



# Simplifying the Machine Learning Lifecycle



- **/** Broad Adoption of ML ... and its issues
- / The need for standardization
- / ML development challenges
- / How MLflow tackles these

# Login to Databricks Community Edition

#### https://databricks.com/try

- Sign up for Databricks Community Edition for free
- We will use this for the tutorial
- Once you sign up, you can continue to use it to learn and experiment on a dedicated data sciences engineering environment



## Go to databricks.com/try

databricks	Platform	Solutions	Customers L	Learn F	Partners	Events	Open So	urce Co	ompany		Q EN	SUPPORT	CONTACT	LOG IN	TRY DATABRICKS
Try Databricks															
		An onen :	and unified	data ar		-			ineering, ma	chine lear	ning ar	nd analyt	ice		
		Anopena			-			-	elta Lake, MLflow		-	ia analy c	103		
			Tell us a lif	ttle abou	it yoursel	f to get st	arted.								
			<ul> <li>First Name:</li> </ul>					* Last Name	2:						
			* Company N	lame				* Work Ema							
			Company N	ame				Work Lina							
* How would you describe your role? * What is your intended use case?															
Select Select															
			Phone Numb	er											
			By Clicking "S	Sign Up". vo	ou agree to t	he Privacy F	Policy								
									ache Spark™.						
			La Keep me i	Informed wi	th occasion	ai updates a	ibout Databr	ricks and Apa	ache Spark						
			SIG	SN UP											
														() I	lelp



## Sign up for Community Edition

Solutions Customers Learn Partners Events Open Source Company 😥

Q EN SUPPORT CONTACT LOG IN

TRY DATABRICKS

#### Try Databricks

An open and unified data analytics platform for data engineering, machine learning, and analytics

From the original creators of Apache Spark™, Delta Lake, MLflow, and Koalas

#### Select a platform

#### DATABRICKS PLATFORM - FREE TRIAL

For businesses

#### COMMUNITY EDITION

For students and educational institutions

- Collaborative environment for Data teams to build solutions together
- Unlimited clusters that can scale to any size, processing data in your own account
- Job scheduler to execute jobs for production pipelines
- Fully collaborative notebooks with multi-language support, dashboards, REST APIs
- Native integration with the most popular ML frameworks (scikit-learn, TensorFlow, Keras,...), Apache SparkTM, Delta Lake, and MLflow
- Advanced security, role-based access controls, and audit logs
- Single Sign On support
- Integration with BI tools such as Tableau, Qlik, and Looker
- 14-day full feature trial (excludes cloud charges)

#### GET STARTED ON





Please note that Azure Databricks is provided by Microsoft and is subject to Microsoft's terms.

By clicking on the "AWS" button to get started, you agree to the Databricks Terms of Service.



- · Basic notebooks without collaboration
- Limited to 3 max users
- · Public environment to share your work

#### GET STARTED

By clicking 'Get Started' for the Community Edition, you agree to the Databricks Community Edition Terms of Service.

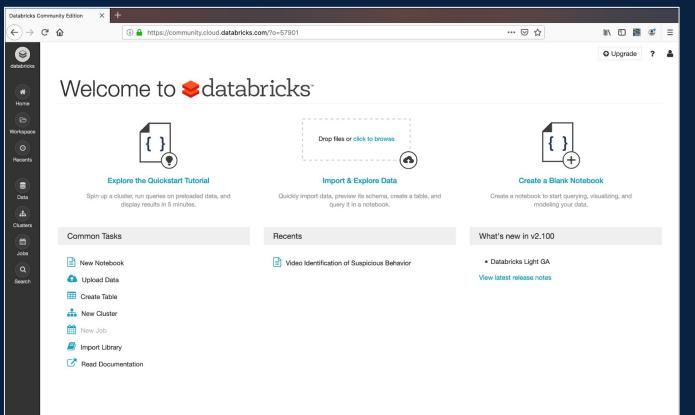


## Sign up for Community Edition

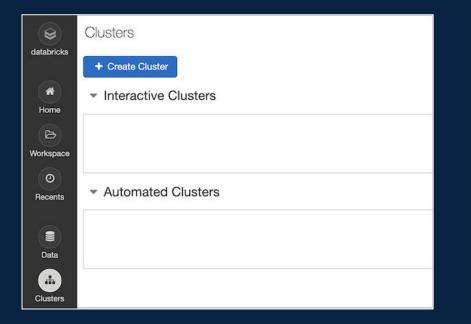
jin - Databricks Community Editic 🗙 📑								
$ ightarrow$ $ ightarrow$ $ m C^{\prime}$ $ m G$	🛈 🔒 https://community.cloud.databricks.com/login.html 🛛 😶 😭	lii\	5		5	≡		
portant Notice: Acceptable use and unused account termination policy and Terms of Use update.								
	edatabricks <sup>•</sup>							
	Sign In to Databricks							
	denny.g.lee@gmail.com							
	Forgot Password?							
	Sign In New to Databricks? Sign Up.							
	Privacy Policy   Terms of Use							



# Log into DBCE









	Create Cluster
databricks	New Cluster Cancel Create Cluster 0 Workers: 0.0 GB Memory, 0 Cores, 0 DBU 1 Driver: 6.0 GB Memory, 0.88 Cores, 1 DBU 0
Home	Cluster Name
Ъ	
Workspace	Databricks Runtime Version @
0	Runtime: 5.5 LTS (Scala 2.11, Spark 2.4.3)
Recents	Python Version @
(	3
	Instance
Data	Free 6GB Memory: As a Community Edition user, your cluster will automatically terminate after an idle period of two hours. For more configuration options, please upgrade your Databricks subscription.
Clusters	Instances Spark
Jobs	Availability Zone @
Q	us-west-2c \$
Search	



databricks	Create Cluster			
	New Cluster	Cancel Create Clu		orkers: 0.0 GB Memory, 0 Cores, 0 DBU iver: 6.0 GB Memory, 0.88 Cores, 1 DBU 🚱
Home	Cluster Name			
Ъ	delta-rocks		83	
Vorkspace	Databricks Runtime Vers	ion 😧		
0	Runtime: 6.1 Beta (Scal	a 2.11, Spark 2.4.4)	~	
Recents	Databricks Runtime			
	6.1 Beta	Scala 2.11, Sp	oark 2.4.4	
	6.1 ML Beta	GPU, Scala 2.11, Sp	oark 2.4.4	
	6.1 ML Beta	Scala 2.11, Sp	oark 2.4.4	utomatically terminate after an idle period of two hours.
Data	20 more			
(A)	Instances Spark			
Clusters				
	Availability Zone 🚱			
Jobs	us-west-2c		\$	
Q				
Search				



8	Clusters								
databricks	+ Create Cluster							All	Created by me
A Home	<ul> <li>Interactive Clusters</li> </ul>								
Ъ	Name	State	Nodes	Driver	Worker	Runtime	Creator		
Workspace	delta-rocks	Running	1 (0 spot)	Community	. Community	. 6.1 Beta (inclu	denny.g.lee@.	0	
O     Recents	<ul> <li>Automated Clusters</li> </ul>								
<b>))</b> Data					No cluster	's found			
Clusters									



Databricks Co	mmunity Edition X	01-Delta Lake Workshop - Delta L	ak × +					
$\leftrightarrow$ $\rightarrow$ (	C 🕜	i 🔒 https://communit	ty.cloud.databricks.cor	n/?o=57901	☺ ☆	III\ 🗊 🇱	۲	≡
databricks	Workspace		Ŧ			O Upgrade	?	4
ualabilitika	workspace	✓ Delta Lake Lab	× 1	• 1				
<b>*</b>	🗅 archive		Create	cks				
Home	🗅 Delta Lake Lab		Clone Rename					
B			Move					
Workspace			Delete	Drop files or click to browse	[ ]			
Ø Recents			Import					
			Export Permissions			0		
			<u> </u>	Import & Explore Data	Create a Blank Noteboo			
Data			nd	Quickly import data, preview its schema, create a table, and query it in a notebook.	Create a notebook to start querying, visu modeling your data.	anzing, and		
Clusters								
				Recents	What's new in v2.100			
Jobs	<			-				
( <b>Q</b> )			1	Video Identification of Suspicious Behavior	<ul> <li>Databricks Light GA</li> </ul>			
Search					View latest release notes			
https://commu	nity.cloud.databricks.com/?	o=57901#						



Databricks (	community Edition X	01-Delta Lake Workshop - Delta Lak 🗙	+			
	C' û	🛈 🔒 https://community.cle	ud.databricks.com/?o=57901	🛛 ☆	III\ 🗉 🦉 📽	≡
databricks	Workspace	✓ Delta Lake Lab	±		O Upgrade ?	4
	C archive		Import Notebooks			
Home Developments Workspace O Recents Data Clusters Jobs O Search	C Delta Lake Lab		Import rom: File • URL Accepted formats: .dbc, .scala, .py, .sql, .r, .ipynb, .Rmd, .html (To import a library, such as a jar or egg, click here) Cancel import query rt in a nonebook. Recents Video Identification of Suspicious Behavior	Create a Blank Noteboo Create a Blank Noteboo Create a notebook to start querying, vi modeling your data What's new in v2.100 • Databricks Light GA View latest release notes		



Databricks Co	ommunity Edition X	01-Delt	a Lake Workshop - Delta La	× +						
	C û	G	🔒 https://community	.cloud.dat	tabricks.com/?o=57901	⊠ ☆		III\ 🗊 🎽	۲	≡
databricks	Workspace	~	Delta Lake Lab	±	1 . 1			O Upgrade	?	4
Home Dorkapace O Recents Data Clusters Clusters Search	C archive			Impo	port Notebooks or from: File URL a01-Delta%20Lake%20Workshop%20-%20Delta%20Lake%20Primer.html Accepted formats: .dbo, .scala, .py, .sql, .r, .jpynb, .Rmd, .html (To import a library, such as a jar or egg, .click here) Cancel Import Query II III a NOTEBOOK Pecents Video Identification of Suspicious Behavior	Create a notebook to	eling your data.			



Attach: • delta-rocks 6.00 GB   0.88 Cores   DBR 6.1 Beta   Spark 2.4.4   Scala 2.11 An open-source storage format that brings ACID transactions to Apache Spark™ and big data workloads. • Open format: Stored as Parquet format in blob storage. • ACID Transactions: Ensures data integrity and read consistency with complex, concurrent data pipelines. • Schema Enforcement and Evolution: Ensures data cleanliness by blocking writes with unexpected. • Audit History: History of all the operations that happened in the table. • Time Travel: Query previous versions of the table by time or version number. • Deletes and upserts: Supports deleting and upserting into tables with programmatic APIs. • Scalable Metadata management: Able to handle millions of files are scaling the metadata operations with Spark.		• •			C	2	
An open-source storage format that brings ACID transactions to Apache Spark <sup>™</sup> and big data workloads. • Open format: Stored as Parquet format in blob storage. • ACID Transactions: Ensures data integrity and read consistency with complex, concurrent data pipelines. • Schema Enforcement and Evolution: Ensures data cleanliness by blocking writes with unexpected. • Audit History: History of all the operations that happened in the table. • Time Travel: Query previous versions of the table by time or version number. • Deletes and upserts: Supports deleting and upserting into tables with programmatic APIs. • Scalable Metadata management: Able to handle millions of files are scaling the metadata operations with Spark.	Attach:						
<ul> <li>An open-source storage format that brings ACID transactions to Apache Spark<sup>™</sup> and big data workloads.</li> <li>Open format: Stored as Parquet format in blob storage.</li> <li>ACID Transactions: Ensures data integrity and read consistency with complex, concurrent data pipelines.</li> <li>Schema Enforcement and Evolution: Ensures data cleanliness by blocking writes with unexpected.</li> <li>Audit History: History of all the operations that happened in the table.</li> <li>Time Travel: Query previous versions of the table by time or version number.</li> <li>Deletes and upserts: Supports deleting and upserting into tables with programmatic APIs.</li> <li>Scalable Metadata management: Able to handle millions of files are scaling the metadata operations with Spark.</li> </ul>		C				111	× -
<ul> <li>Open format: Stored as Parquet format in blob storage.</li> <li>ACID Transactions: Ensures data integrity and read consistency with complex, concurrent data pipelines.</li> <li>Schema Enforcement and Evolution: Ensures data cleanliness by blocking writes with unexpected.</li> <li>Audit History: History of all the operations that happened in the table.</li> <li>Time Travel: Query previous versions of the table by time or version number.</li> <li>Deletes and upserts: Supports deleting and upserting into tables with programmatic APIs.</li> <li>Scalable Metadata management: Able to handle millions of files are scaling the metadata operations with Spark.</li> </ul>							
<ul> <li>Open format: Stored as Parquet format in blob storage.</li> <li>ACID Transactions: Ensures data integrity and read consistency with complex, concurrent data pipelines.</li> <li>Schema Enforcement and Evolution: Ensures data cleanliness by blocking writes with unexpected.</li> <li>Audit History: History of all the operations that happened in the table.</li> <li>Time Travel: Query previous versions of the table by time or version number.</li> <li>Deletes and upserts: Supports deleting and upserting into tables with programmatic APIs.</li> <li>Scalable Metadata management: Able to handle millions of files are scaling the metadata operations with Spark.</li> </ul>	$\sim$						
<ul> <li>ACID Transactions: Ensures data integrity and read consistency with complex, concurrent data pipelines.</li> <li>Schema Enforcement and Evolution: Ensures data cleanliness by blocking writes with unexpected.</li> <li>Audit History: History of all the operations that happened in the table.</li> <li>Time Travel: Query previous versions of the table by time or version number.</li> <li>Deletes and upserts: Supports deleting and upserting into tables with programmatic APIs.</li> <li>Scalable Metadata management: Able to handle millions of files are scaling the metadata operations with Spark.</li> </ul>	An open-source storage format that brings ACID	transactions to Apach	ne Spark™ and big data workloads.				
<ul> <li>Schema Enforcement and Evolution: Ensures data cleanliness by blocking writes with unexpected.</li> <li>Audit History: History of all the operations that happened in the table.</li> <li>Time Travel: Query previous versions of the table by time or version number.</li> <li>Deletes and upserts: Supports deleting and upserting into tables with programmatic APIs.</li> <li>Scalable Metadata management: Able to handle millions of files are scaling the metadata operations with Spark.</li> </ul>	Open format: Stored as Parquet format in blo	lob storage.					
<ul> <li>Audit History: History of all the operations that happened in the table.</li> <li>Time Travel: Query previous versions of the table by time or version number.</li> <li>Deletes and upserts: Supports deleting and upserting into tables with programmatic APIs.</li> <li>Scalable Metadata management: Able to handle millions of files are scaling the metadata operations with Spark.</li> </ul>	ACID Transactions: Ensures data integrity ar	nd read consistency w	ith complex, concurrent data pipelines				
<ul> <li>Time Travel: Query previous versions of the table by time or version number.</li> <li>Deletes and upserts: Supports deleting and upserting into tables with programmatic APIs.</li> <li>Scalable Metadata management: Able to handle millions of files are scaling the metadata operations with Spark.</li> </ul>	Schema Enforcement and Evolution: Ensur	res data cleanliness by	blocking writes with unexpected.				
<ul> <li>Deletes and upserts: Supports deleting and upserting into tables with programmatic APIs.</li> <li>Scalable Metadata management: Able to handle millions of files are scaling the metadata operations with Spark.</li> </ul>	Audit History: History of all the operations th	nat happened in the tak	ble.				
• Scalable Metadata management: Able to handle millions of files are scaling the metadata operations with Spark.	• Time Travel: Query previous versions of the t	table by time or version	n number.				
	• Deletes and upserts: Supports deleting and	upserting into tables v	with programmatic APIs.				
	Scalable Metadata management: Able to ha	andle millions of files a	re scaling the metadata operations wit	h Spark.			
Unified Batch and Streaming Source and Sink: A table in Delta Lake is both a batch table, as well as a streaming source and sink	Unified Batch and Streaming Source and S	Sink: A table in Delta L	ake is both a batch table, as well as a	streaming s	ource	and sir	٦k.
Streaming data ingest, batch historic backfill, and interactive queries all just work out of the box.	Streaming data ingest, batch historic backfill,	, and interactive querie	s all just work out of the box.				
	Cmd 2						
and a							
Cmd 2							
	SPARK+AI						
	<b>V</b> SUMMIT EURO	)PE					
	15 - 17 OCTOBER 2019   AMSTERDAM   ORGANIZED BY	latabricks					
SPARK+AI SUMMIT EUROPE		latabricks					



## Broad Adoption of ML

Huge disruptive innovations are affecting most enterprises on the planet



Healthcare and Genomics

Fraud Prevention

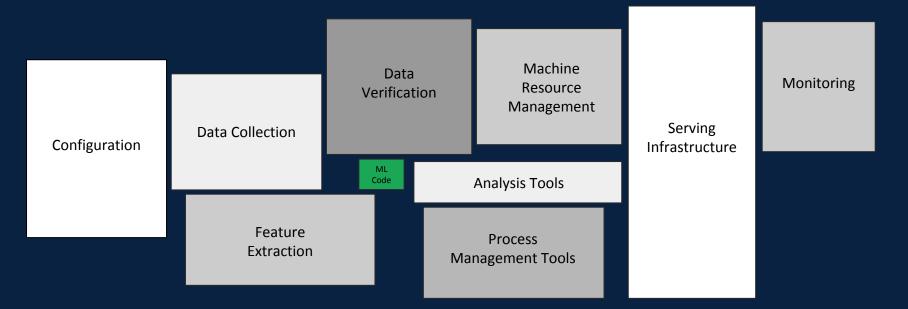
Digital Personalization

Internet of Things

and many many more customers in different industries and segments

## Hardest Part of ML isn't ML, it's Data

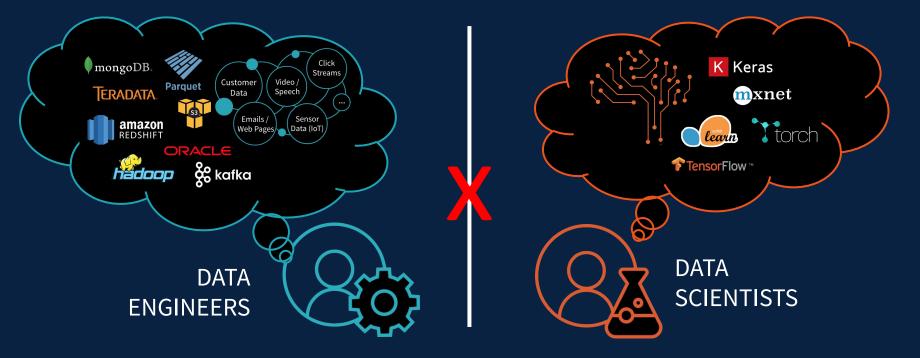
*"Hidden Technical Debt in Machine Learning Systems," Google NIPS 2015* 



Only a small fraction of real-world ML systems is composed of the ML code, as shown by the small green box in the middle. The required surrounding infrastructure is vast and complex.



## Data & ML Tech and People are in Silos





# ML Lifecycle is Manual, Inconsistent and Disconnected

#### **Prep Data**

- Low level integrations for  $\bullet$ Data and ML
- Difficult to track data used  $\bullet$ for a model





#### **Build Model**

- Ad hoc approach to track • experiments
- Very hard to reproduce  $\bullet$ experiments

GitHub CONDA



TensorFlow 1F

#### **Deploy Model**

- Multiple tightly coupled  $\bullet$ deployment options
- Different monitoring approach for each framework



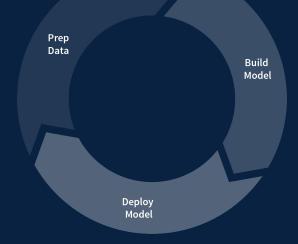
Amazon SageMaker

Azure Machine Learning





# The need for standardization





```
Elasticnet model (alpha=0.01, l1_ratio=1.0):

RMSE: ??

MAE: 51.051828604086325

R2: 0.3951809598912357
```

```
Elasticnet model (alpha=?, l1_ratio=0.75):

RMSE: 65.28994906390733

MAE: 53.759148284349266

R2: ??
```

```
Elasticnet model (alpha=0.01, l1_ratio=?):

RMSE: 71.40362571026475

MAE: ??

R2: 0.2291130640003659
```



```
Elasticnet model (alpha=0.01, l1_ratio=1.0):

RMSE: ??

MAE: 51.051828604086325

R2: 0.3951809598912357
```

```
Elasticnet model (alpha=?, l1_ratio=0.75):

RMSE: 65.28994906390733

MAE: 53.759148284349266

R2: ??
```

```
Elasticnet model (alpha=0.01, l1_ratio=?):

RMSE: 71.40362571026475

MAE: ??

R2: 0.2291130640003659
```



Did anything change in the feature engineering?



```
Elasticnet model (alpha=0.01, l1_ratio=1.0):

RMSE: ??

MAE: 51.051828604086325

R2: 0.3951809598912357
```

```
Elasticnet model (alpha=?, l1_ratio=0.75):

RMSE: 65.28994906390733

MAE: 53.759148284349266

R2: ??
```

```
Elasticnet model (alpha=0.01, l1_ratio=?):

RMSE: 71.40362571026475

MAE: ??

R2: 0.2291130640003659
```

How did the hyperparameters change?



```
Elasticnet model (alpha=0.01, l1_ratio=1.0):

RMSE: ??

MAE: 51.051828604086325

R2: 0.3951809598912357
```

```
Elasticnet model (alpha=?, l1_ratio=0.75):

RMSE: 65.28994906390733

MAE: 53.759148284349266

R2: ??
```

```
Elasticnet model (alpha=0.01, l1_ratio=?):

RMSE: 71.40362571026475

MAE: ??

R2: 0.2291130640003659
```

What data was this model trained on?



```
Elasticnet model (alpha=0.01, l1_ratio=1.0):

RMSE: ??

MAE: 51.051828604086325

R2: 0.3951809598912357
```

```
Elasticnet model (alpha=?, l1_ratio=0.75):

RMSE: 65.28994906390733

MAE: 53.759148284349266

R2: ??
```

```
Elasticnet model (alpha=0.01, l1_ratio=?):

RMSE: 71.40362571026475

MAE: ??

R2: 0.2291130640003659
```

How did the offline metrics change?



```
Elasticnet model (alpha=0.01, l1_ratio=1.0):

RMSE: ??

MAE: 51.051828604086325

R2: 0.3951809598912357
```

```
Elasticnet model (alpha=?, l1_ratio=0.75):

RMSE: 65.28994906390733

MAE: 53.759148284349266

R2: ??
```

```
Elasticnet model (alpha=0.01, l1_ratio=?):

RMSE: 71.40362571026475

MAE: ??

R2: 0.2291130640003659
```



#### What else am I missing?



The difference between releasing Software and deploying ML Models







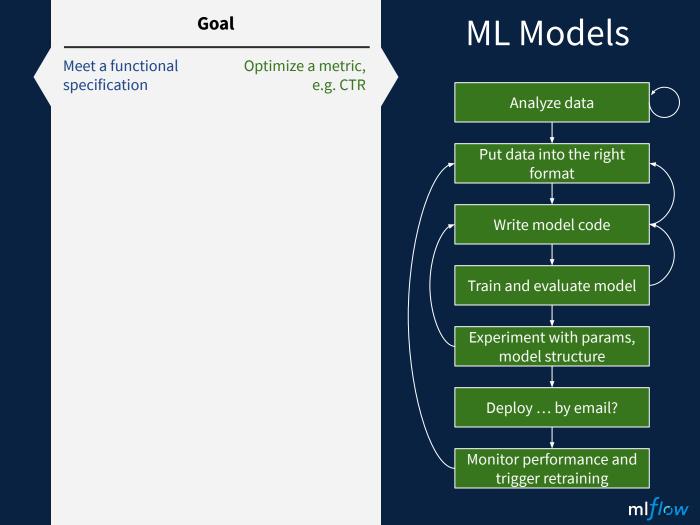
#### ML Models



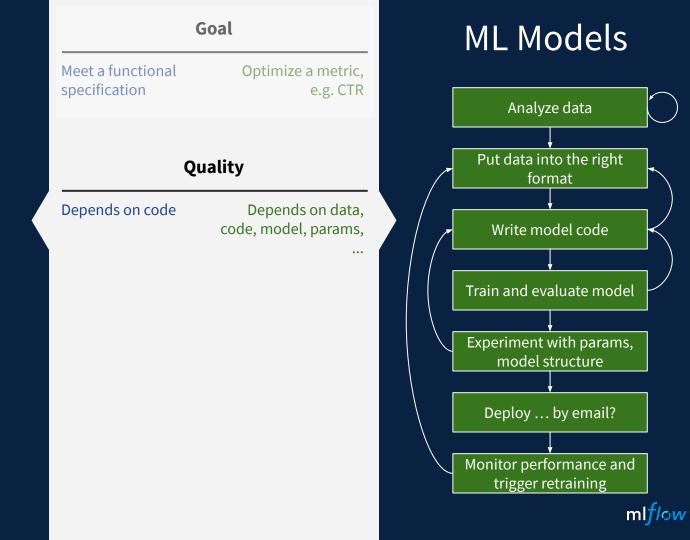








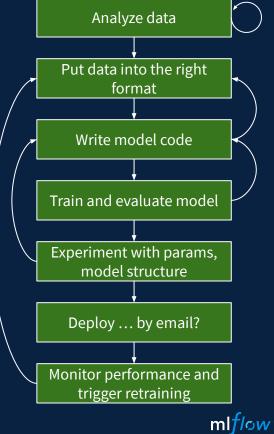




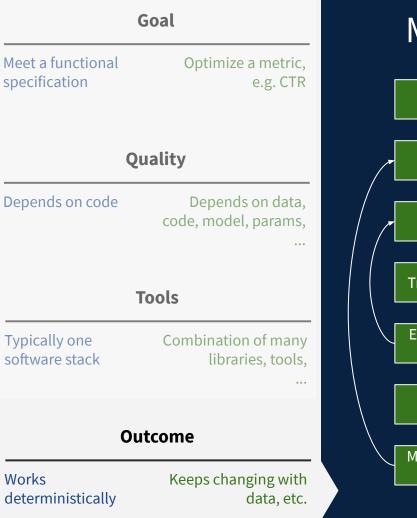


	Goal	MLN
Meet a functional specification	Optimize a metric, e.g. CTR	Ana
Q	uality	Put data
Depends on code	Depends on data, code, model, params, 	Write
-	rools	Train and
Typically one software stack	Combination of many libraries, tools,	Experime mod
		Deploy
		Monitor p trigge

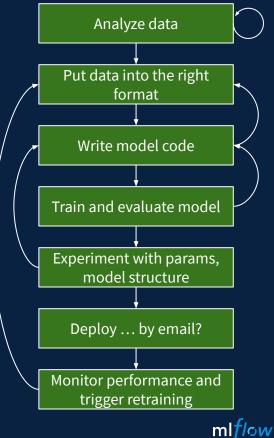
### L Models





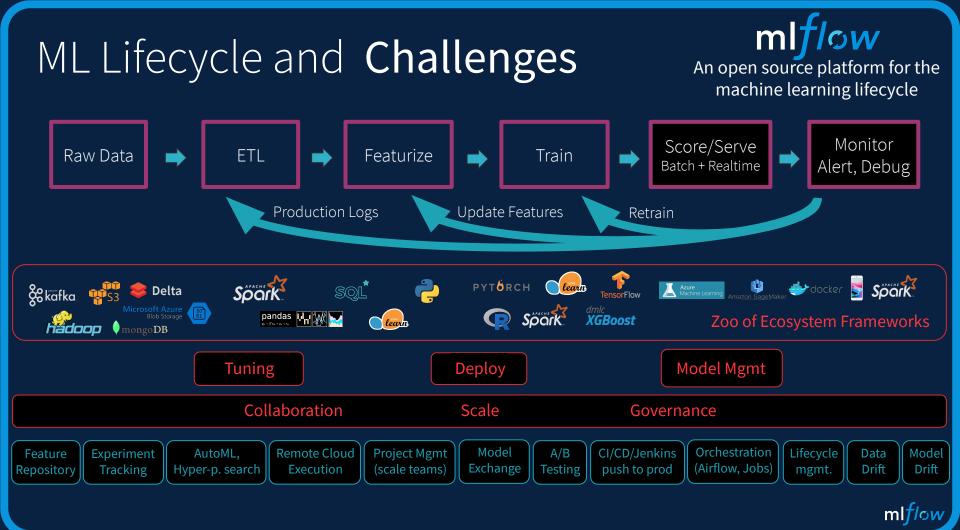


#### ML Models



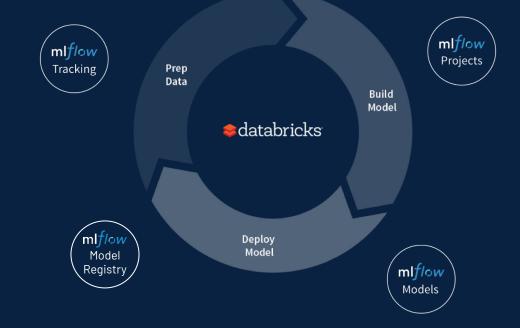


## In summary, deploying ML Models is hard!



# Introducing MLflow

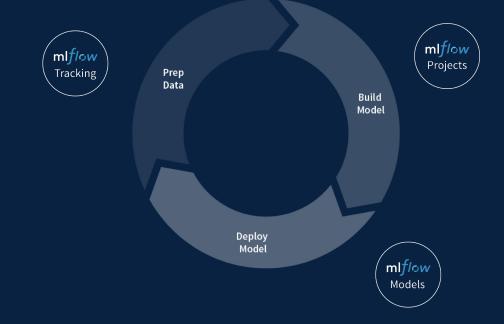
Unveiled in June 2018, MLflow is the only open source framework designed to manage the complete Machine Learning Lifecycle.





## Introducing MLflow

Unveiled in June 2018, MLflow is the only open source framework designed to manage the complete Machine Learning Lifecycle.







Months since Project Launch



# mlflow Components

#### ml**flow** Tracking

Record and query experiments: code, data, config, results

#### mlflow Projects

Packaging format for reproducible runs on any platform

#### ml*flow* Models

General format that standardizes deployment paths ml**flow** Model Registry

Centralized and collaborative model lifecycle management

mlflow.org







# mlflow Components

#### ml**flow** Tracking

Record and query experiments: code, data, config, results

#### mlflow Projects

Packaging format for reproducible runs on any platform

#### ml*flow* Models

General format that standardizes deployment paths men **mlflow** Model Registry

> Centralized and collaborative model lifecycle management

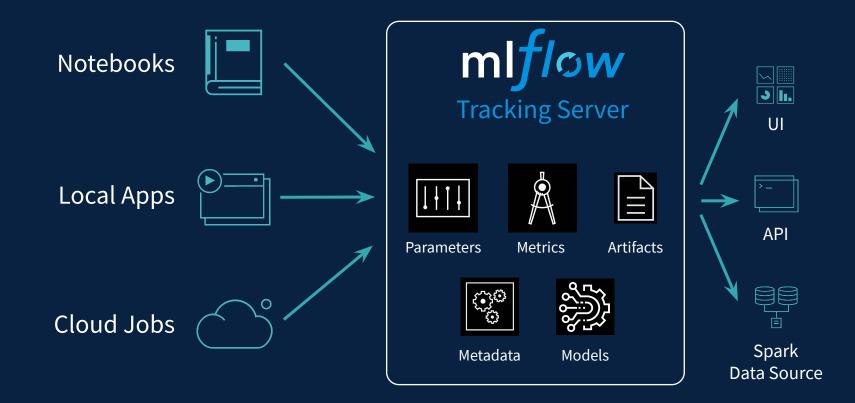
mlflow.org







# mlflow Tracking



mlflow

## Key Concepts in Tracking

Parameters: key-value inputs to your code Metrics: numeric values (can update over time) Artifacts: arbitrary files, including models Source: what code ran?

ml <i>flow</i>				GitHub Docs			
Experiments	Default						
Default	Experiment ID: 0 Artifact Location: /Users/matei/mlflow/mlruns/0						
Something	Search Runs: metrics.rmse <	Search Runs: metrics.mse < 1 and params.model = "tree"					
	Filter Params: alpha, Ir	Filter Metrics:	mse, r2	Clear			
	4 matching runs Compare Sel	Download CSV 🛓					
	Date	User Source	F Version (n/a	Parameters Metrics			
	2018-06-28 17:09:49	matei matei_test.py	7cff8e	2.123			
	2018-06-28 17:09:06	matei matei_test.py	7cff8e	4.543			
	2018-06-28 17:09:05	matei matei_test.py	7cff8e	4.543			
	2018-06-25 13:08:12	matei matei_test.py	53ccdc	4.543			

```
# Scikit Learn Linear Regression via ElasticNet
lr = ElasticNet(alpha=alpha, l1_ratio=l1_ratio, random_state=42)
lr.fit(train_x, train_y)
```

# Predict
predicted\_qualities = lr.predict(test\_x)

# Evaluate Metrics
(rmse, mae, r2) = eval\_metrics(test\_y, predicted\_qualities)



with mlflow.start\_run() as run:

# Scikit Learn Linear Regression via ElasticNet lr = ElasticNet(alpha=alpha, l1\_ratio=l1\_ratio, random\_state=42) lr.fit(train\_x, train\_y)

# Predict
predicted\_qualities = lr.predict(test\_x)

```
# Evaluate Metrics
(rmse, mae, r2) = eval_metrics(test_y, predicted_qualities)
```

# Log
mlflow.log\_param("alpha", alpha)

• • •

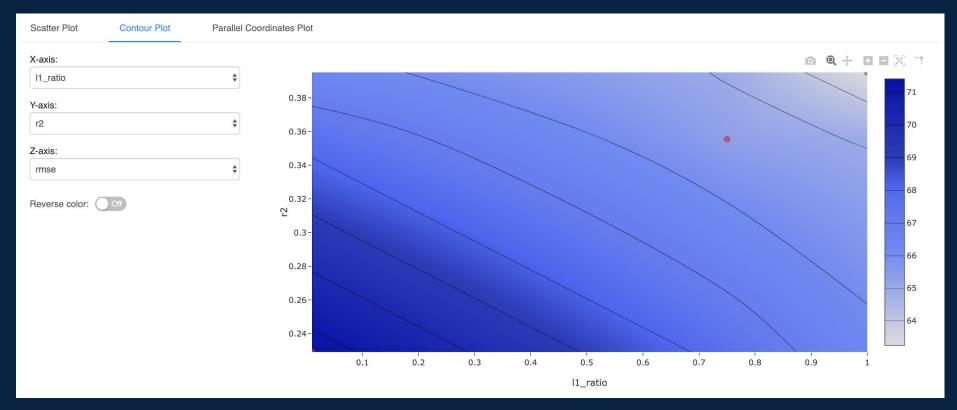


### GitHub Demo

https://github.com/dennyglee/mlflow-diabetes-example

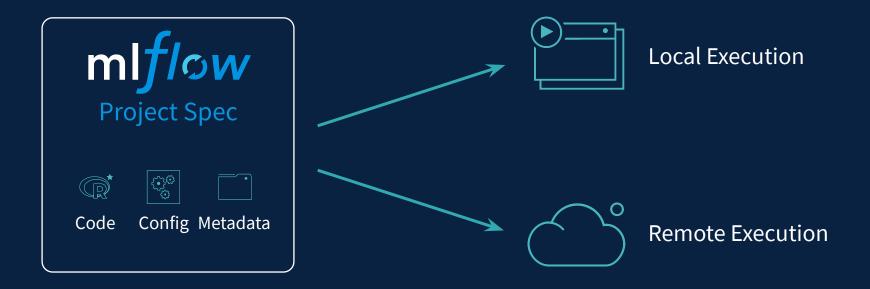


#### Comparing Runs Contour Plot







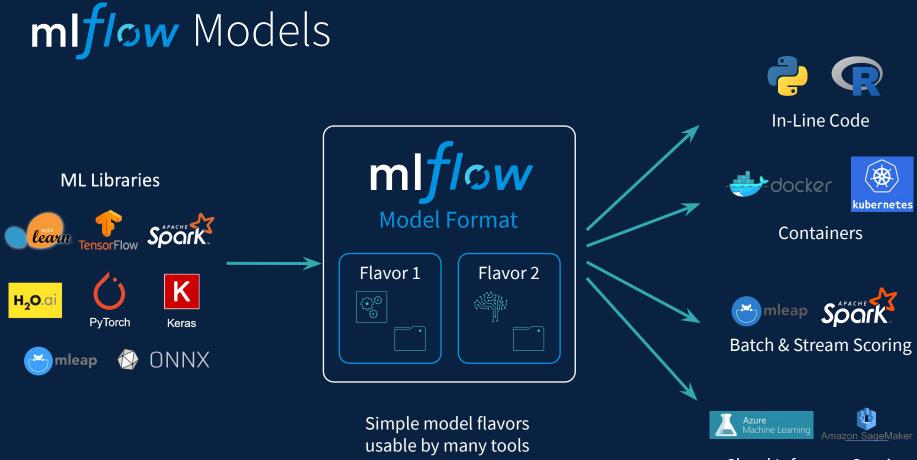




## Example MLflow Project

my\_project/ MLproject conda env: conda.yaml entry points: main: parameters: training data: path lambda: {type: float, default: 0.1} command: python main.py {training data} {lambda} conda.yaml main.py \$ mlflow run git://<my\_project> model.py mlflow.run("git://<my project>", ...)

mlflow



Cloud Inference Services

### Example MLflow Model

my\_model/ |---- MLmodel

run\_id: 769915006efd4c4bbd662461
time\_created: 2018-06-28T12:34
flavors:

tensorflow: saved\_model\_dir: estimator signature\_def\_key: predict python\_function:

loader\_module: mlflow.tensorflow

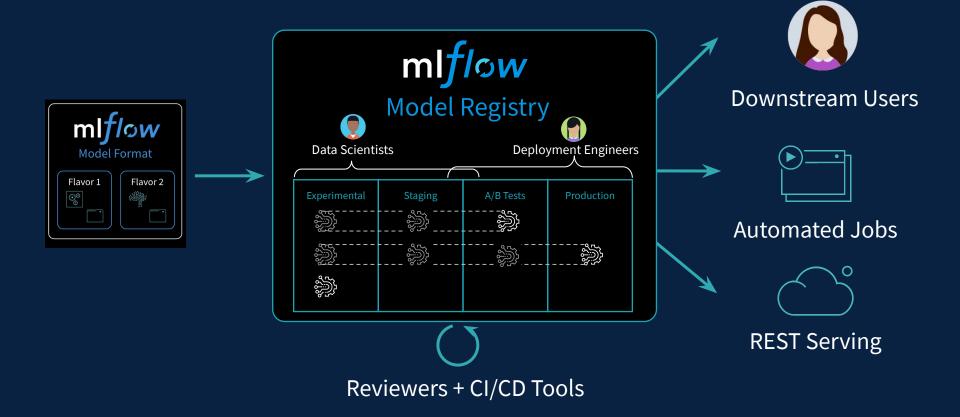
Usable by tools that understand TensorFlow model format Usable by any tool that can run Python (Docker, Spark, etc!)

estimator/

--- saved\_model.pb
--- variables/

• • •

# **mlflow** Model Registry



Registered Models			3	
negistered models			search model name Q	
Name 🔶	Latest Version	Staging	Production	Last Modified 👙
AaronModel	Version 108	_		2019-10-18 09:04:20
Airline_Delay_Scikit	Version 3	-	Version 1	2019-10-11 12:41:43
Airline_Delay_SparkML	Version 11	Version 6	Version 5	2019-10-22 10:30:21
Andre_BasicModels_02_Sklearn_Train_Predict	Version 2	Version 1	_	2019-10-19 14:38:11
BertisLarge	Version 1	-		2019-10-11 15:18:05
Brooke Keras Model	Version 1	-	-	2019-10-12 08:20:12
holland-forecast-model	Version 1	-	Version 1	2019-10-07 15:38:27
joytesting	20	_		2019-10-15 18:23:17
ManiErrorModel		_	-	2019-10-14 16:53:10
MatelModel	Version 5	Version 5	Version 3	9-10-10 14:07:07
				< 1 2 3

#### One Collaborative Hub

- Central Model Repository
- 2 Overview of versions in Staging/Production/etc.
- **3** Search/filter/pagination



Registered Models > Airline_Delay_SparkML -	Stage: Staging V
Created Time: 2019-10-10 15:20:29 Last Modified: 2019-10-22 17:08:29	2 Request transition to → None
Description      Predicts airline delays (in minutes) using the best Spark RF model from the AutoML Toolkit.	Request transition to → Production
1 Versions All Active(2)	Request transition to → Archived
Version Registered at Created by Stage Pending Requests	Transition to → None
⊘ Version 5 2019-10-11 12:44:44 clemens.mewald@databricks.com Production -	Transition to - Production
⊘ Version 6 2019-10-16 03:15:56 clemens.mewald@databricks.com Staging 1	Transition to → Archived
< 1 >	

#### Management of the entire ML Lifecycle (MLOps)



2 Request/Approval workflow for transitioning deployment stages



#### Registered Models > Airline\_Delay\_SparkML > Version 5 -

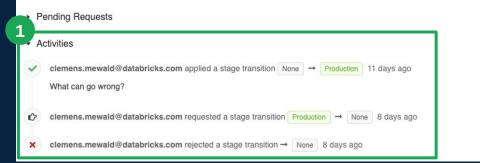
Registered At: 2019-10-11 12:44:44

Creator: clemens.mewald@databricks.com Stage: Production v

Last Modified : 2019-10-22 09:03:28 Source Run : Run 6151fe768a5e49d39076b07448e60d57

Description

Improved the Airline delay model using a GBDT. See run for improved metrics.



#### Visibility

 Full activity log of stage transition requests, approvals, etc.



Oct 11 2019, 10:09 AM PDT clemens.mewald@databricks.com

Oct 11 2019, 9:39 AM PDT

Oct 10 2019, 23:00 PM PDT

clemens.mewald@databricks.com

clemens.mewald@databricks.com

#### Registered Models > Airline\_Delay\_SparkML > Version 5 -Registered At: 2019-10-11 12:44:44 Creator: clemens.mewald@databricks.com Stage: Production ~ Last Modified: 2019-10-22 09:03:28 Source Run: Run 6151fe768a5e49d39076b07448e60d57 /Users/clemens.mewald@databricks.com/Airline Demo/... Run 6151fe768a5e49d39076b07448e60d57 1.b Date: 2019-10-11 12:20:44 Source: 02.2 Model Search User: clemens.mewald@databricks.com Duration: 12.8 02.2 Model Search (Python) O Git: Not linked You are viewing a notebook revision from Oct 11 2019, 10:09 AM PDT. Exit Oct 11 2019, 10:09 AM PDT clemens.mewald@databricks.com All changes saved Save now **Define training function**

Cmd 4

def train(params):
 if params['type'] == 'spark\_rf':
 regressor = RandomForestRegressor(featuresCol="features",
 labelCol="ArrPoleLay", maxBins=348, seed=42.

#### Governance and Auditability

1.a

1.b

- Full provenance from Model marked production in the Registry to ...
  - Run that produced the model Notebook that produced the run
  - **Exact revision history of the** notebook that produced the run

ml7

### Notebook Demo

https://github.com/dennyglee/tech-talks/blob/master/sa mples/MLflow%20Diabetes%20Example%20(with%20MLfl ow%20Registry).ipynb



# **mlflow**: An Open Source ML Platform

### Towards more principled Data Science and ML

mlflow.org







### Hands-on Workshop

bit.ly/mlflow-boss-2020

