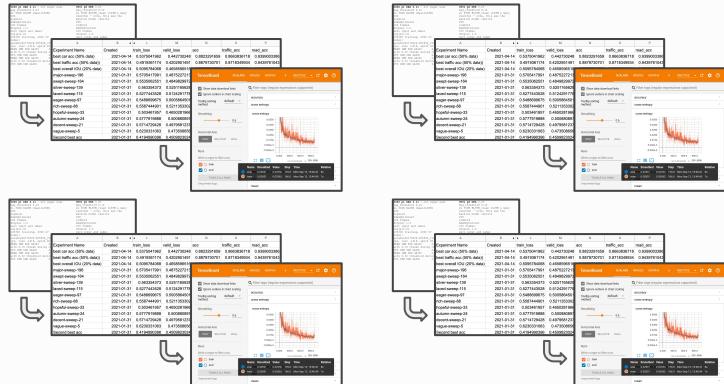
to ensure knowledge transfer
across the organization



A system of record for all ML workflows



A system of record for all ML workflows

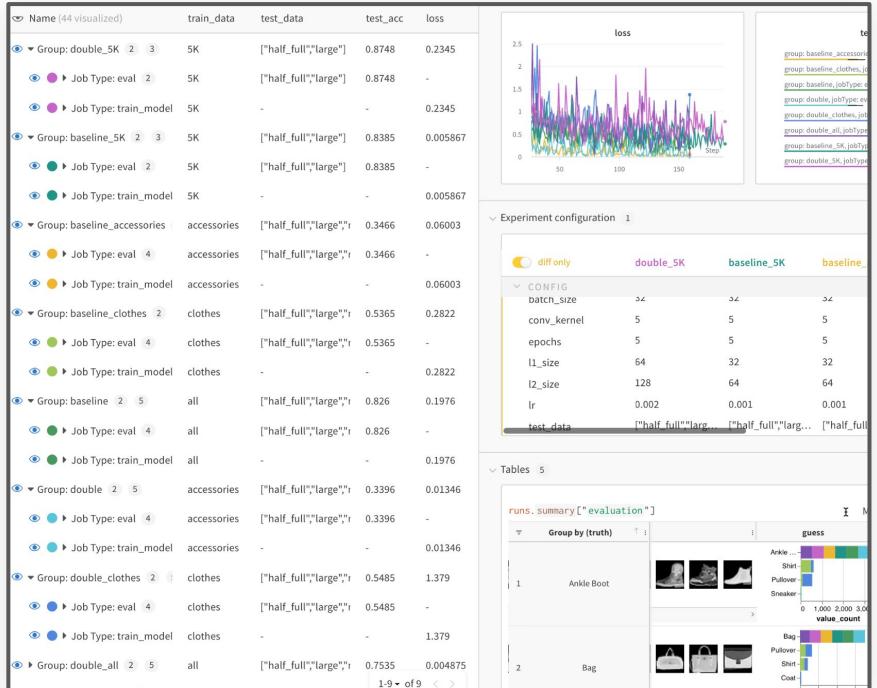


A system of record for all ML workflows



Get started in 60 seconds

```
!pip install wandb # Install W&B  
  
wandb.init() # Start experiment  
wandb.log(metrics) # Log metrics + more!
```

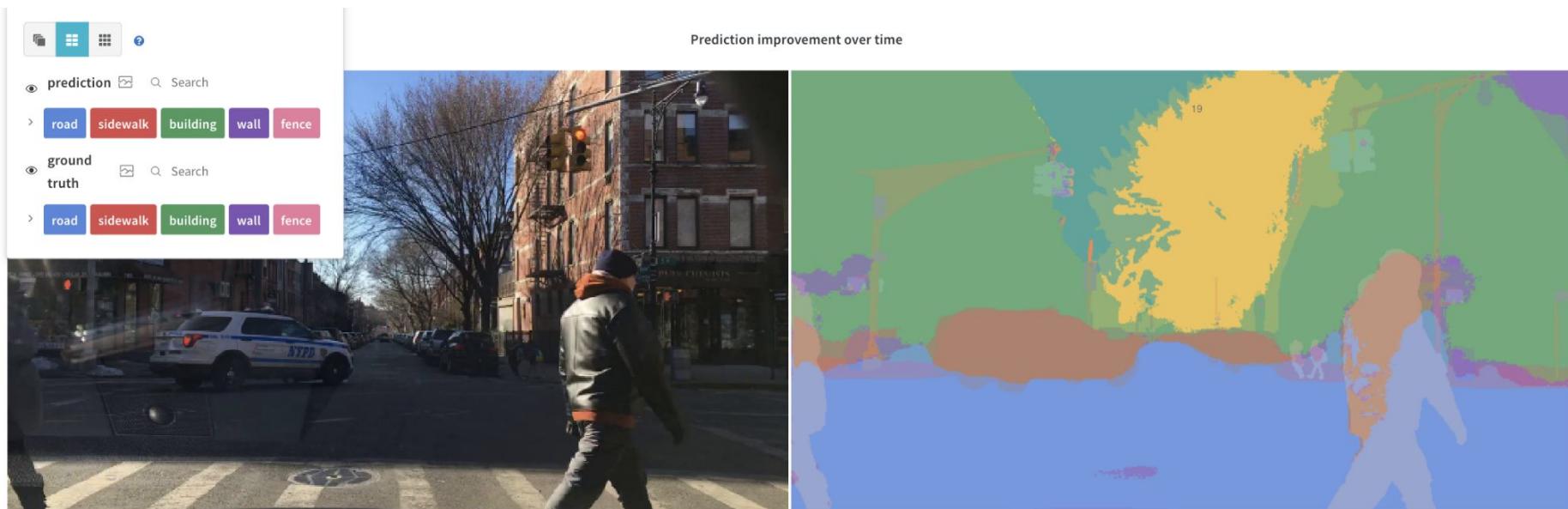




Yes, you really can get started in 60 seconds.

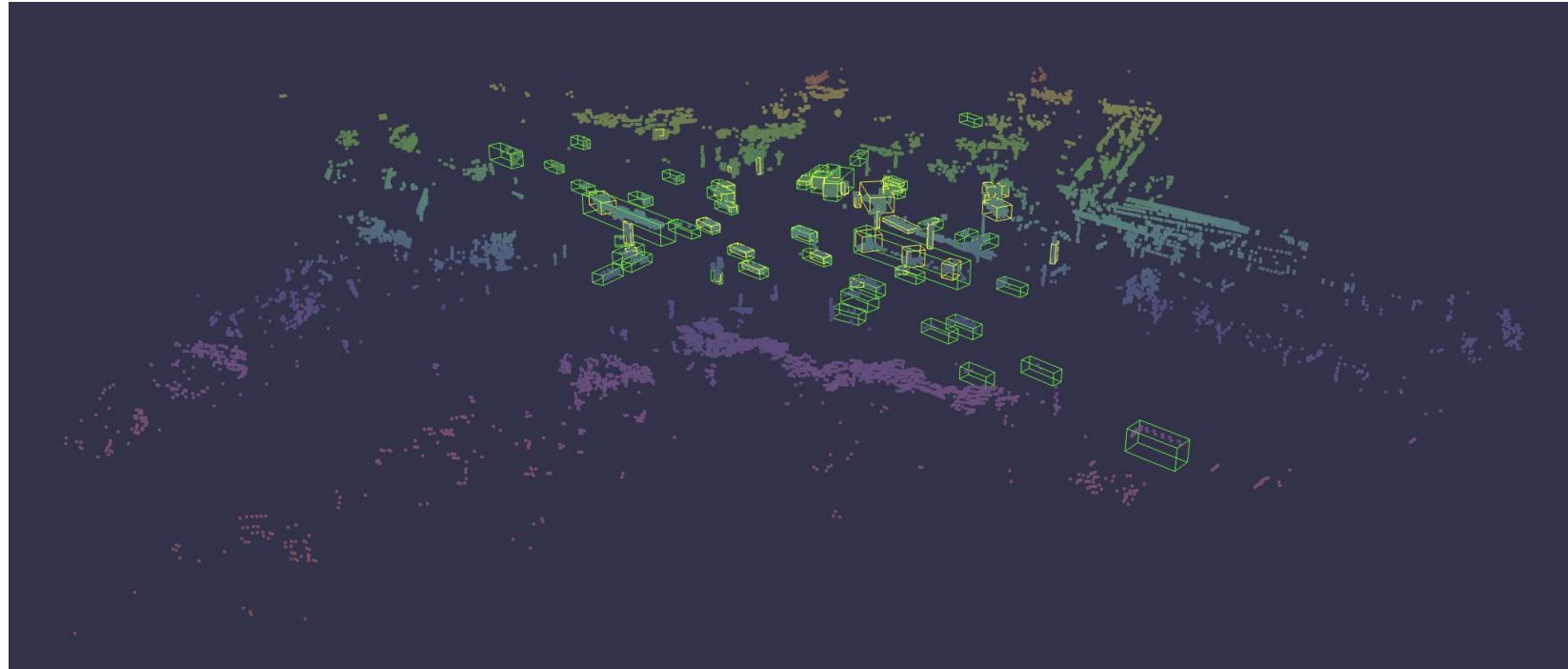
Experiment Tracking

A system of record for `wandb.Image`



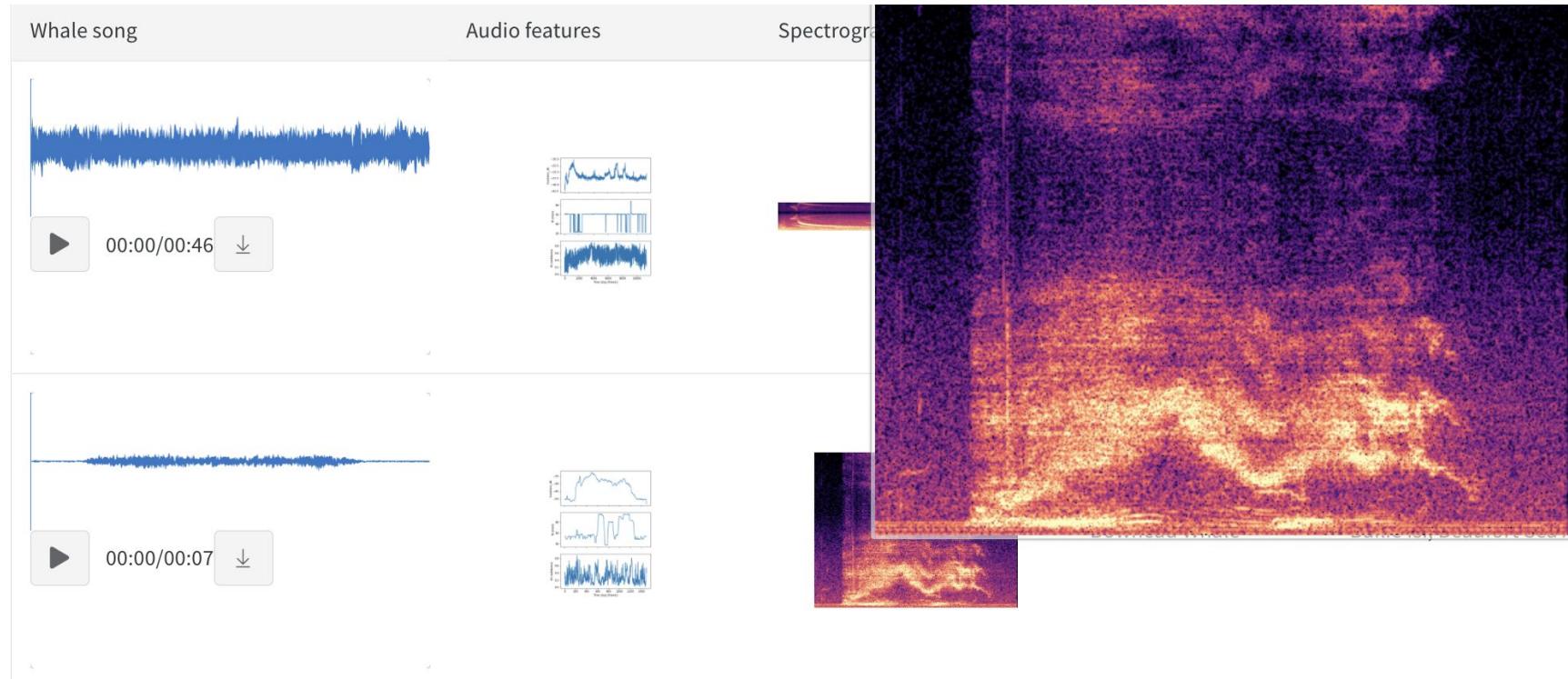
Experiment Tracking

A system of record for `wandb.Object3D`



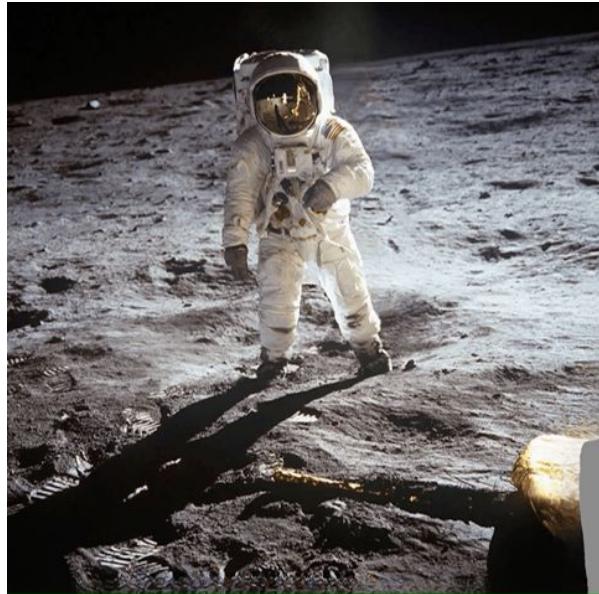
Experiment Tracking

A system of record for `wandb.Audio`





A system of record for `wandb.Video`



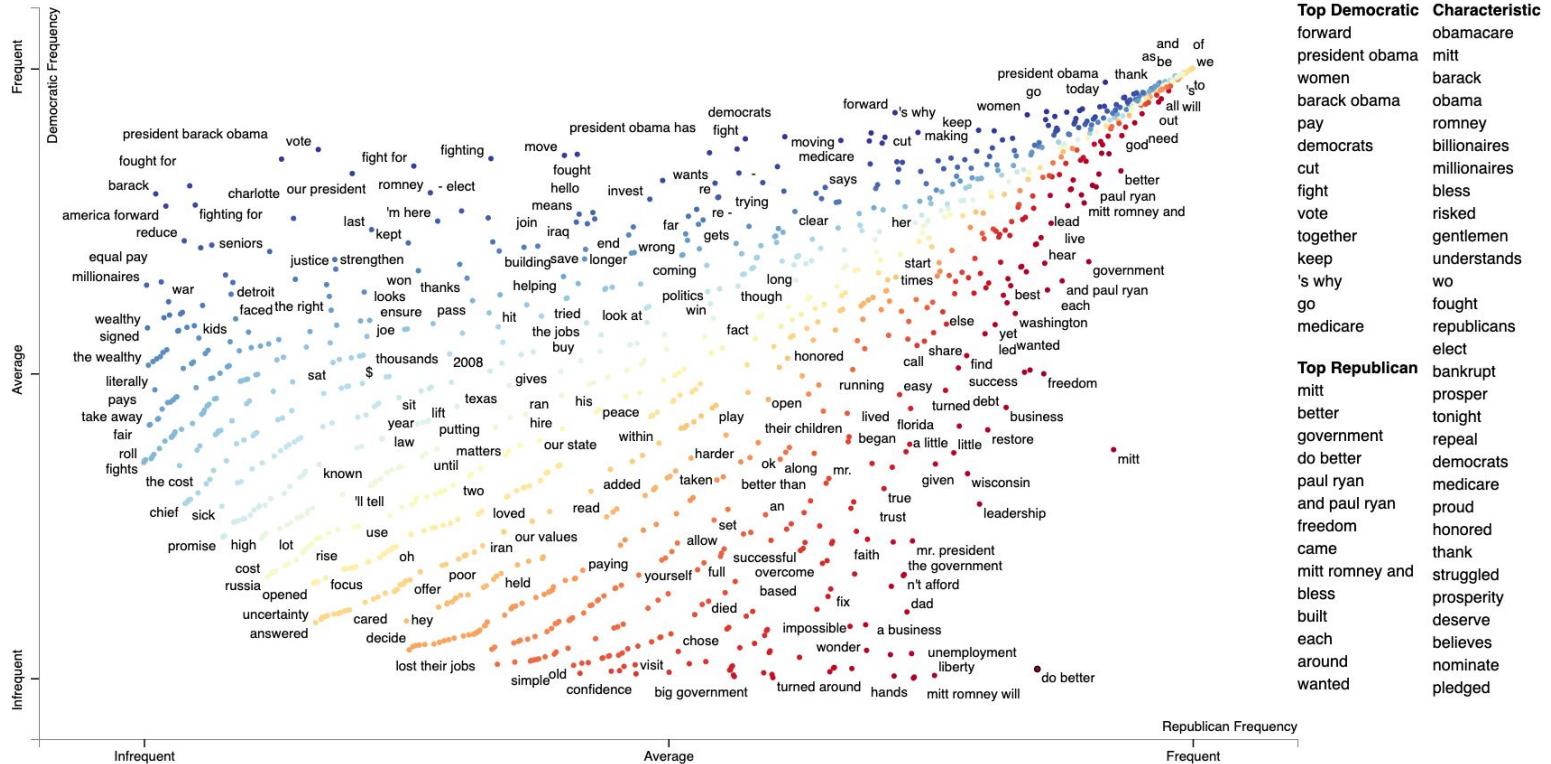
Experiment Tracking

A system of record for wandb.plots.POS

		displacy NER
365	Beyond Silicon Valley	LOCATION :Net Knows No Borders And Startups Abound
366	Uber	ORG Nears Deal for Self-Driving Car Simulation Startup: The Information
367	Shopin	ORG Blockchain Startup CEO Pleads Guilty to Fraud: NY LOCATION Attorney General
368	Medical Treatment Startup Gets a little Help from Hedge Funds	
369	Energy Dept. Aide Sees Delay in Reactor Startup	
370	Davis to	PERSON Return to Cosmos PRODUCT ' Startup Lineup Tonight; To Play Is the Thing
371	GM	ORG Lending \$40M to Startup Company Buying Closed Ohio LOCATION Plant
372	Intel	ORG Buys Israeli AI Startup Habana Labs ORG for \$2 Billion
373	Fintech Startup	Qwil ORG Raises \$24.4 Million in Equity and \$200 Million in Debt

Experiment Tracking

A system of record for wandb.html

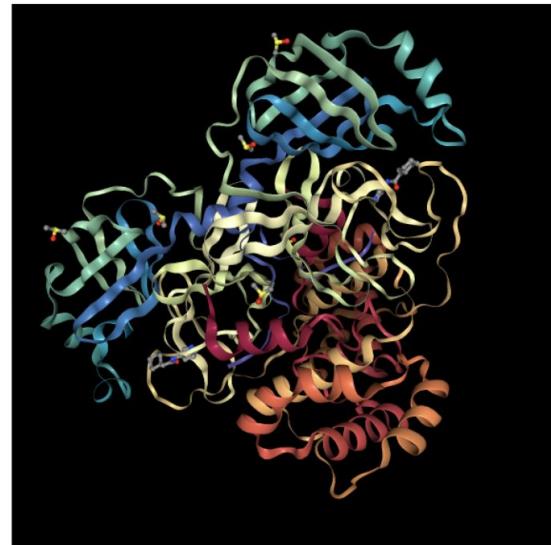




A system of record for all types of data!

wandb.Image
wandb.Object3D
wandb.log(wandb.Molecule)
wandb.Video
wandb.Html

COVID-19 main protease in complex with Z31792168





Ask questions about your data and models



gradio

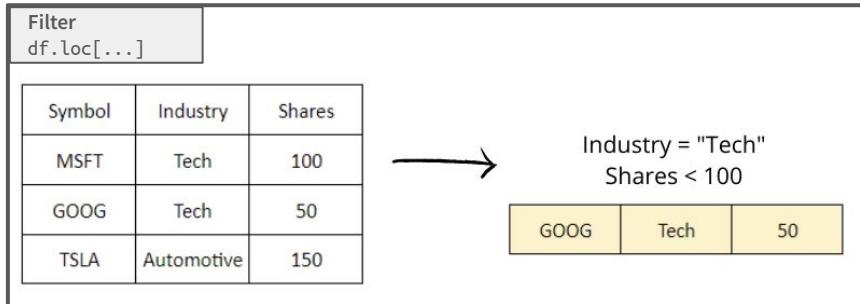
Streamlit



plotly | Dash

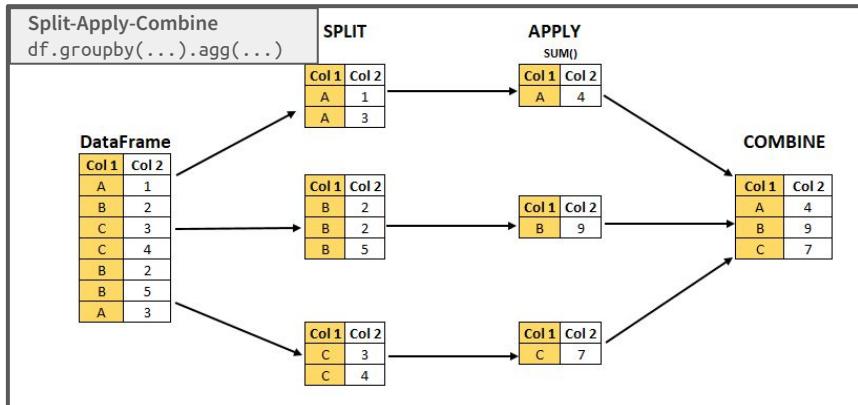
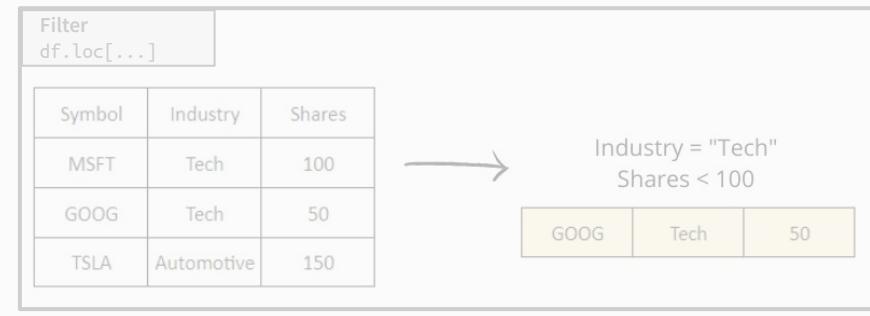
Tables – DataFrames with rich media support

Ask questions about your data (and models)



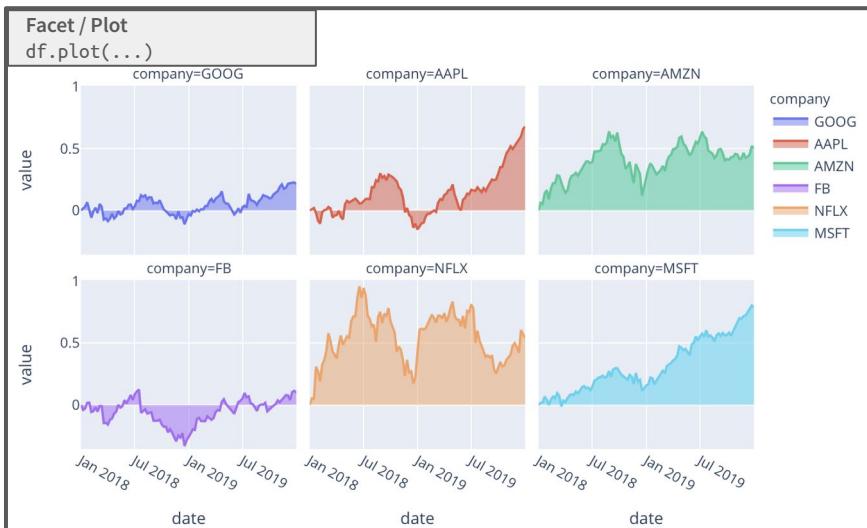
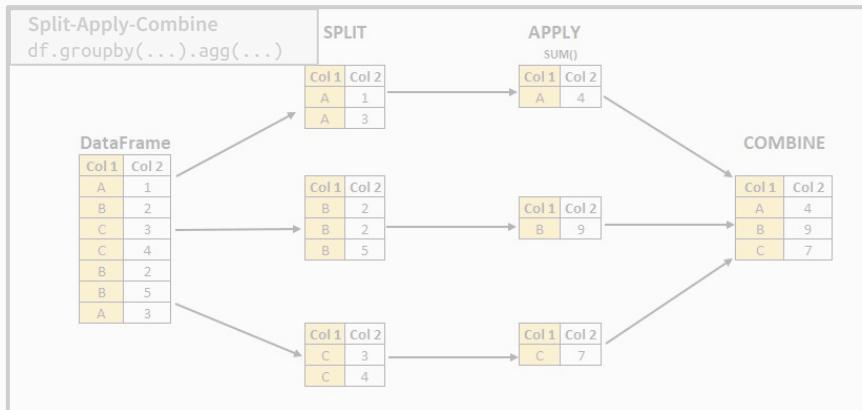
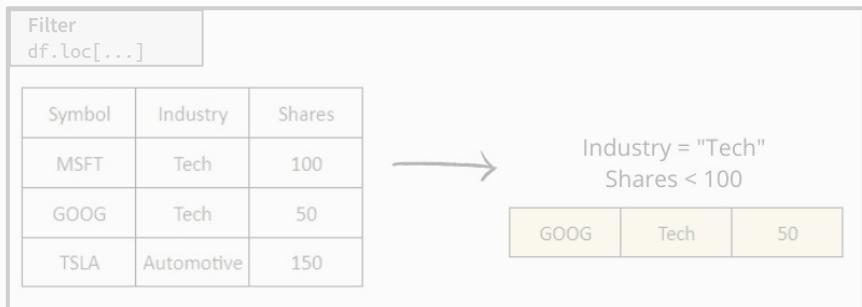
Tables – DataFrames with rich media support

Ask questions about your data (and models)



Tables – DataFrames with rich media support

Ask questions about your data (and models)





How do I ask questions about non-tabular data?

Tables – DataFrames with rich media support

Ask questions about your data (and models)

✗ Classic DataFrames

img	label
image mode=RGB size=500x333 at 0x7F56F6360ED0>	samoyed
image mode=RGB size=443x435 at 0x7F56F6379BD0>	shiba_inu
image mode=RGB size=600x437 at 0x7F56F6379E50>	Egyptian_Mau
image mode=RGB size=375x500 at 0x7F56F6379F90>	Birman
image mode=RGB size=500x375 at 0x7F56F637B350>	great_pyrenees
...	...
image mode=RGB size=500x335 at 0x7F56F623BD50>	Bengal
image mode=RGB size=403x500 at 0x7F56F623BF90>	leonberger
image mode=RGB size=500x375 at 0x7F56F6241390>	beagle
image mode=RGB size=192x288 at 0x7F56F62416D0>	Abyssinian
image mode=RGB size=288x300 at 0x7F56F6241910>	american_pit_bull_terrrier

Tables – DataFrames with rich media support

Ask questions about your data (and models)

✗ Classic DataFrames

img	label
image mode=RGB size=500x333 at 0x7F56F6360ED0>	samoyed
image mode=RGB size=443x435 at 0x7F56F6379BD0>	shiba_inu
image mode=RGB size=600x437 at 0x7F56F6379E50>	Egyptian_Mau
image mode=RGB size=375x500 at 0x7F56F6379F90>	Birman
image mode=RGB size=500x375 at 0x7F56F637B350>	great_pyrenees
...	...
image mode=RGB size=500x335 at 0x7F56F623BD50>	Bengal
image mode=RGB size=403x500 at 0x7F56F623BF90>	leonberger
image mode=RGB size=500x375 at 0x7F56F6241390>	beagle
image mode=RGB size=192x288 at 0x7F56F62416D0>	Abyssinian
image mode=RGB size=288x300 at 0x7F56F6241910>	american_pit_bull_terrrier

✓ wandb.Table

	img	label
1		samoyed
2		shiba_inu
3		Egyptian_Mau
4		Birman

Tables – DataFrames with rich media support

Ask questions about your data (and models)

✓ wandb.Table

	img	label
1		samoyed
2		shiba_inu
3		Egyptian_Mau
4		Birman

Table.groupby("label")

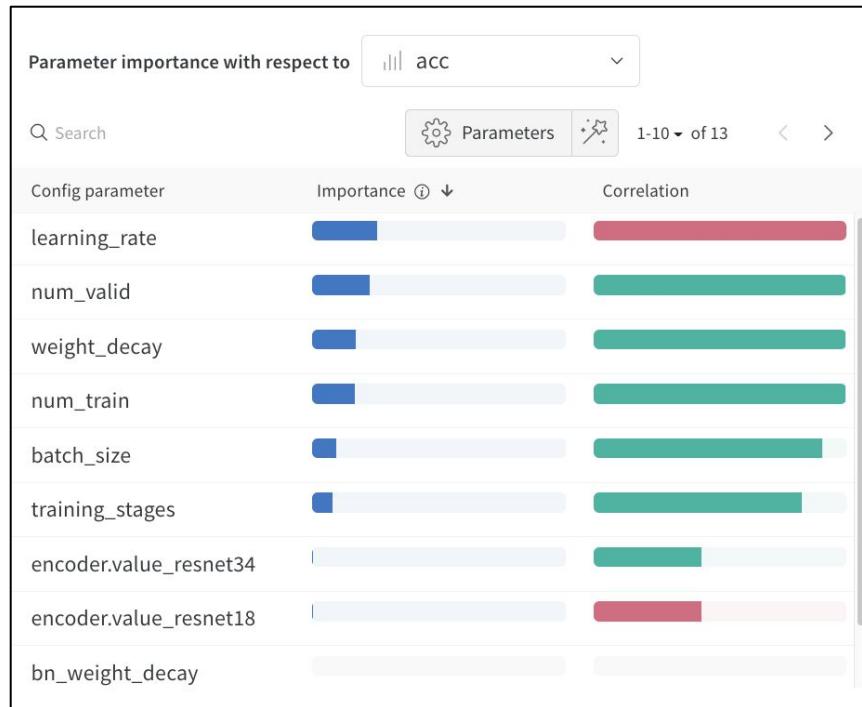
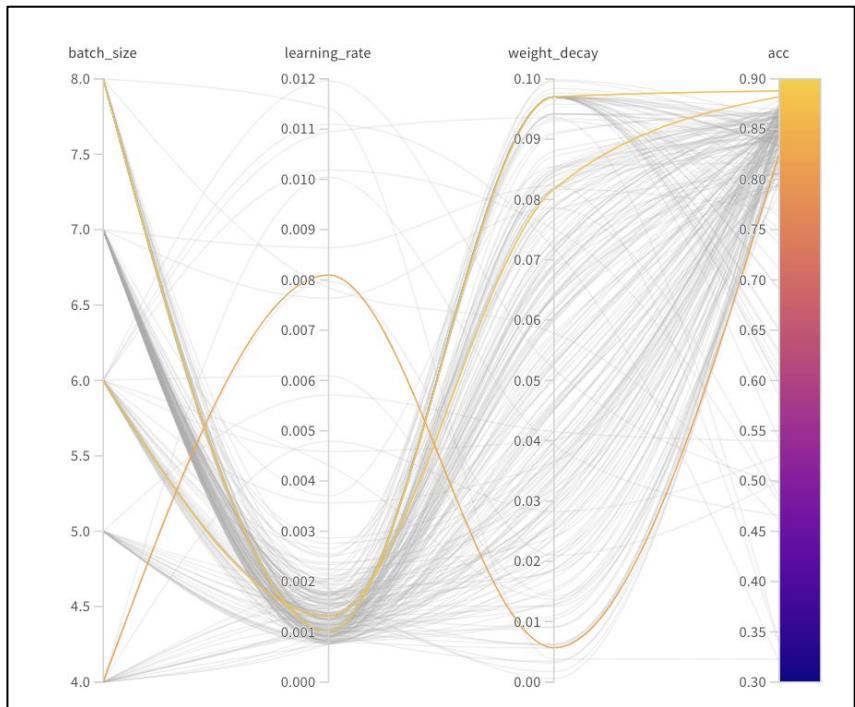
	Group by (label)	img
1	japanese_chin	    
2	Ragdoll	    
3	leonberger	    
4	shiba_inu	    



How do I ask questions about training?



Easily and systematically search hyperparameters

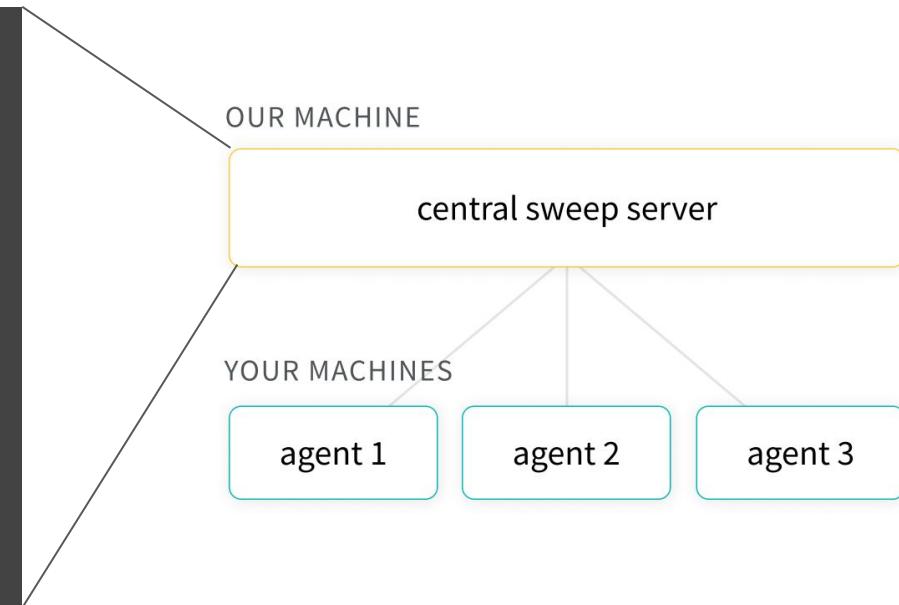




Sweeps – Easily search hyperparameters

Ask questions about your data (and models)

```
program: train.py
method: bayes
metric:
  name: valid_loss
  goal: minimize
parameters:
  batch_size:
    values: [32, 64]
  mixup_alpha:
    values: [0.2, 0.5, 0.8]
  optimizer:
    values: ["adam", "ranger"]
  encoder:
    values: ["resnet18", "resnet34", "resnet50", "resnet101"]
```





Three principles of an ideal ML workflow



Rapidly iterate

to continuously refine
and optimize models



Reproduce

to reduce key-person
dependencies



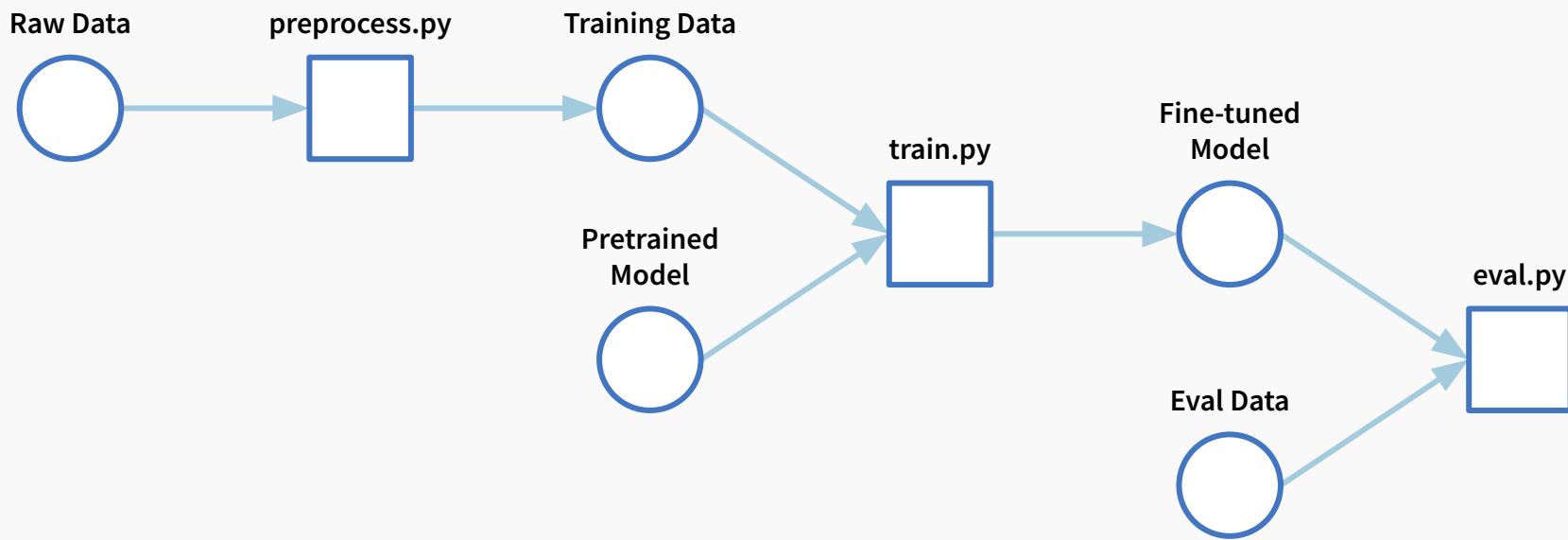
Collaborate

to ensure knowledge transfer
across the organization

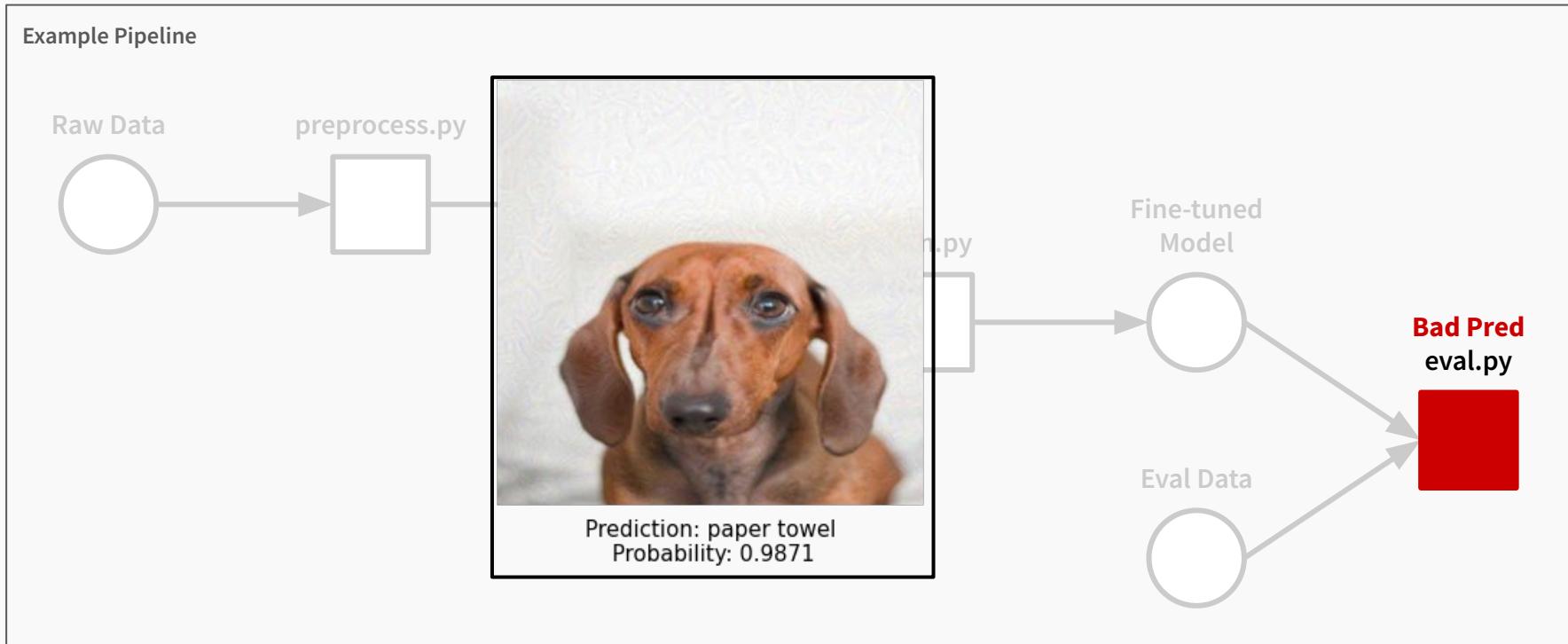


Artifacts – Version data and models across your pipeline

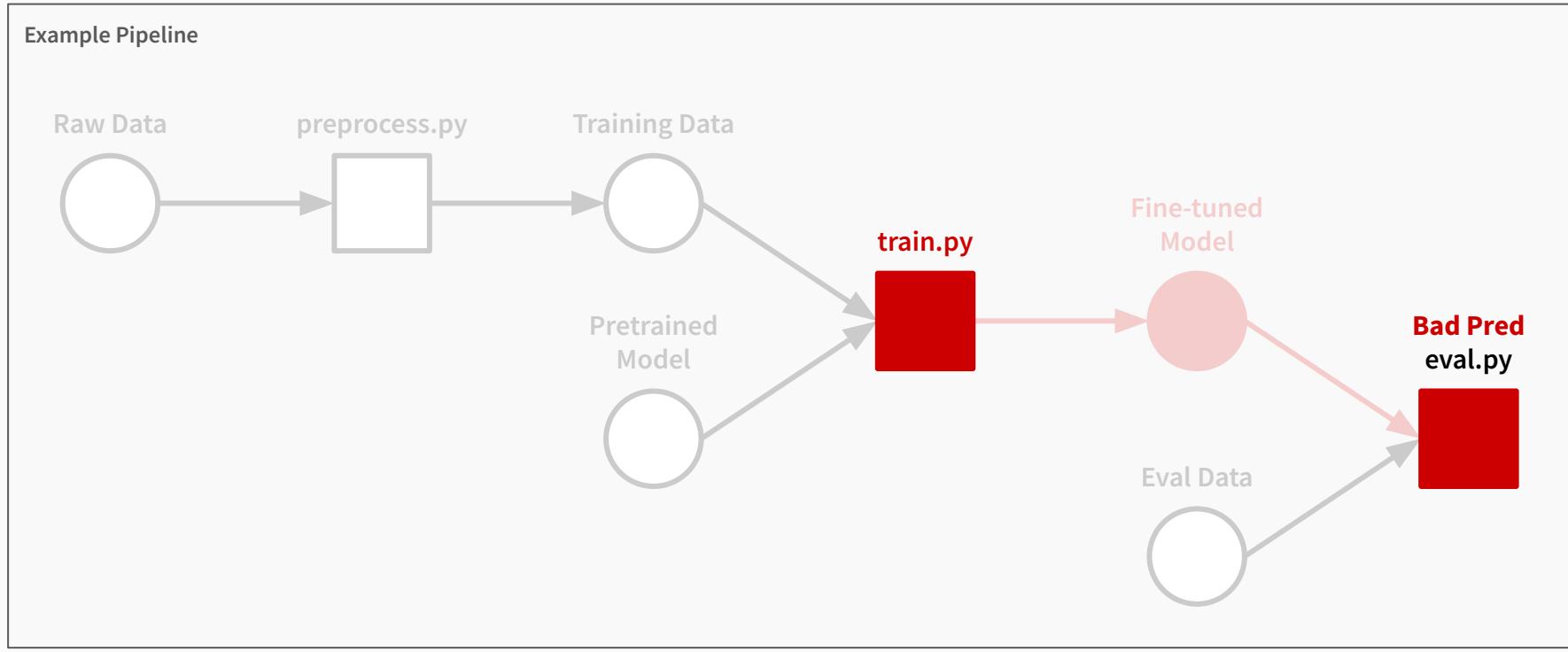
Example Pipeline



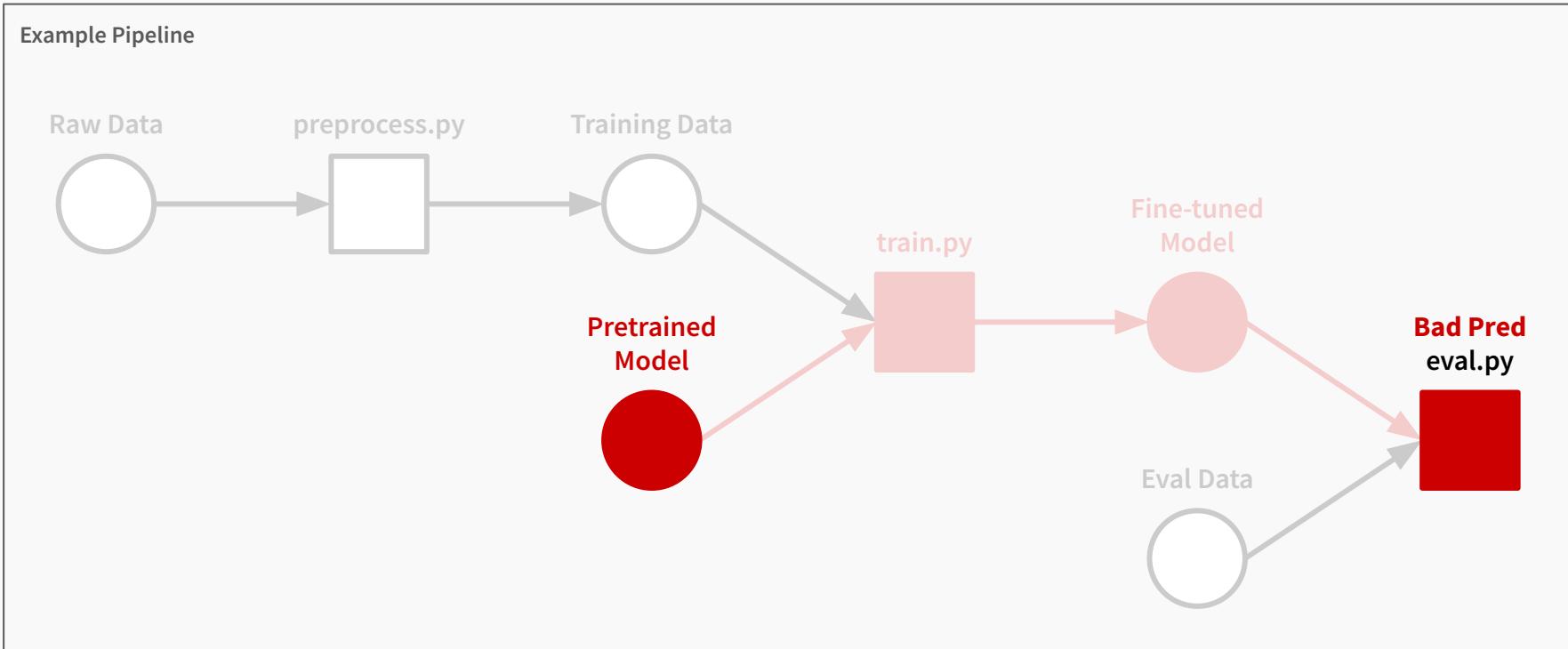
Bad predictions!



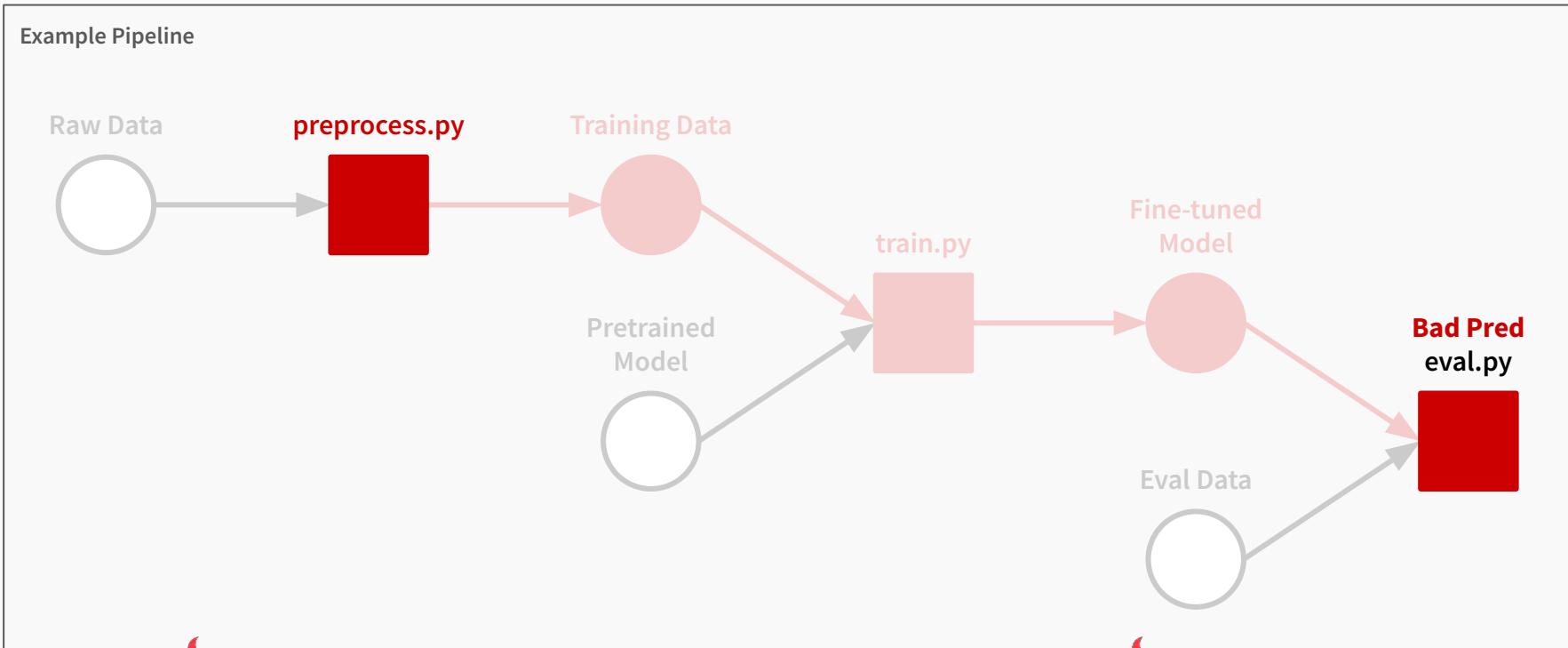
Bad predictions because of **training?**



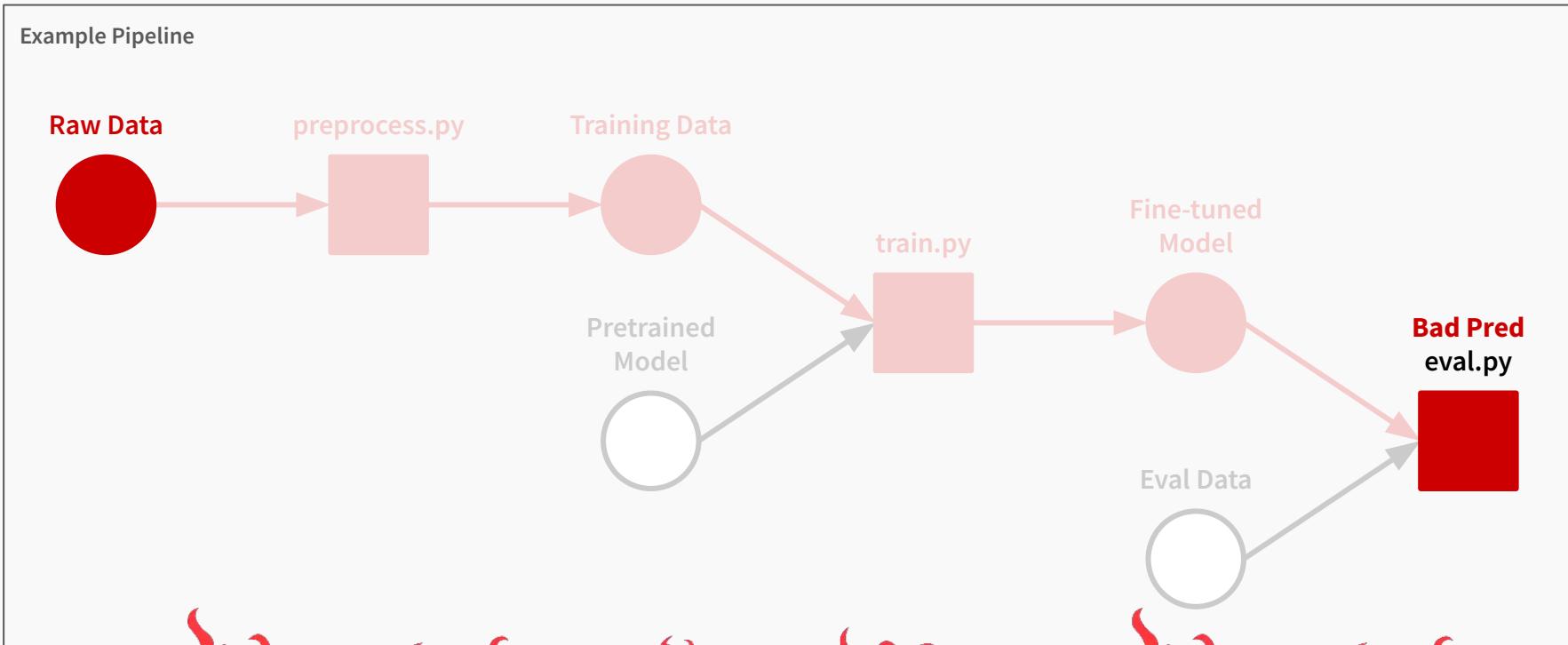
Bad predictions because of **pretrained models?**



Bad predictions because of **preprocessing?**

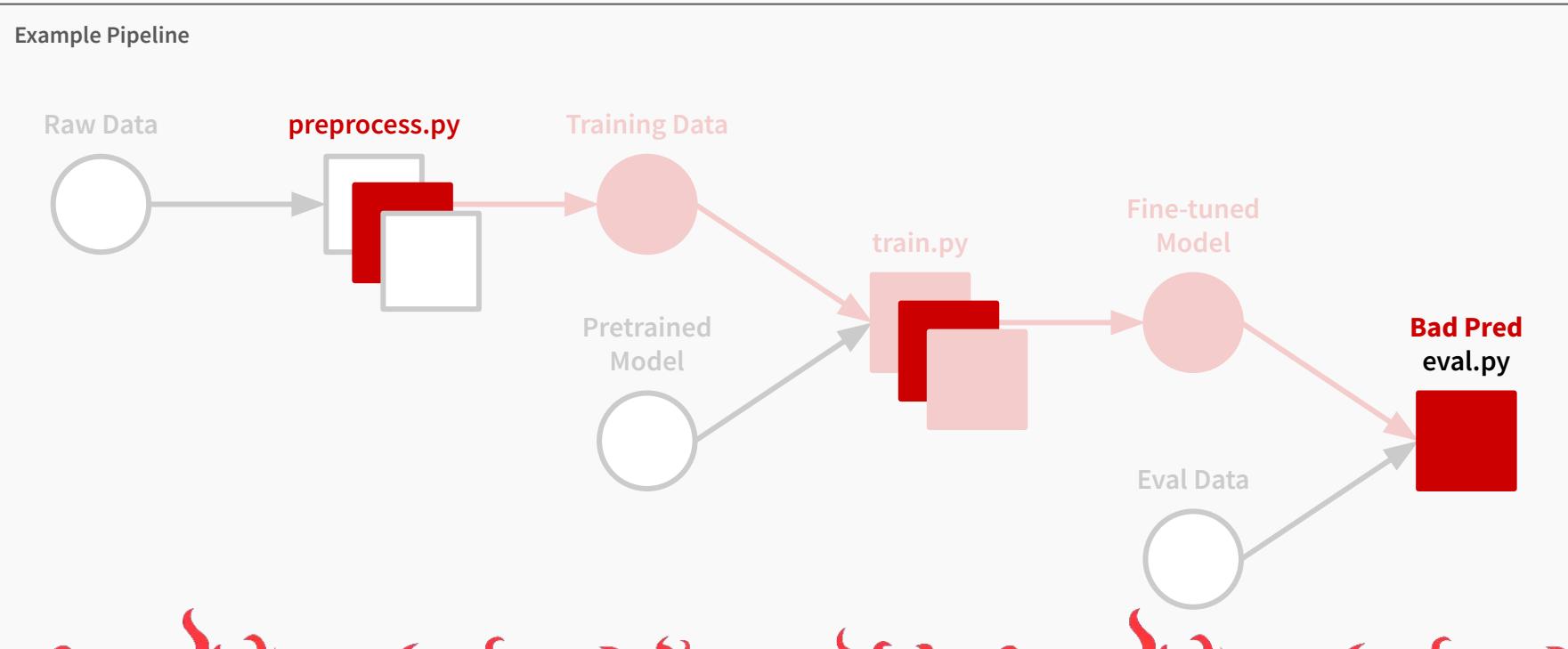


Bad predictions because of **raw data?**



All inputs - Version of a model across your pipeline

Bad predictions because of **multiprocessing?**



Bad predictions because of **multiprocessing?**

Example Pip

Raw Data



Bad predictions because of multithreading?

Example Pipeline

THIS IS FINE.

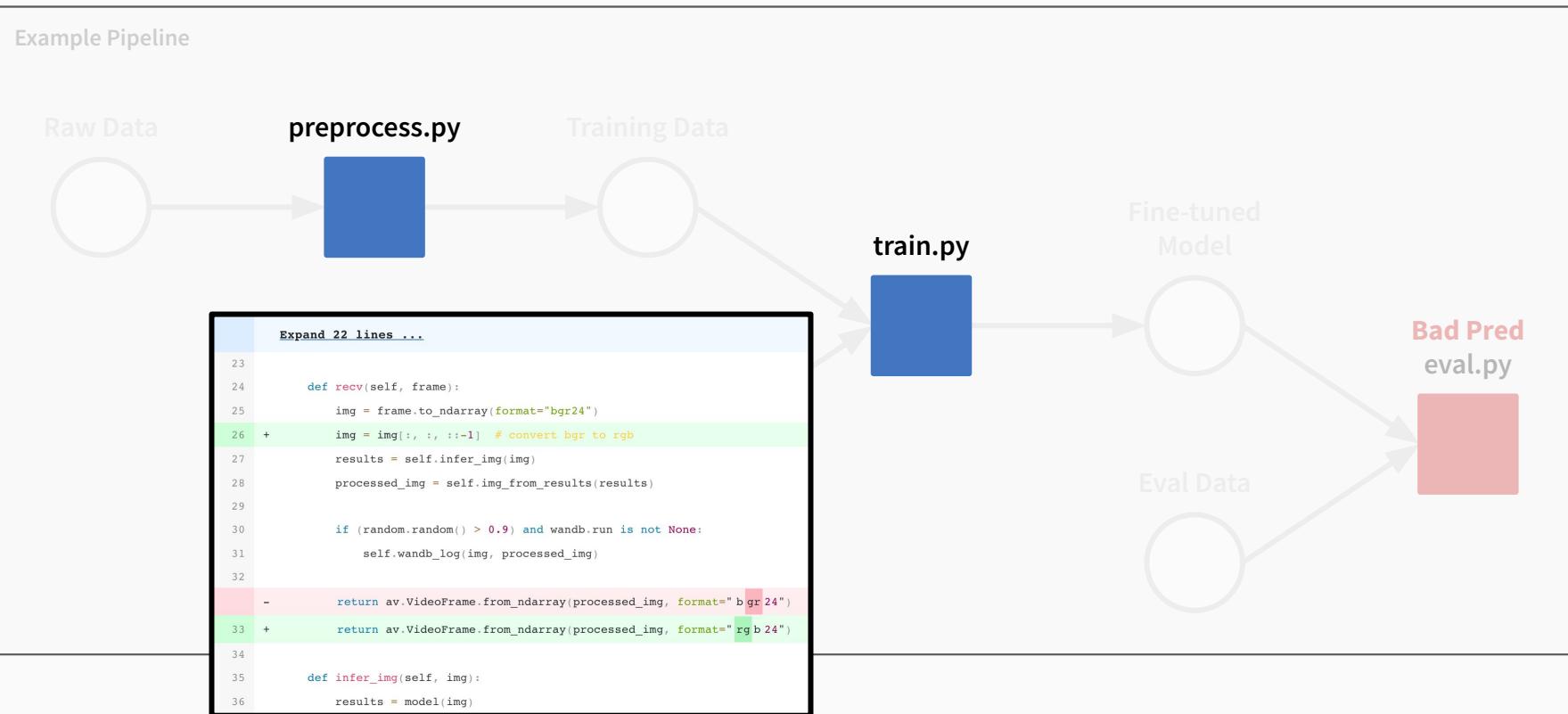
How do you debug a model pipeline?



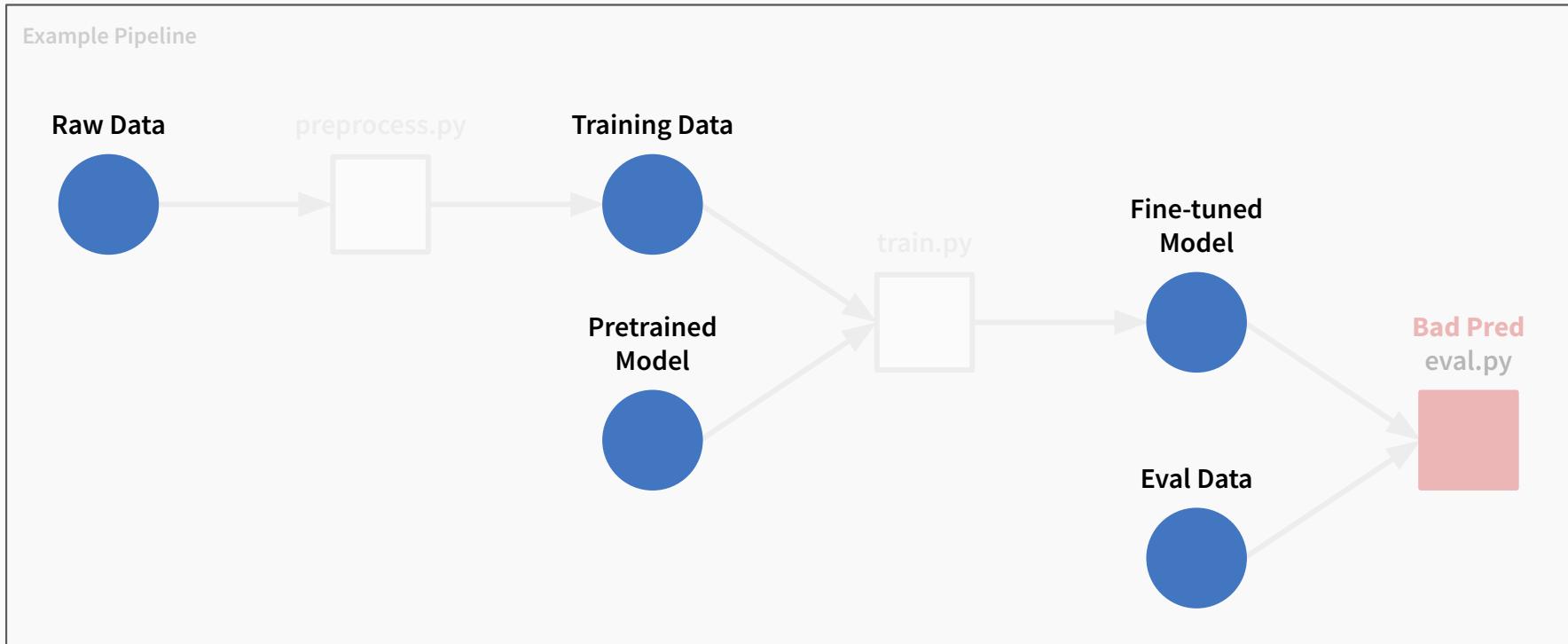


Check the Code

Problem with code? More than `git diff`



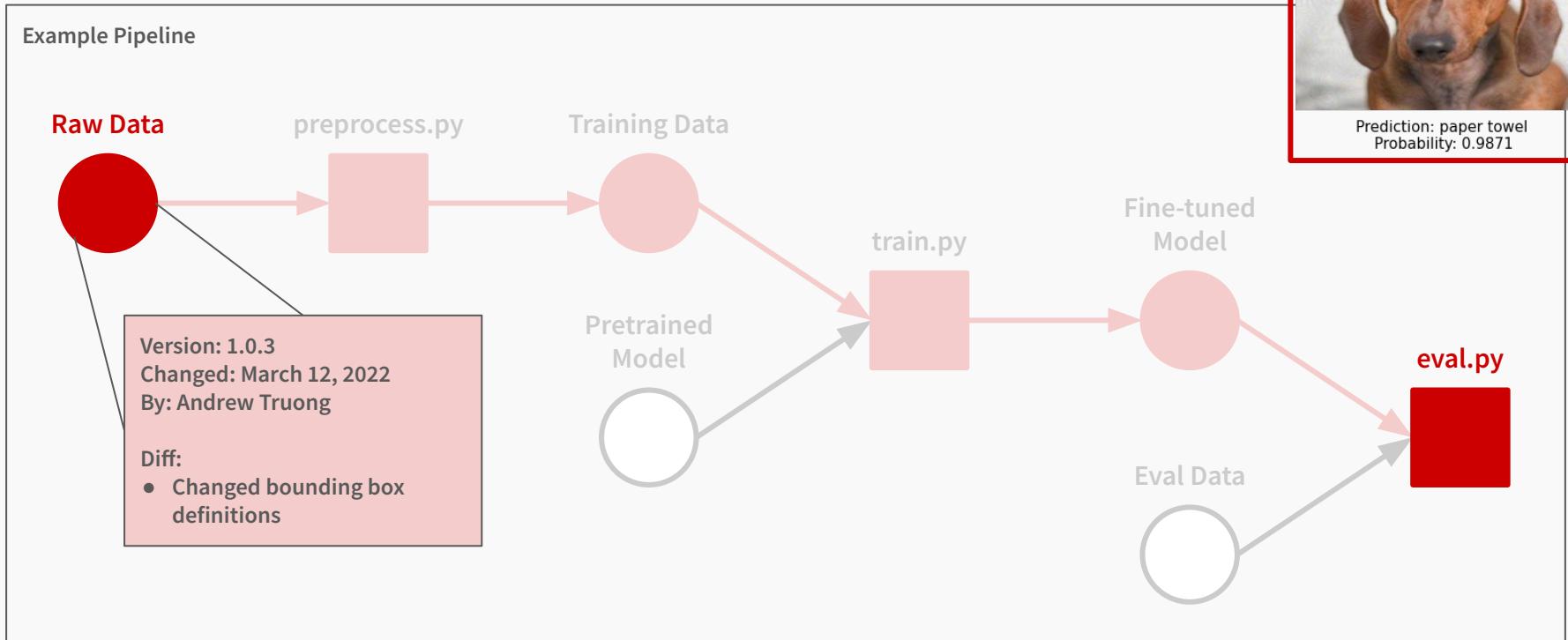
What about the input/output artifacts?



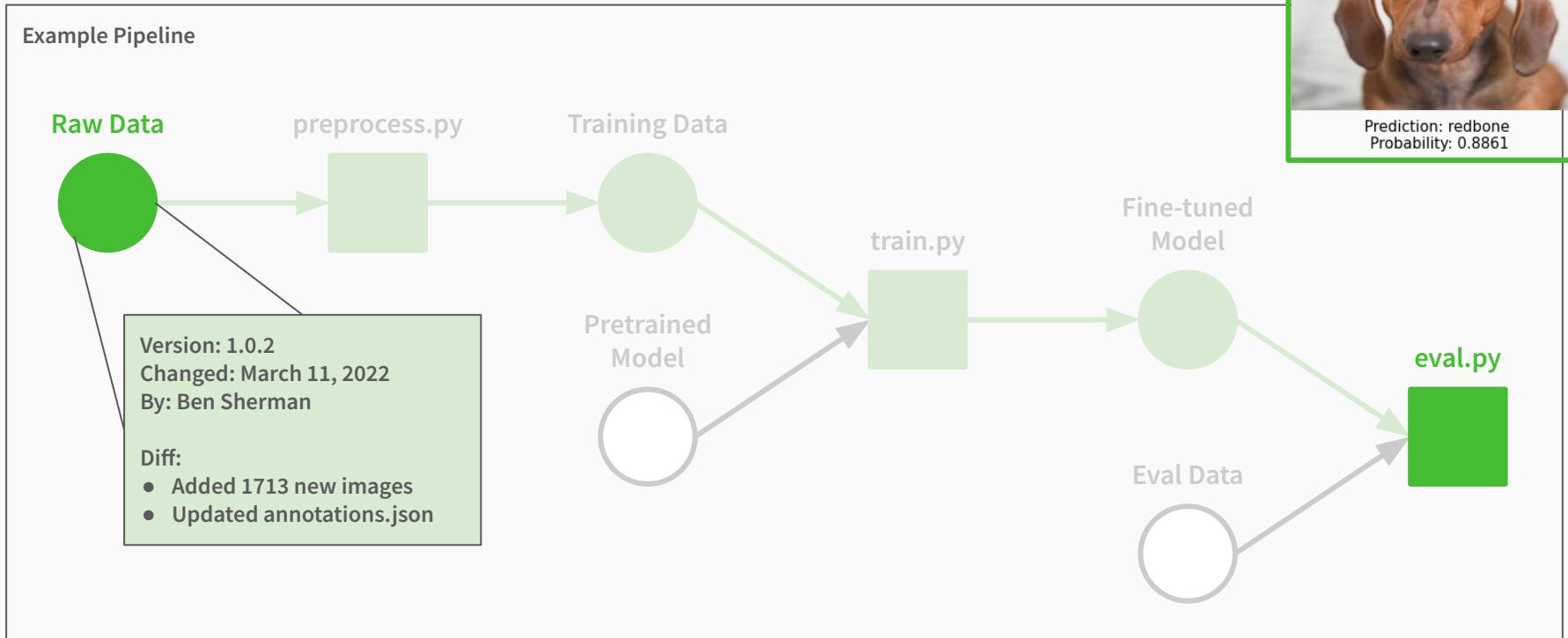


Check the inputs

Bad predictions because of **Raw Data**



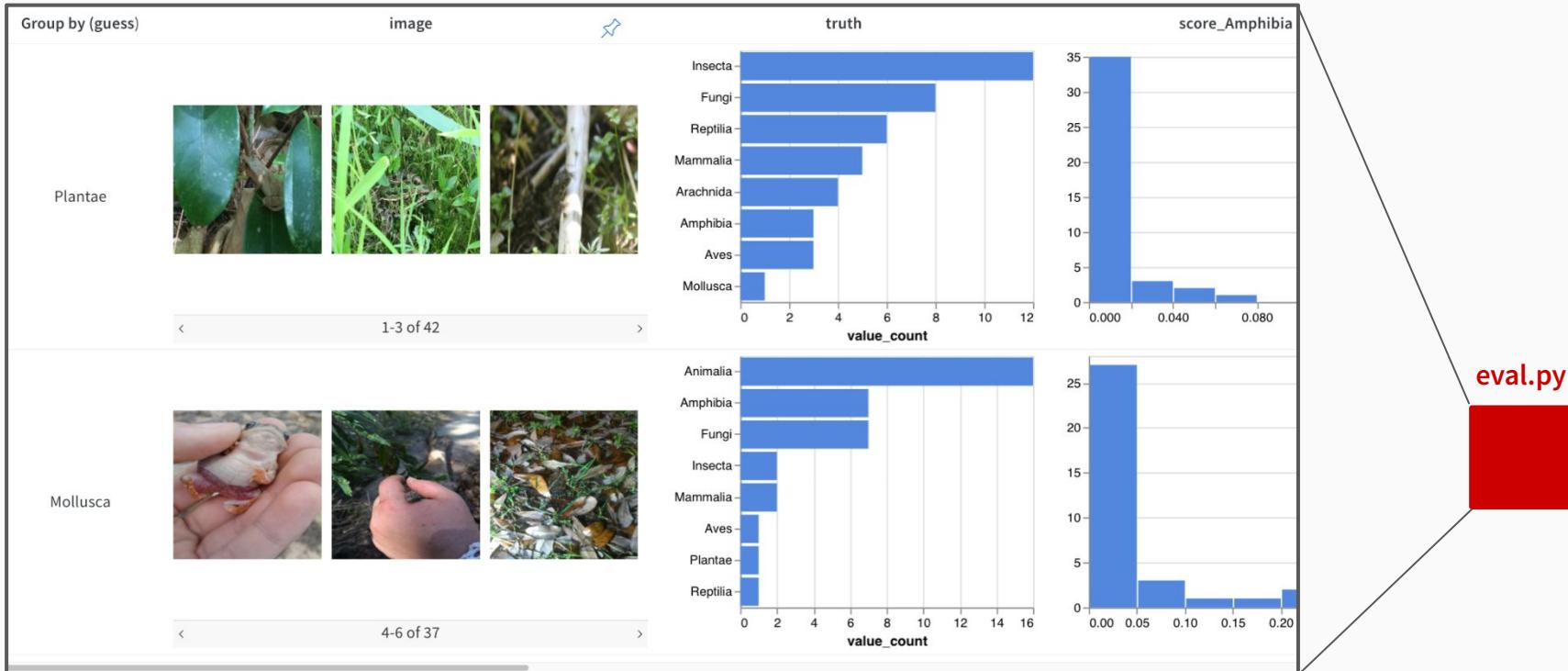
Good predictions after reverting Raw Data





Check the outputs

Tables – DataFrames with rich media support





Three principles of an ideal ML workflow



Rapidly iterate

to continuously refine
and optimize models



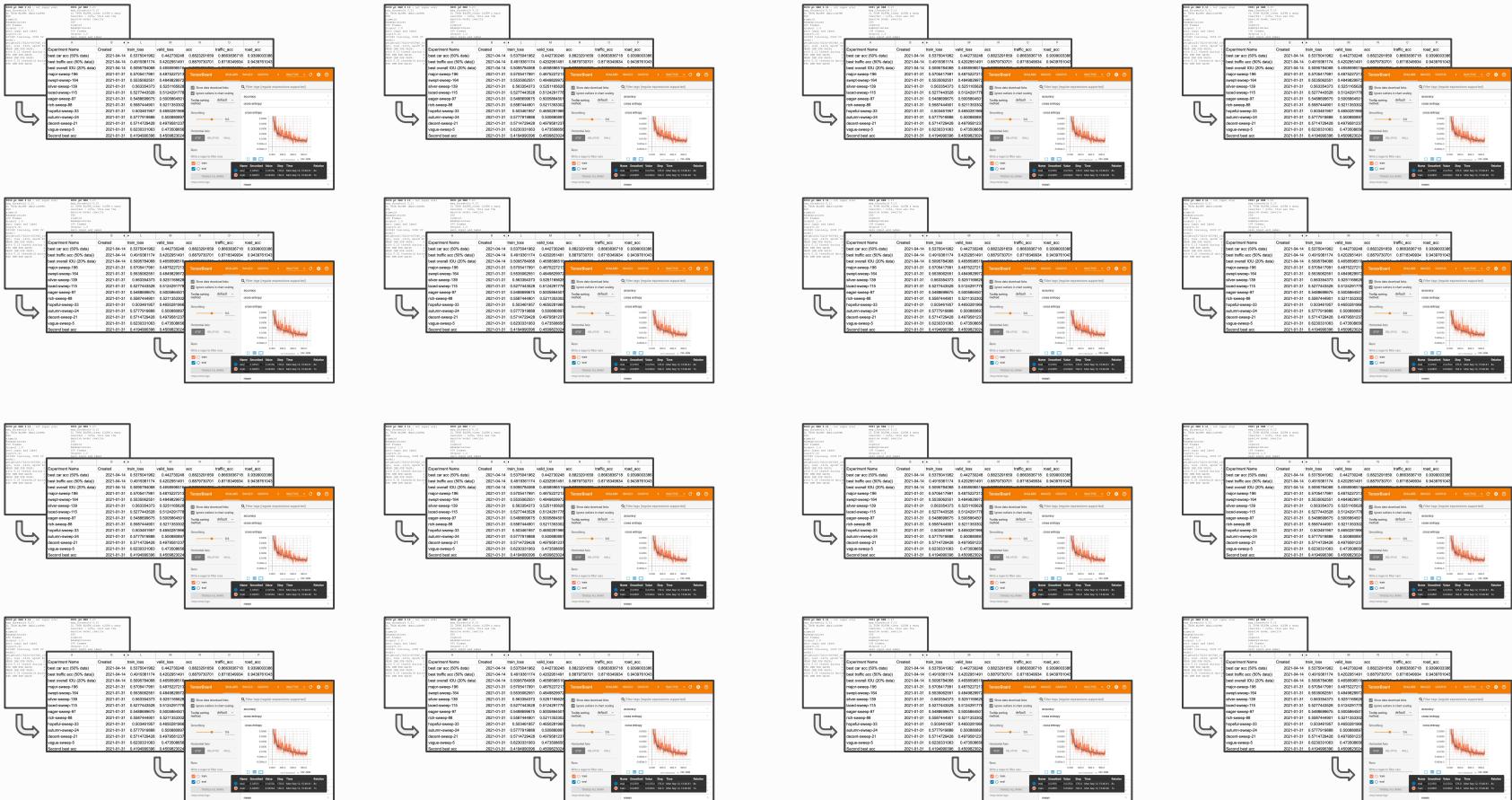
Reproduce

to reduce key-person
dependencies



Collaborate

to ensure knowledge transfer
across the organization



Name (44 visualized)	train_data	test_data	test_acc	loss
Group: double_5K 2 3	5K	["half_full","large"]	0.8748	0.2345
Job Type: eval 2	5K	["half_full","large"]	0.8748	-
Job Type: train_model	5K	-	-	0.2345
Group: baseline_5K 2 3	5K	["half_full","large"]	0.8385	0.005867
Job Type: eval 2	5K	["half_full","large"]	0.8385	-
Job Type: train_model	5K	-	-	0.005867
Group: baseline_accessories	accessories	["half_full","large"]	0.3466	0.06003
Job Type: eval 4	accessories	["half_full","large"]	0.3466	-
Job Type: train_model	accessories	-	-	0.06003
Group: baseline_clothes	clothes	["half_full","large"]	0.5365	0.2822
Job Type: eval 4	clothes	["half_full","large"]	0.5365	-
Job Type: train_model	clothes	-	-	0.2822
Group: baseline	2 5	all	0.826	0.1976
Job Type: eval 4	all	["half_full","large"]	0.826	-
Job Type: train_model	all	-	-	0.1976
Group: double	2 5	accessories	0.3396	0.01346
Job Type: eval 4	accessories	["half_full","large"]	0.3396	-
Job Type: train_model	accessories	-	-	0.01346
Group: double_clothes	2 4	clothes	0.5485	1.379
Job Type: eval 4	clothes	["half_full","large"]	0.5485	-
Job Type: train_model	clothes	-	-	1.379
Group: double_all	2 5	all	0.7535	0.004875

1-9 ▾ of 9 < >

loss

test_acc

Experiment configuration 1

+ Add Panel

	diff only	double_5K	baseline_5K	baseline_accessories	baseline_clothes
CONFIG					
batch_size	32	32	32	32	32
conv_kernel	5	5	5	5	5
epochs	5	5	5	5	5
l1_size	64	32	32	32	32
l2_size	128	64	64	64	64
lr	0.002	0.001	0.001	0.001	0.001
test_data	["half_full","large"]	["half_full","large"]	["half_full","large"]	["half_full","large"]	["half_full","large"]

Tables 5

+ Add Panel

runs.summary ["evaluation"]		Merge Tables: Table	⚙️
Group by (truth)			
1	Ankle Boot	guess	s_Ankle Boot
2	Bag	guess	s_Bag

Interactive Dashboards

Check any metrics



Gourab 6:26 PM

Cool report, what about policy loss and CPU utilization?



Andrew Truong 6:27 PM

I didn't think it was important – let me get back to you later...



Gourab 6:28 PM

It seems you don't think anything I suggest as important.

Interactive Dashboards

Check any metrics



Gourab 6:26 PM

Cool report, what about policy loss and CPU utilization?



Andrew Truong 6:27 PM

I didn't think it was important — lo

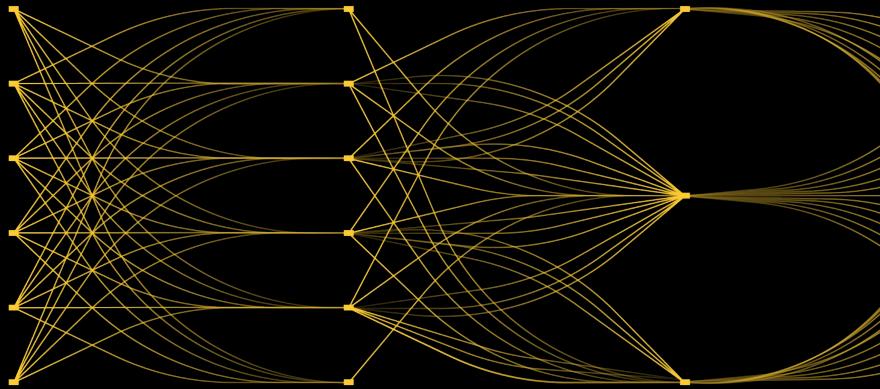
The screenshot shows a user interface for managing dashboards. On the left, there's a sidebar with a list of projects: brain-tumor-test-viz, g2net-cqt, trace, hf-flax-alberti-poetry, hf-flax-gpt2-tamil, hf-flax-transcoder, hf-flax-roberta-marathi, custom_yolov5, cleanRL, and nutorch_mnist_recognizer. On the right, there are two search dropdowns: 'Search projects' and 'Search panels'. The 'Search panels' dropdown is open, showing a list of panels grouped under 'charts': charts/episode_reward, charts/episode_reward/AttackRe..., charts/episode_reward/Produce..., charts/episode_reward/Produce..., charts/episode_reward/ProduceWorkerRe..., charts/episode_reward/Resourc..., charts/episode_reward/WinLoss..., charts/learning_rate, and charts/sps. At the bottom of the interface are two buttons: 'Import panel' and 'Add panel'.



Unified Reporting and Dashboarding



Open Source Research





The world's leading ML teams trust us



COATUE INSIGHT BOND





The top open source research orgs use us



EleutherAI



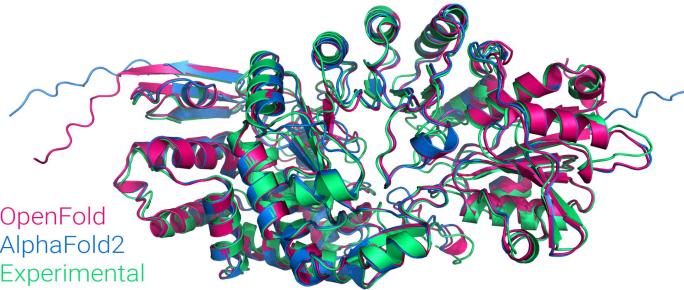
LAION



CarperAI



Harmonai



OpenFold

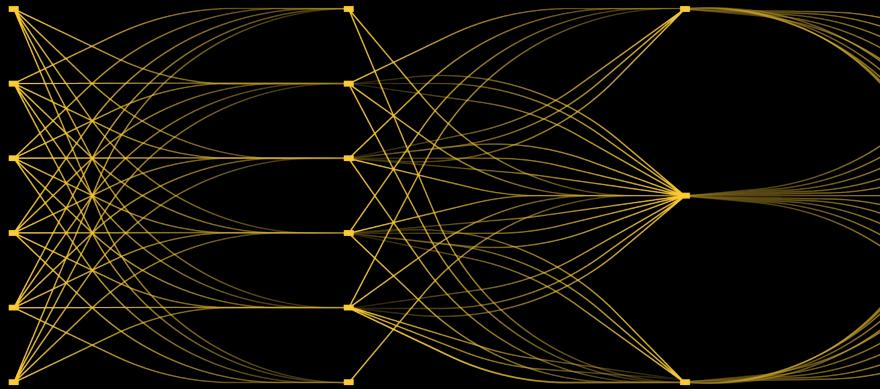


Craiyon



Weights & Biases

Integrations



Fits into your workflow

Integrated into every popular ML framework



Instrumented into over 6,000 popular ML repos



Runs on every cloud or in your own infra



Google Cloud



NVIDIA



Competition Time!

wandb.me/xx



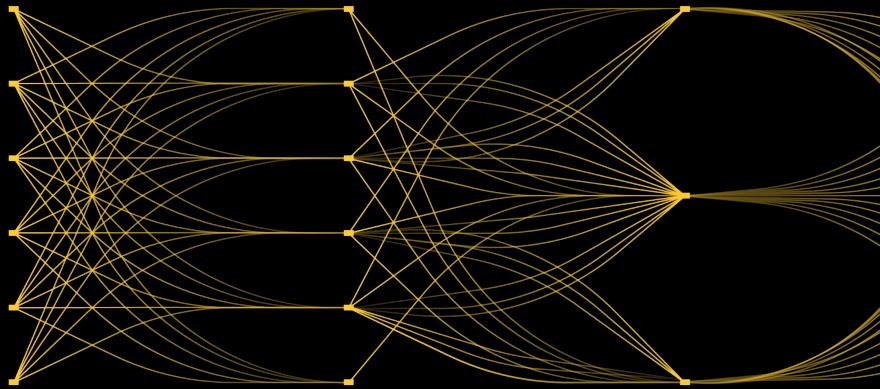
The W&B Course

www.wandb.courses



Weights & Biases

Thank You!





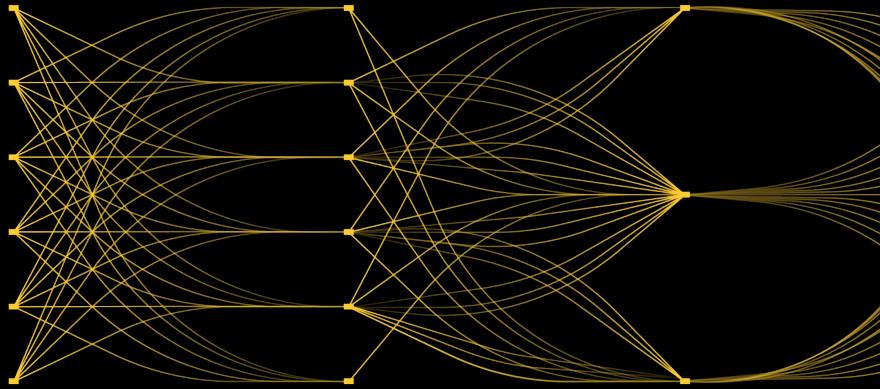
Join us and our ML community!

- Fully Connected – wandb.me/fc
- YouTube – wandb.me/youtube
- Twitter – wandb.me/twitter



Weights & Biases

Appendix



October 2022



Andrew's Presentation

<https://www.youtube.com/watch?v=Se1HvbAM0O4&t=12s>