

Hands on Introduction to IBM's Watson Studio



Power of data. Simplicity of design. Speed of innovation.

Bernie Beekman
Executive I/T Architect

Watson Studio is the new name for the IBM Data Science Experience on Cloud

Watson Knowledge Catalog is the new name for the IBM Data Catalog

Get started with Watson Studio at datascience.ibm.com

Agenda

Time	Description
9:00 AM - 10:00 AM	Overview of Data Science and the Watson Studio Platform Lab Orientation
10:00 AM - 11:00 AM	Lab 1 - Watson Machine Learning
11:00 AM - 12:00 PM	Lab 2 – Watson Studio SPSS Modeler
12:00 PM – 12:30 PM	Lunch
12:30 PM - 1:30 PM	Lab 3 - Machine Learning with SparkML and Jupyter Notebooks
1:30 PM - 2:00 PM	Lab 4 - Data Refinery
2:00 PM – 3:00 PM	Lab-5 - Build, Train, Deploy a Neural Network Model
3:00 PM – 3:30 PM	Questions and Wrap-Up

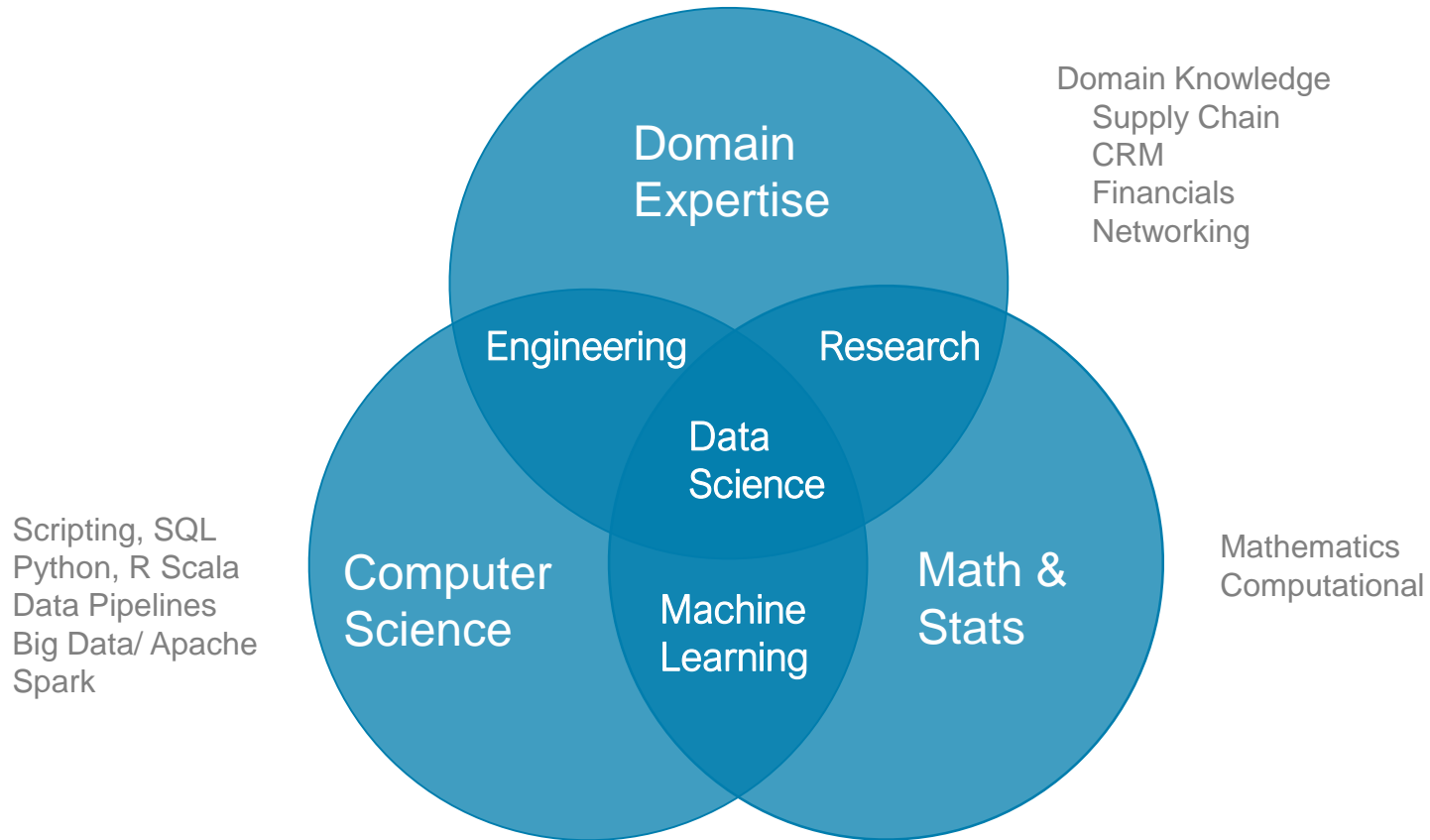
Participant Background

- R/Python/Scala
- Jupyter Notebook
- Spark
- IBM Cloud/Bluemix
- Machine Learning
- Deep Learning/Neural Networks
- Github

Outline

- **Data Science Introduction**
- **Watson Studio Overview**
- **Lab Overview**

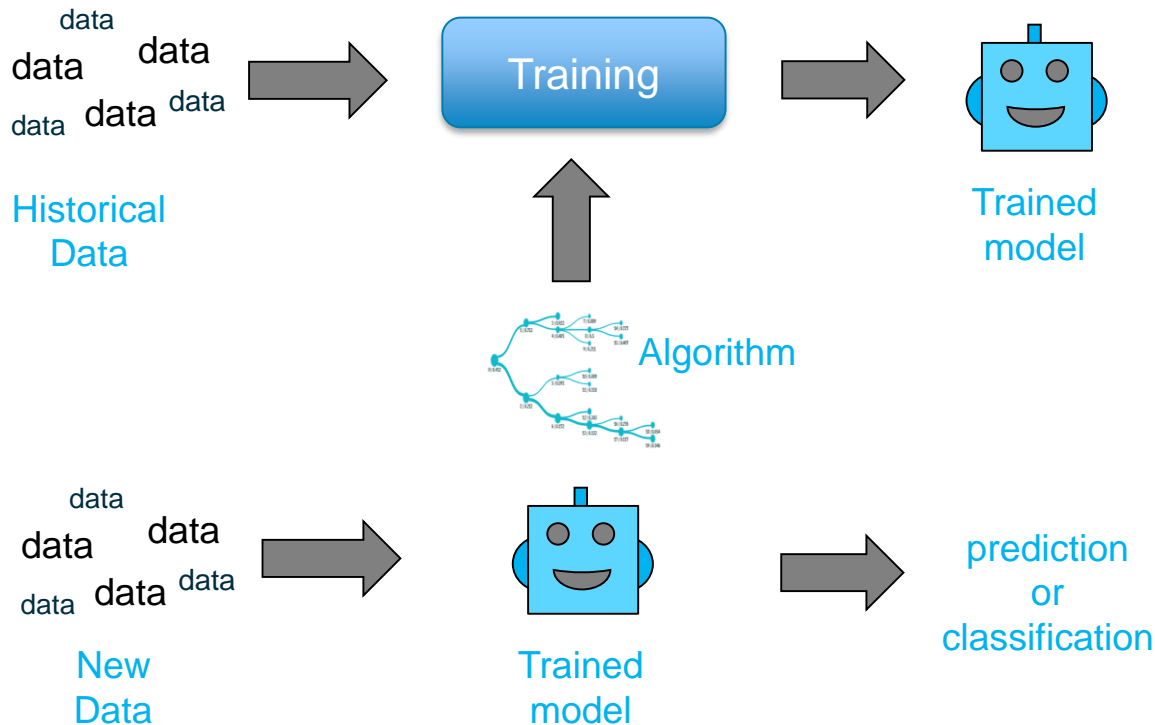
What is Data Science?



Data Science Projects Require Multiple Skills

But what is Machine Learning?

*“Computers that learn without being **explicitly programmed**”*



1

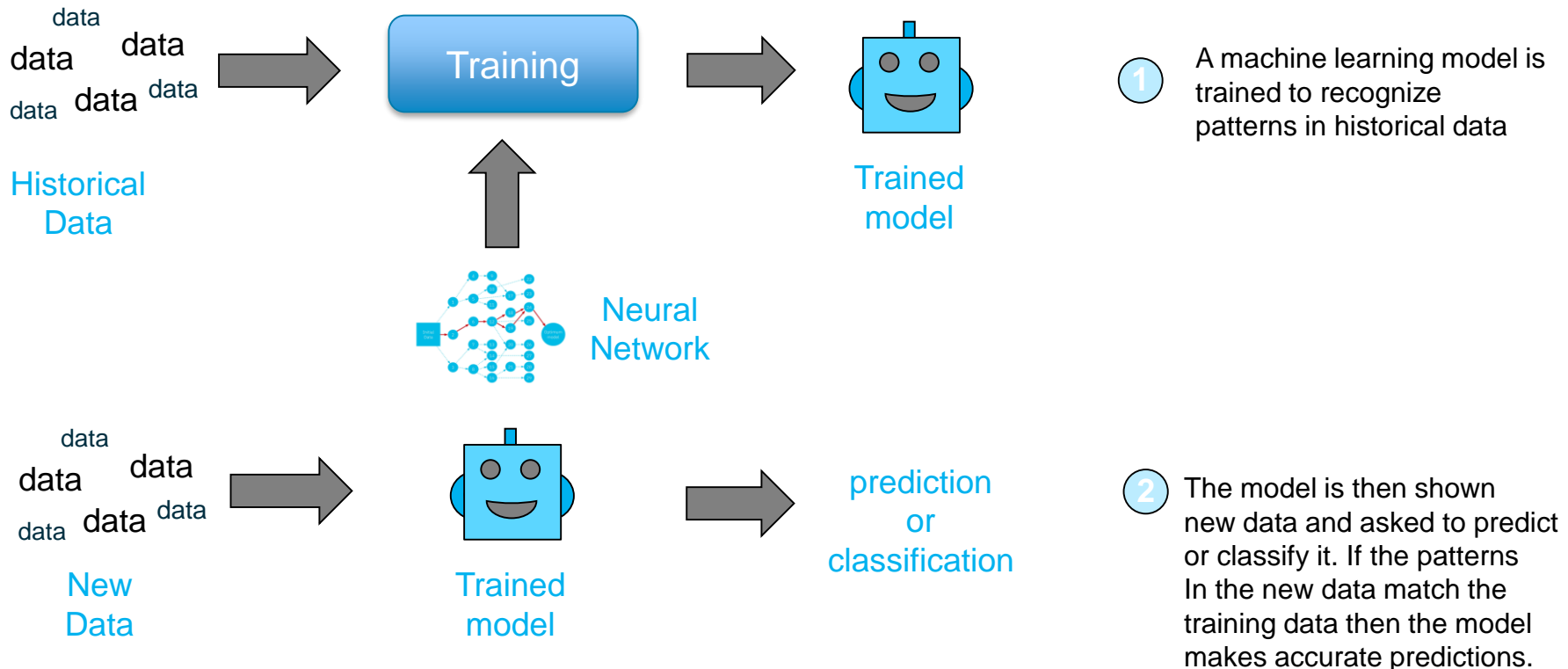
A machine learning model is trained to recognize patterns in historical data

2

The model is then shown new data and asked to predict or classify it. If the patterns in the new data match the training data then the model makes accurate predictions.

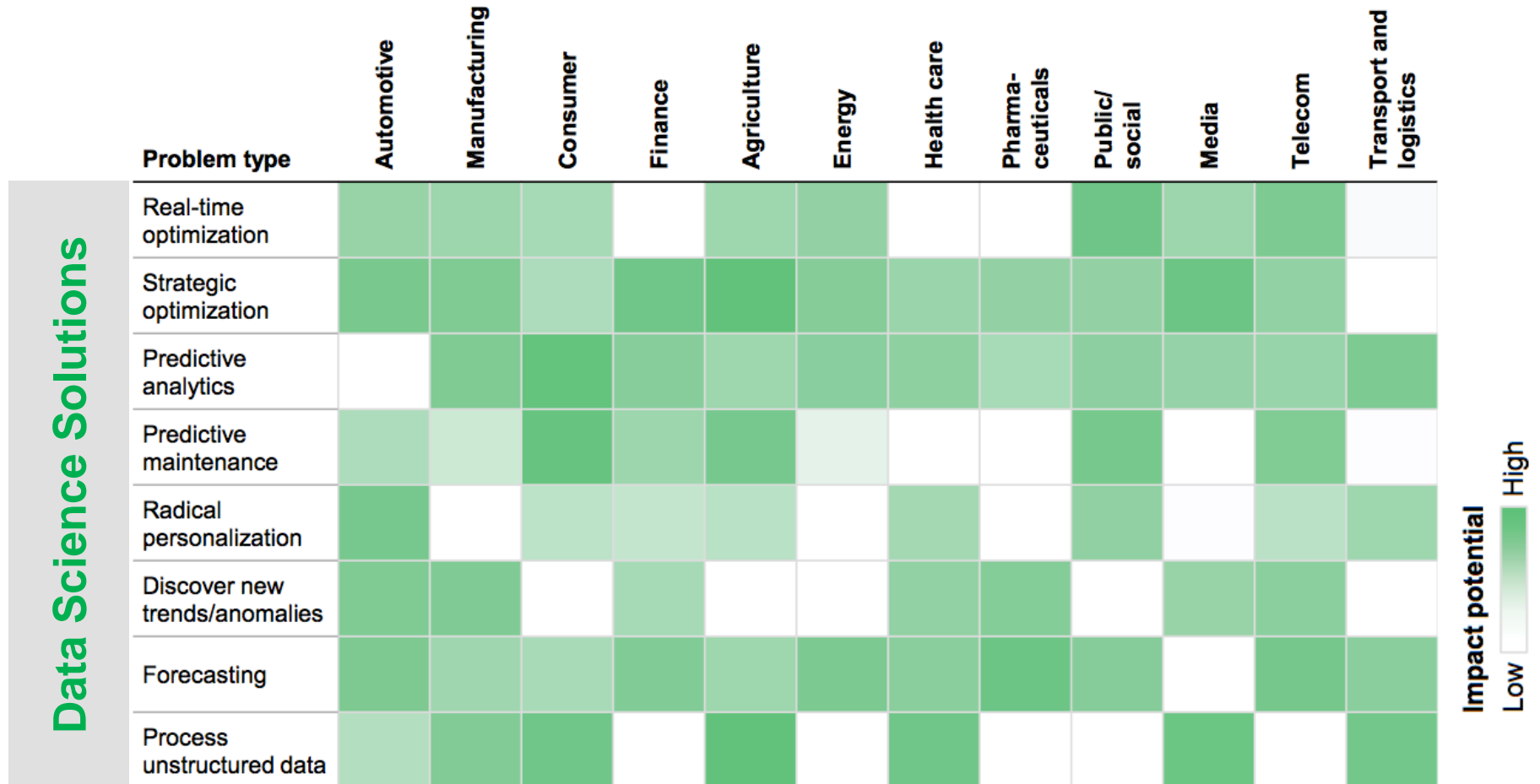
But what is Deep Learning?

*“Computers that learn without being **explicitly programmed**”*



Data Science Impact Across Industries and Use Cases

\$10s of Billions in each industry and use case



SOURCE: McKinsey Global Institute analysis

Challenges in delivering value with Data Science

Data

- Data resides in silos and difficult to access
- Unstructured and external data wasn't considered

Skills

- Data Science skills are in low supply and high demand

Governance

- Self-service isn't a reality, if the data isn't secure
- Understanding lineage and getting to a system of truth

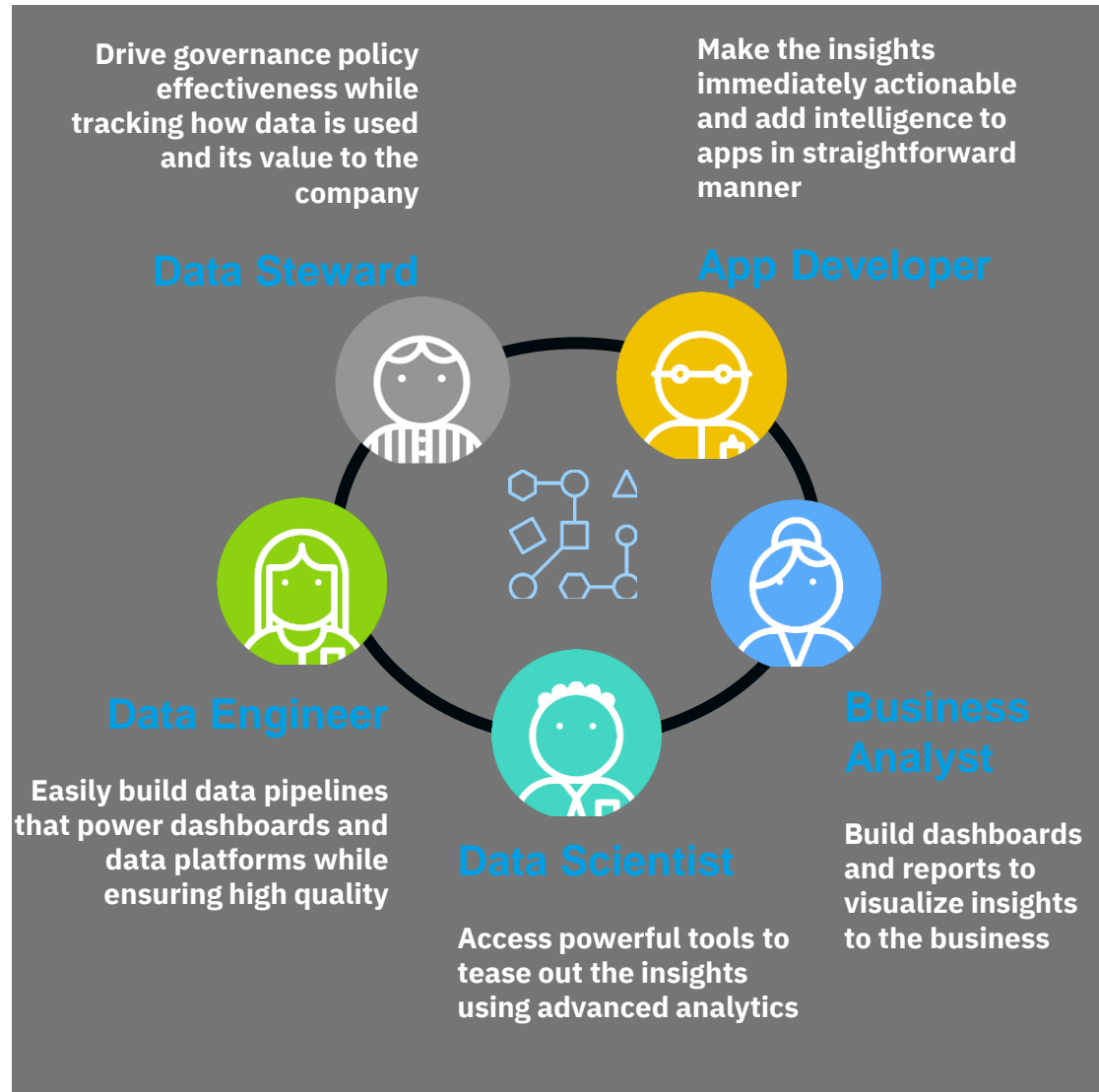
Infrastructure

- Need an environment that enables collaboration and deployment to production
- Discrete tools present barriers to progress

Watson Studio Platform

IBM Watson Studio Platform

An integrated platform of tools, services, and data that help companies or agencies accelerate their shift to be data-driven organizations.



Watson Studio supports end-to-end AI workflow

Build, train, deploy, and monitor at scale ML/DL workflows to infuse AI into the enterprise to drive innovation.

Connect &
Access Data

Search and Find
Relevant Data

Prepare Data
for Analysis

Build and Train
ML/DL Models

Deploy Models

Monitor, Analyze
and Manage

Connect and discover content from multiple data sources in the cloud or on premises. Bring **structured** and **unstructured** data to one toolkit.

Find data (structured, unstructured) and AI assets (e.g., ML/DL models, notebooks, Watson Data Kits) in the **Knowledge Catalog** with intelligent search and giving the right access to the right users.

Clean and prepare your data with **Data Refinery**, a tool to create data preparation pipelines visually. Use popular open source libraries to prepare unstructured data.

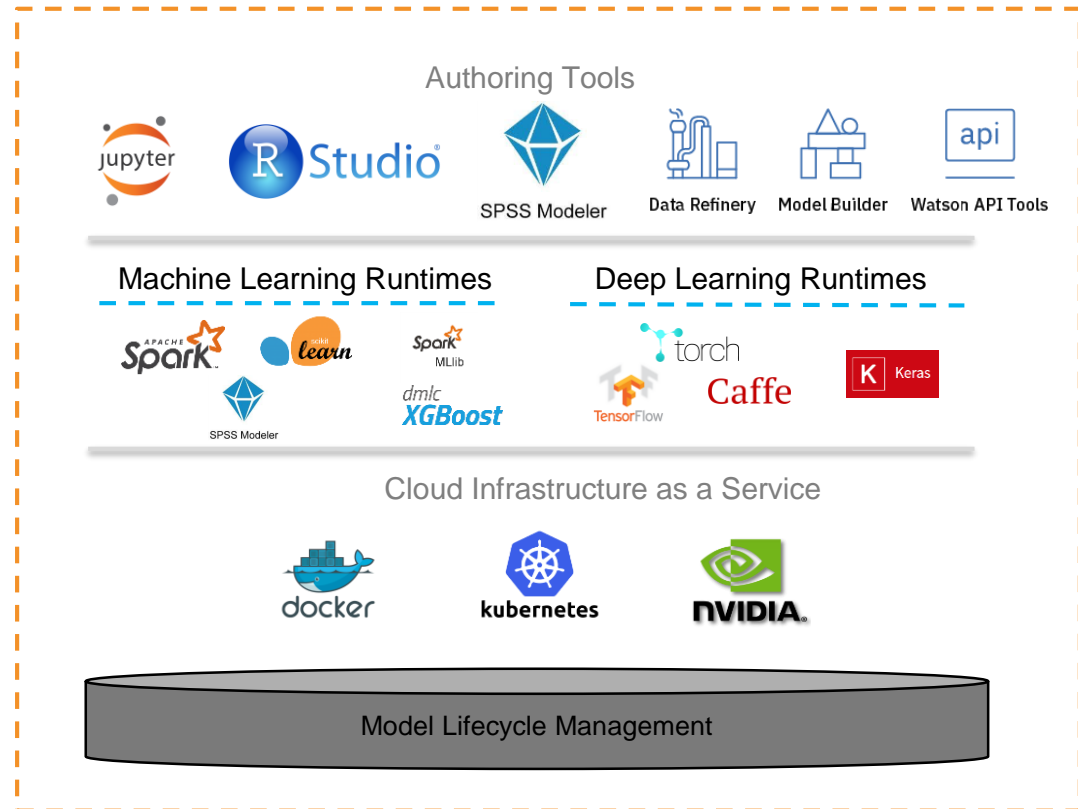
Democratize the creation of ML and DL models. Design your AI models **programmatically** or **visually** with the most popular **open source** and IBM ML/DL frameworks. Train at scale on **GPUs** and **distributed** compute

Deploy your models easily and have them **scale automatically** for online, batch or streaming use cases

Monitor the performance of the models in production and trigger automatic retraining and redeployment of models.

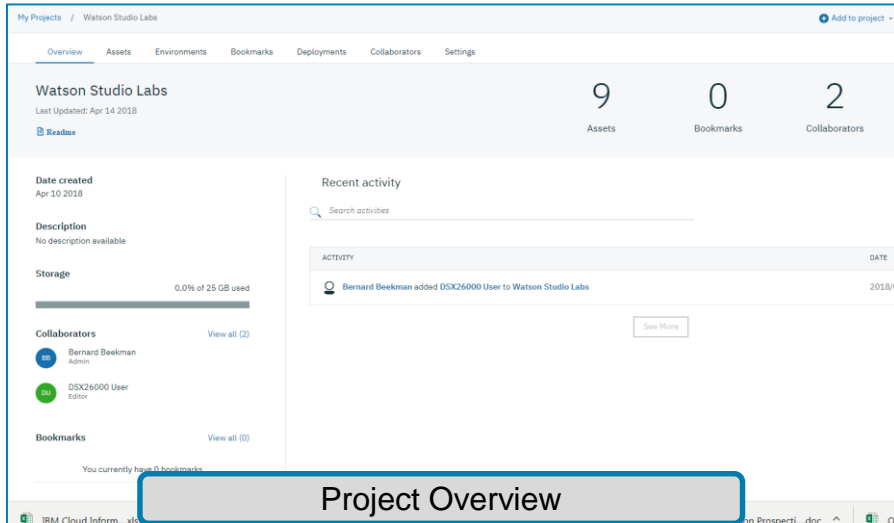
Watson Studio Tools

- Create, collaborate, deploy, and monitor
- Best of breed open source & IBM tools
- Code (R, Python or Scala) and no-code/visual modeling tools
- Open Source and IBM libraries/frameworks
- Fully managed service
- Container-based resource management
- Elastic pay as you go cpu/gpu power

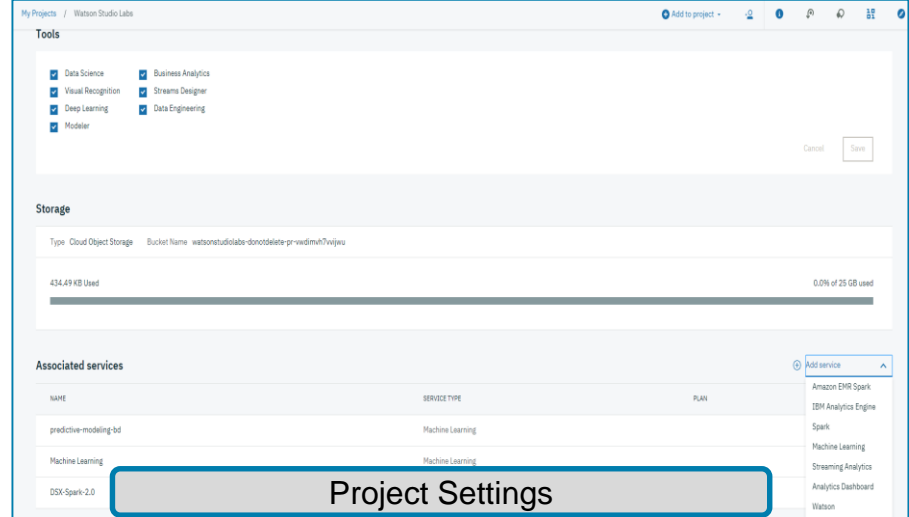


Watson Studio – Projects

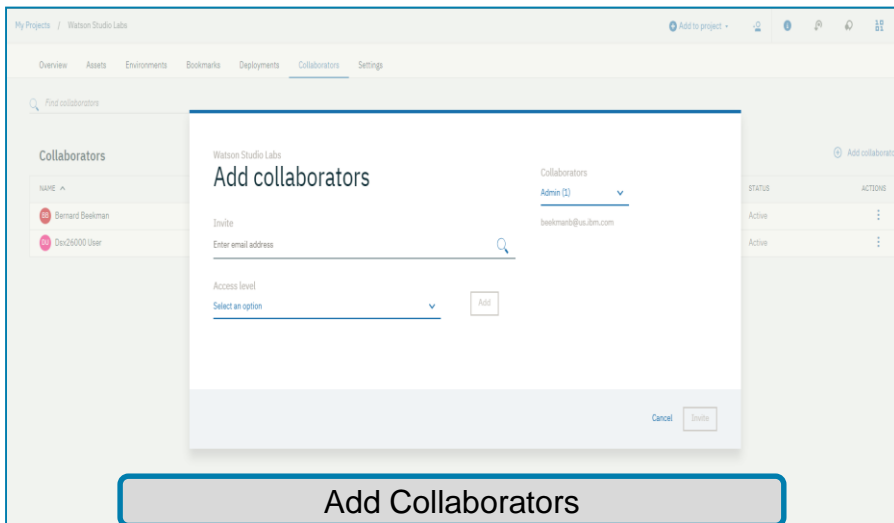
Making Data Science a Team Sport



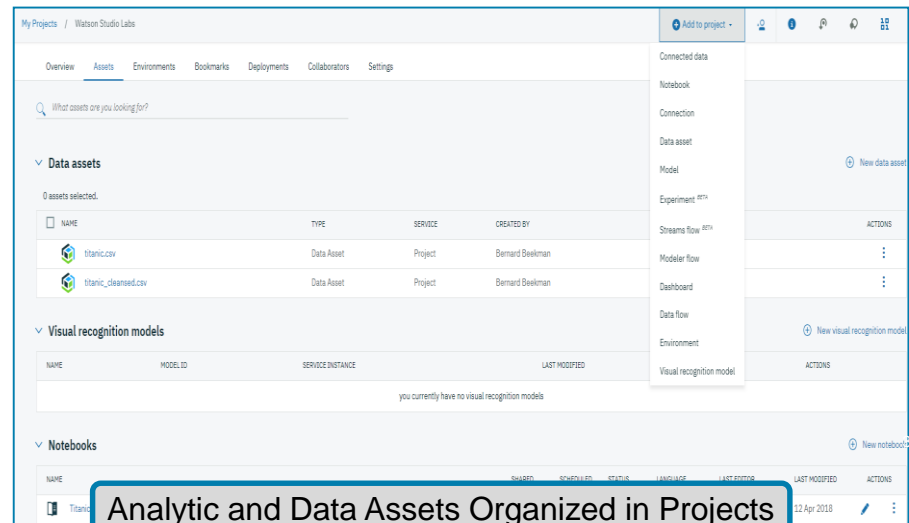
Project Overview



Project Settings



Add Collaborators



Analytic and Data Assets Organized in Projects

Watson Studio – Community Cards

Built-in learning to get started

Search results (355) Sort by: Most Related

Popular filters: Spark Deep Learning Brunel

Articles

ARTICLE	AUTHOR	DATE	TOPIC	FORMAT
Leaflet: Interactive web maps with R	RStudio Blog	May 20, 2016	Visualization	Web page
Open Sourcing 223GB of Driving Data –...	Udacity	Nov 09, 2016	Open Data	Web page
Learn TensorFlow and Deep Learning Together...	Big Data University	May 01, 2017	Deep Learning	Web page
sparklyr – R interface for Apache Spark	RStudio Blog	Oct 06, 2016	Analytics +1	Web page
This Week in Data Science (April 11, 2017)	Big Data University	Apr 14, 2017		
This Week in Data Science (October 18, 2016)	Big Data University	Oct 21, 2016		
Some Random Weekend Reading	R Views	Apr 10, 2017		
Using Deep Learning to Reconstruct...	Jeffrey Hetherly	Jun 26, 2017		

Search results (78) Sort by: Most Related

Popular filters: Spark Deep Learning Brunel

Notebooks

NOTEBOOK	AUTHOR	DATE	TOPIC
A TensorFlow regression model to predict...	IBM	Apr 06, 2018	Economy & Business
Access Db2 Warehouse on Cloud and Db2 with...	IBM	Mar 20, 2018	Economy & Business
Access MySQL with Python	IBM	Mar 27, 2018	Transportation
Access MySQL with R	IBM	Mar 27, 2018	Transportation
Access PostgreSQL with Python	IBM	Mar 20, 2018	Transportation
Access PostgreSQL with R	IBM	Mar 20, 2018	Transportation
Analyze Facebook Data Using IBM Watson and...	IBM	Mar 20, 2018	Transportation
Analyze accident reports on Amazon EMR Spark	IBM	Oct 12, 2017	Transportation

Search results (119) Sort by: Most Related

Popular filters: Spark Deep Learning Brunel

Tutorials

TUTORIAL	AUTHOR	DATE	LEVEL	TOPIC
What I Learned Implementing a Classifier...	Jean-Nicholas Houde	Apr 17, 2017	Intermediate	Machine Learning
Best packages for data manipulation in R	DataScience+	Jul 12, 2016	Intermediate	Data Science
Common Excel Tasks Demonstrated in Pandas	Practical Business Python	Sep 15, 2016	Beginner	Visualization
An Introduction to Stock Market Data...	Curtis Miller	Jun 13, 2017	Beginner	Visualization
Pulling and Displaying ETF Data	RStudio	Feb 09, 2017	Intermediate	
Super Fast String Matching in Python	van den Blog	Nov 20, 2017		
Understanding empirical Bayes estimation...	Variance Explained	Mar 13, 2018		
Brunel interactive visualizations in Jupyter...	Data Science Experience Blog	Jul 01, 2016		

Search results (295) Sort by: Most Related

Popular filters: Spark Deep Learning Brunel

Data Sets

DATA SET	AUTHOR	DATE	TOPIC
Adolescent fertility rate (births per 1,000...	IBM	May 22, 2016	Society
Agriculture, value added (% of GDP) by...	IBM	May 22, 2016	Economy & Business
Airbnb Data for Analytics: Amsterdam Calendar	IBM	Dec 20, 2016	Economy & Business
Airbnb Data for Analytics: Amsterdam Listings	IBM	Dec 20, 2016	Economy & Business
Airbnb Data for Analytics: Amsterdam Reviews	IBM	Dec 20, 2016	Economy & Business
Airbnb Data for Analytics: Antwerp Calendar	IBM	Dec 20, 2016	Economy & Business
Airbnb Data for Analytics: Antwerp Listings	IBM	Dec 20, 2016	Economy & Business
Airbnb Data for Analytics: Antwerp Listings...	IBM	Dec 20, 2016	Business

The best of open source and IBM Watson tools to create start-of-the-art data products

The screenshot shows the IBM Watson OpenScale JupyterLab environment. The top navigation bar includes 'IBM Watson' and links to 'Projects', 'Tools', 'Catalog', 'Community', and 'Services'. The main workspace is divided into three panes. The left pane shows a file explorer with 'My Projects' and 'demo99'. The center pane displays a Jupyter Notebook with a Python script for plotting a bar chart. The script uses a color palette and includes a title 'Tweets Country Distribution based on the User Profile'. Below the code, a bar chart is visible with labels for Germany, Mexico, Canada, India, Japan, and Spain. The right pane contains a 'Format' panel and a chat window with a message from 'ARMAND RUIZ G...'.

```

colors = ['gray'] * colors
lt.figure(figsize=(10,8))
lt.barh(y_pos, num_tweets, align='center', color=colors)
lt.yticks(y_pos, countries)
lt.xlabel('Number of Tweets')
lt.title('Tweets Country Distribution based on the User Profile')
lt.ylin(1, len(y_pos))
lt.show()

```

Tweets Country Distribution based on the User Profile

GERMANY

MEXICO

CANADA

INDIA

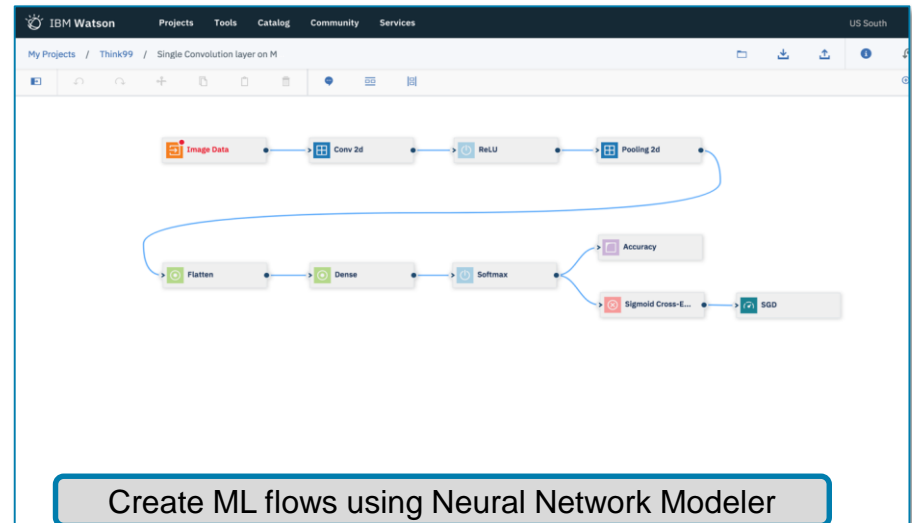
JAPAN

SPAIN

ARMAND RUIZ G... 7:49 AM

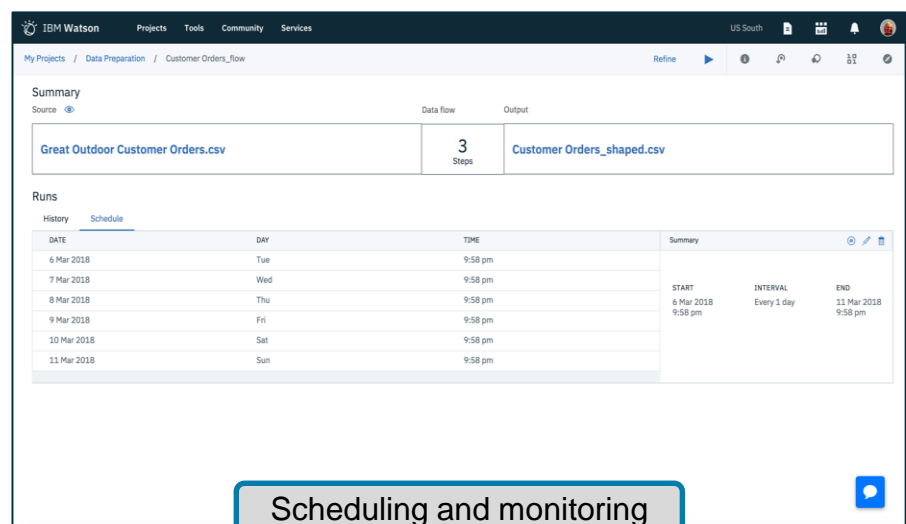
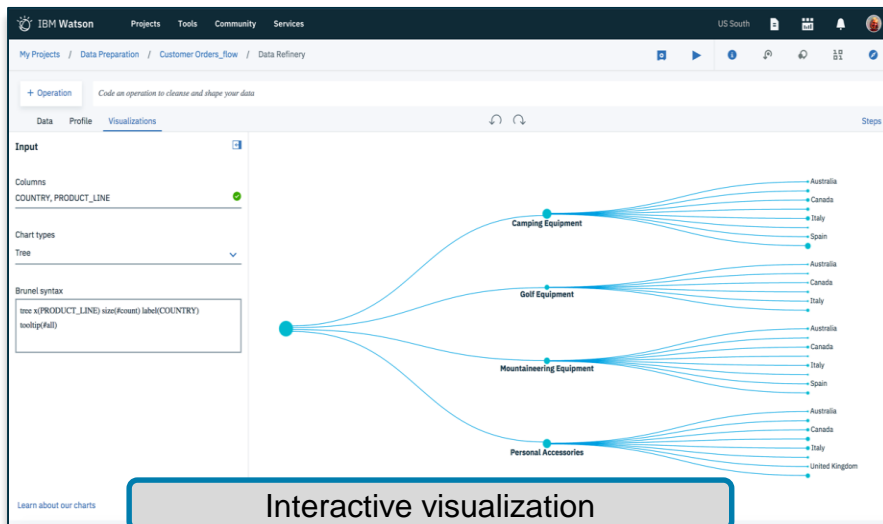
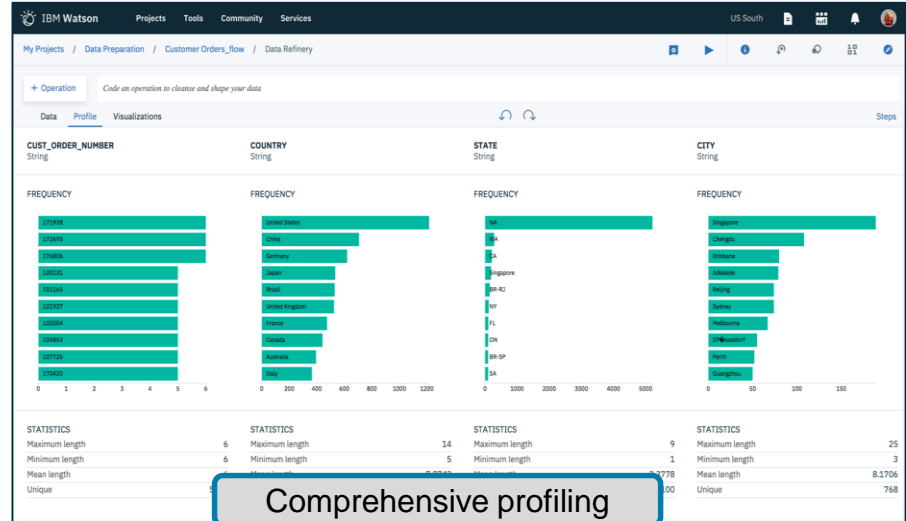
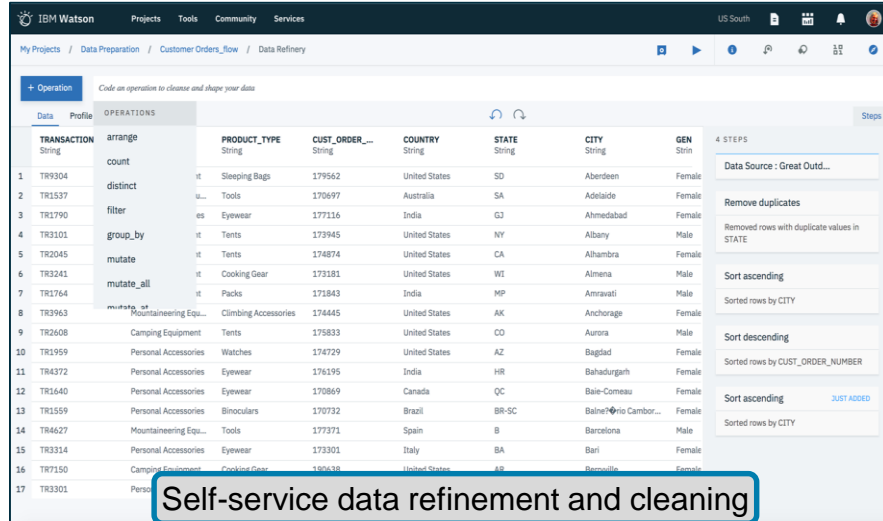
Great work!

Open Source tools – Jupyter and RStudio



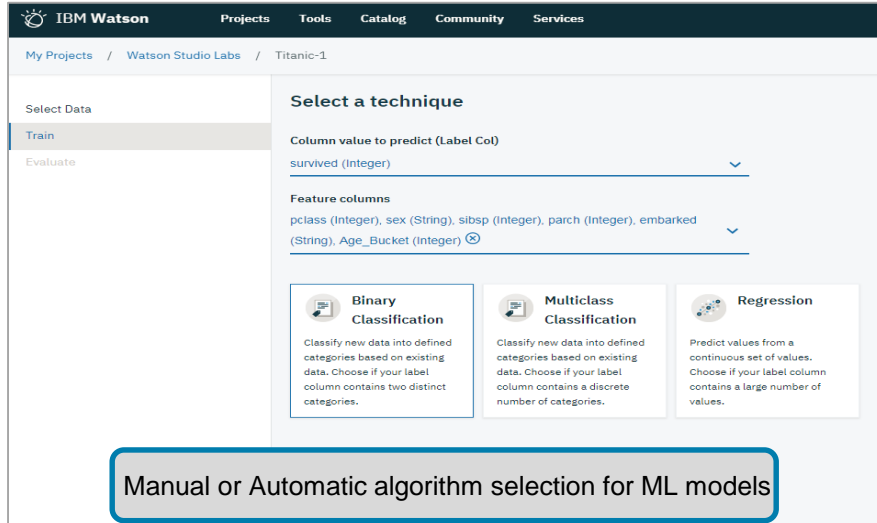
Watson Studio - Data Refinery

Making Data fit for use



Watson Studio – Watson Machine Learning

Simplifying deployment and management of ML models in production



Select a technique

Column value to predict (Label Col)
survived (integer)

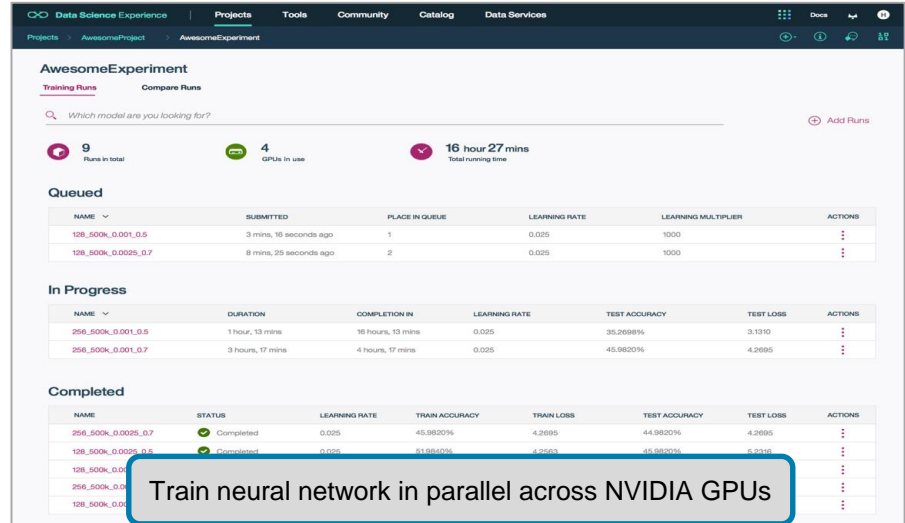
Feature columns
pclass (Integer), sex (String), sibsp (Integer), parch (Integer), embarked (String), Age_Bucket (Integer)

Binary Classification
Classify new data into defined categories based on existing data. Choose if your label column contains two distinct categories.

Multiclass Classification
Classify new data into defined categories based on existing data. Choose if your label column contains a discrete number of categories.

Regression
Predict values from a continuous set of values. Choose if your label column contains a large number of values.

Manual or Automatic algorithm selection for ML models



AwesomeExperiment

Training Runs Compare Runs

Which model are you looking for?

9 Runs in total 4 GPUs in use 16 hour 27 mins Total running time

Queued

NAME	SUBMITTED	PLACE IN QUEUE	LEARNING RATE	LEARNING MULTIPLIER	ACTIONS
128_500k_0.001_0.5	3 mins, 16 seconds ago	1	0.025	1000	
128_500k_0.0025_0.7	8 mins, 25 seconds ago	2	0.025	1000	

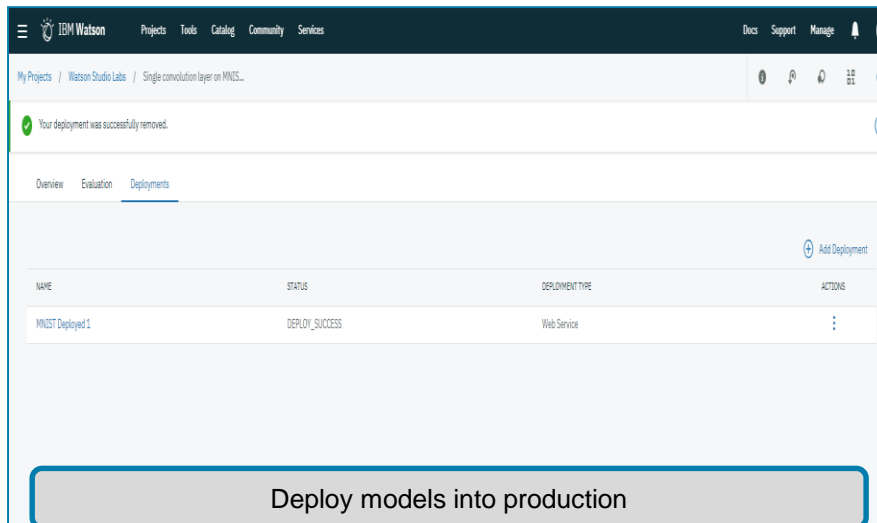
In Progress

NAME	DURATION	COMPLETION IN	LEARNING RATE	TEST ACCURACY	TEST LOSS	ACTIONS
256_500k_0.001_0.5	1 hour, 13 mins	16 hours, 13 mins	0.025	35.2688%	3.1310	
256_500k_0.001_0.7	3 hours, 17 mins	4 hours, 17 mins	0.025	45.9820%	4.2695	

Completed

NAME	STATUS	LEARNING RATE	TRAIN ACCURACY	TRAIN LOSS	TEST ACCURACY	TEST LOSS	ACTIONS
256_500k_0.0025_0.7	Completed	0.025	45.9820%	4.2695	44.9820%	4.2695	
128_500k_0.0025_0.6	Completed	0.025	61.0940%	4.2293	46.0820%	5.9245	
128_500k_0.001_0.5	Completed	0.025	35.2688%	3.1310	35.2688%	3.1310	
256_500k_0.001_0.5	Completed	0.025	35.2688%	3.1310	35.2688%	3.1310	
256_500k_0.001_0.7	Completed	0.025	45.9820%	4.2695	45.9820%	4.2695	
128_500k_0.001_0.7	Completed	0.025	45.9820%	4.2695	45.9820%	4.2695	

Train neural network in parallel across NVIDIA GPUs

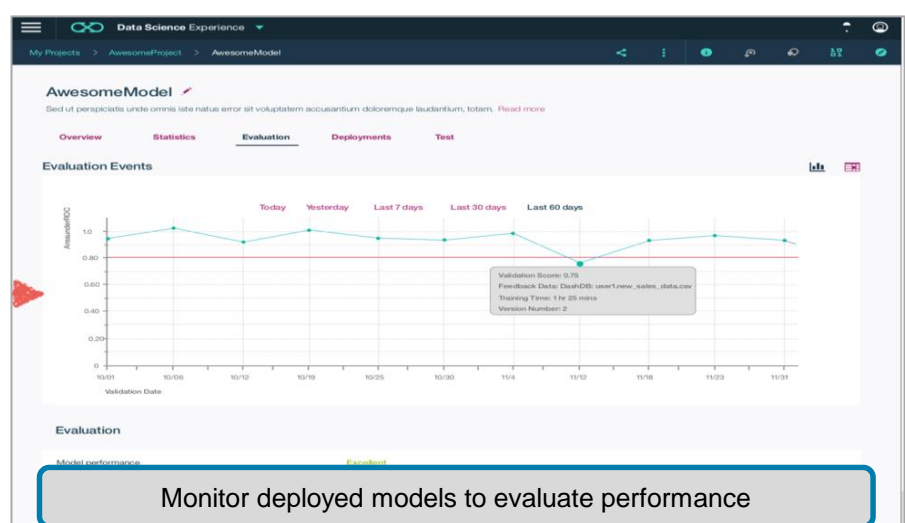


Deployment Events

NAME STATUS DEPLOYMENT TYPE ACTIONS

MNIST Deployed 1 DEPLOY_SUCCESS Web Service

Deploy models into production



AwesomeModel

Overview Statistics Evaluation Deployments Test

Evaluation Events

Area Under the Curve (AUC) vs Validation Date

Validation Score: 0.78
Feedback Data: DeskDB: user1new_sales_data.csv
Training Time: 1 hr 25 mins
Version Number: 2

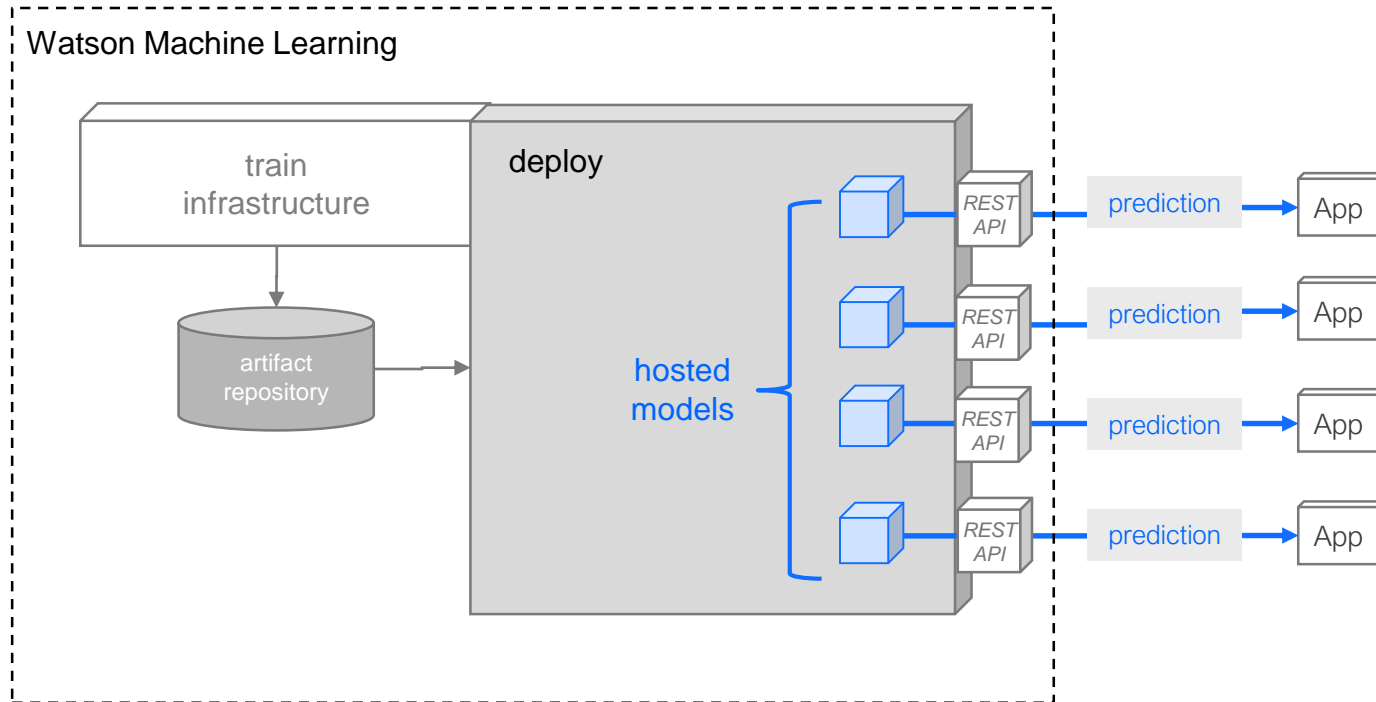
Evaluation

Model performance Evaluation

Monitor deployed models to evaluate performance

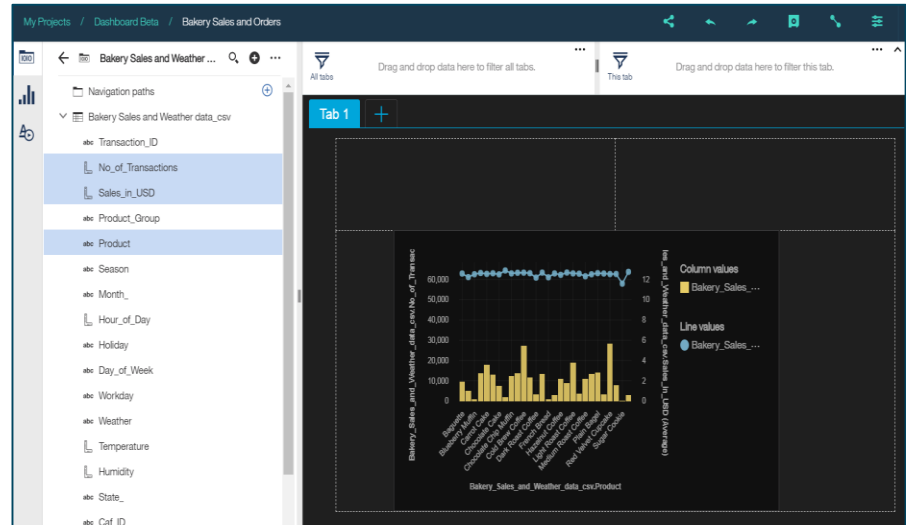
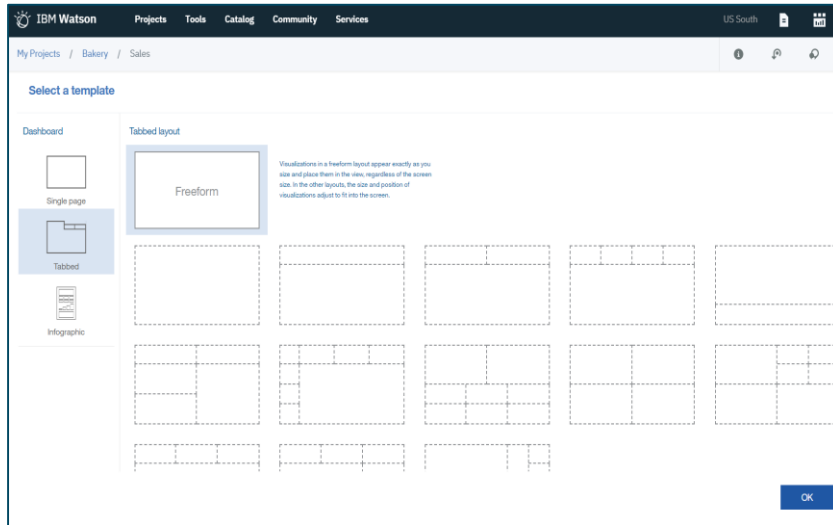
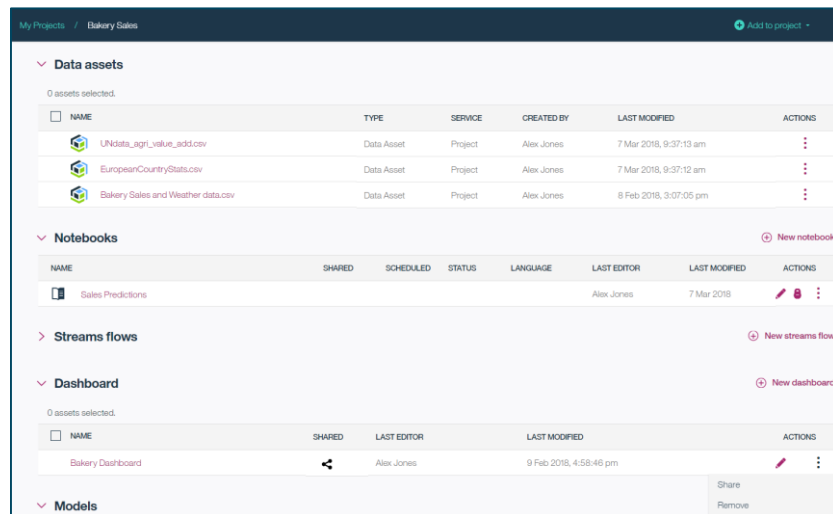
Watson Studio- Deploying Trained Models

Deploy your models within Watson Machine Learning



Watson Studio – Dynamic Dashboards

Making insights available to all

My Projects / Bakery Sales

Add to project

Data assets

0 assets selected.

NAME	TYPE	SERVICE	CREATED BY	LAST MODIFIED	ACTIONS
UNdata_agri_value_add.csv	Data Asset	Project	Alex Jones	7 Mar 2018, 9:37:13 am	
EuropeanCountryStats.csv	Data Asset	Project	Alex Jones	7 Mar 2018, 9:37:12 am	
Bakery Sales and Weather data.csv	Data Asset	Project	Alex Jones	8 Feb 2018, 3:07:05 pm	

Notebooks

NAME SHARED SCHEDULED STATUS LANGUAGE LAST EDITOR LAST MODIFIED ACTIONS

Sales Predictions Alex Jones 7 Mar 2018

Streams flows

0 assets selected.

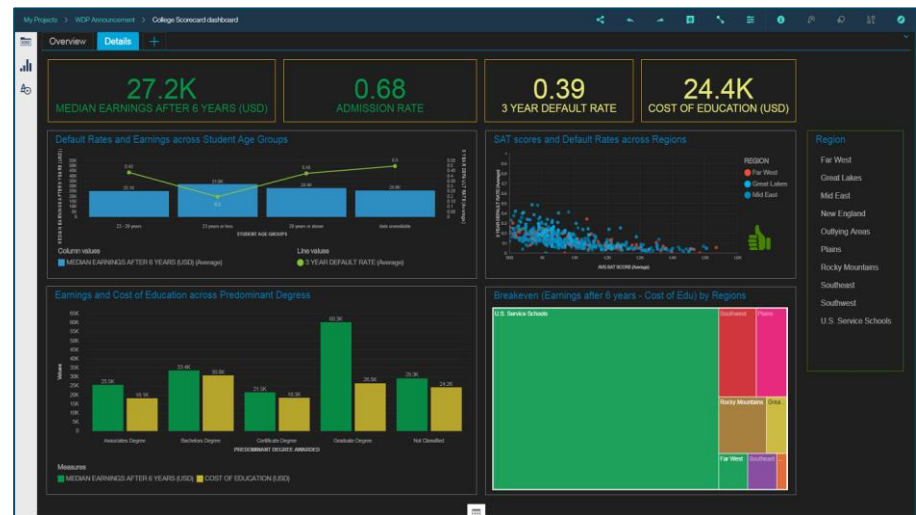
Dashboard

0 assets selected.

NAME	SHARED	LAST EDITOR	LAST MODIFIED	ACTIONS
Bakery Dashboard		Alex Jones	9 Feb 2018, 4:58:46 pm	

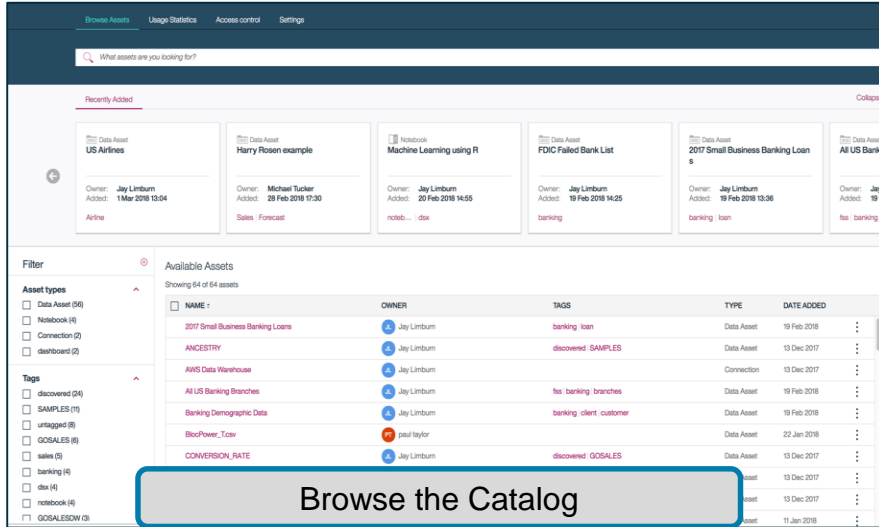
Models

Share Remove

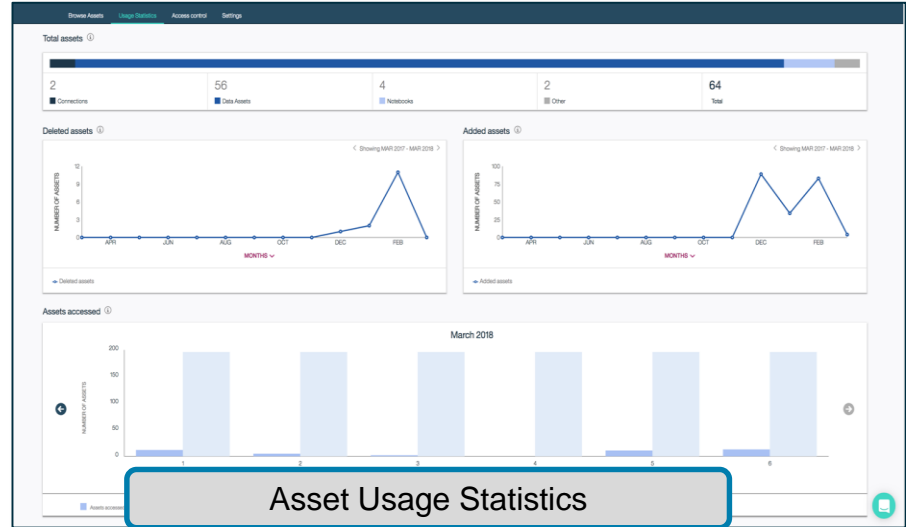


Watson Knowledge Catalog

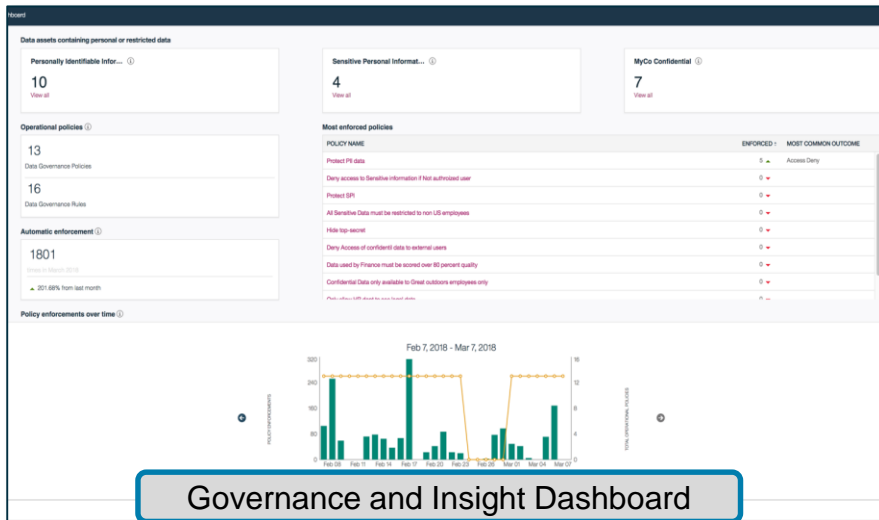
Unlock tribal knowledge and unleash knowledge workers



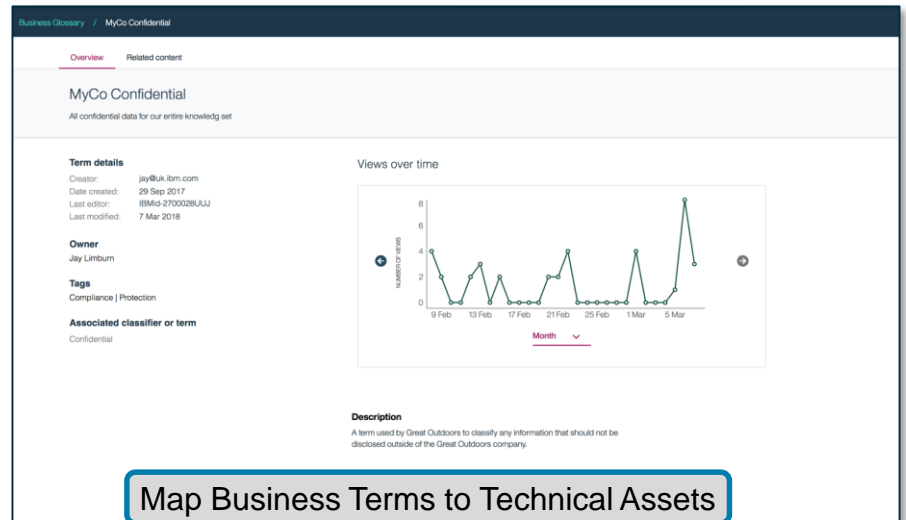
Browse the Catalog



Asset Usage Statistics



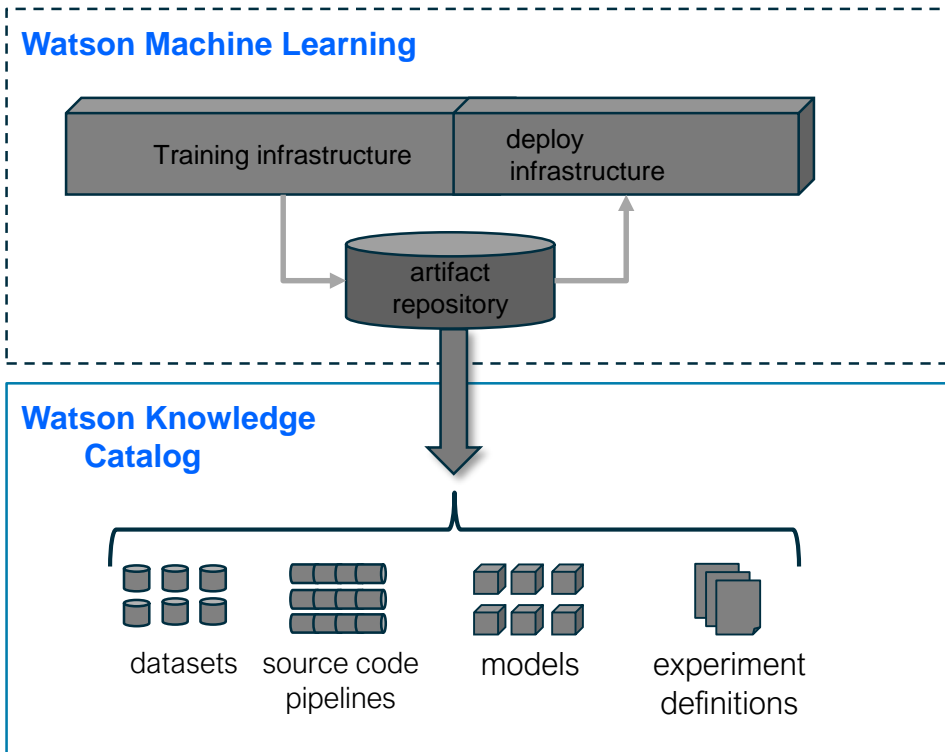
Governance and Insight Dashboard



Map Business Terms to Technical Assets

Watson Studio Model Lifecycle Management

Use the Watson Knowledge Catalog and Watson Studio to manage your AI assets or manage them yourself



Model Explanations

In May 2018, the General Data Protection Regulation (GDPR) takes effect and grants consumers the legal “right to explanation” from organizations that use algorithmic decision making.

Audit Trails

Tracking prediction to each model's unique heritage is critical to regulatory compliance. Enforcing access controls for model sharing and deployment ensure data security and application stability.

Watson Studio Takeaways

Integrated Collaboration Environment

- Data Scientists, Subject Matter experts, Business Analysts & Developers all in one environment to accelerate innovation, collaboration and productivity
- Built-in learning to get started or go the distance with advanced tutorials

Choice of Tools for the full AI lifecycle

- Best in-breed open source and IBM tools that support the end-to-end AI lifecycle
- Choice of code or no-code tools to build and train your own ML/DL models or easily train and customize pre-trained Watson APIs

Support for all levels of expertise

- Use Watson smarts and recommendations for the best algorithms to use given your data, OR
- Use the rich capabilities and controls to fine tune your models

Experiment centric DL workflow

- Monitor batch training experiments then compare cross-model performance without worrying about log transfers and scripts to visualize results.
- You focus on designing your neural networks. We'll manage and track your assets.

Model lifecycle & management

- Deploy models into production then monitor them to evaluate performance.
- Capture new data for continuous learning and retrain models so they continually adapt to changing conditions.

Integrated with Knowledge Catalog

- Intelligent discovery of data and AI assets that enables reuse & improves productivity
- Seamlessly integrated for productive use with Machine Learning and Data science
- Powerful governance tools to control and protect access to data

How does Watson Studio help fulfill the promise of your data?

Data

Puts every important data source at the fingertips of the teams that need it wherever resides

Governance

Enforces your policies without getting in the way of delivering insights

Skills

Makes the most of the data professionals you have and helps them grow and learn from each other as a team

Infrastructure

Brings all the tools in one place. Collaboration capabilities enables Data Science as a team sport.

Watson Studio is the new name for the IBM Data Science Experience on Cloud

Watson Knowledge Catalog is the new name for the IBM Data Catalog

Get started with Watson Studio at datascience.ibm.com

Lab Overview

Lab Overview

Use IBM's Watson Studio to create machine learning models and applications. Participants will be led through 5 labs (time permitting). Lab-1, Lab-2, Lab-3, and Lab-4 all use the Titanic data set, a common one used in Kaggle competitions. Lab-5 will use Watson Studio's Neural Network designer and Experiment Builder to build, train, and deploy a neural network model using the MNIST dataset.

- [Lab-1](#) - The first lab will use the Watson Machine Learning capability to create a machine learning model based on the Titanic data set. The model will be deployed in the IBM Cloud, and an application will be built that uses the deployed machine learning model to predict survivability given passenger characteristics.
- [Lab-2](#) - The second lab will guide participants in using the Watson Studio SPSS Modeler capability to explore, prepare, and model passenger data from the Titanic. The SPSS Modeler is a drag and drop capability to build machine learning pipelines.
- [Lab-3](#) - The third lab will leverage Spark machine learning (SparkML) in a Jupyter notebook to predict survivability using pyspark and a supervised learning model.
- [Lab-4](#) - The fourth lab features the Data Refinery tool a fully managed self-service data preparation facility.
- [Lab-5](#) - The fifth lab will use IBM's Neural Network designer and Experiment builder to build and train a simple Convolutional Neural Network. We will use the well-known MNIST dataset for training, test, and validation [MNIST](#)

Lab Tips

- Labs are all located in www.github.com/bleonardb3/NS repository. Environment set up is located in the repository [README](#) file.
- Instructions for each Lab are in the [README](#) file in the respective Lab folder.
- With cloud development frequent improvements are made in the user interface. We reviewed the lab instructions and made screen updates so they should be pretty faithful to the user interface. Small differences may occur but shouldn't get in the way of successfully completing the labs.
- You need to download the pdfs that are linked to the instructions for Lab-1, Lab-2, Lab4, Lab-5. You will click on the link and then click on the Download option. Otherwise, the links in the pdf will not work when viewing in the github interface.
- When downloading csv data files, make sure you follow the instructions to right click on the Raw button and use the Save link as ... option.
- Do not use Internet Explorer as the browser
- For Lab 3, you execute notebook cells by <Shift><Enter> when your cursor is in a code cell.

Lab 1 – Watson Machine Learning

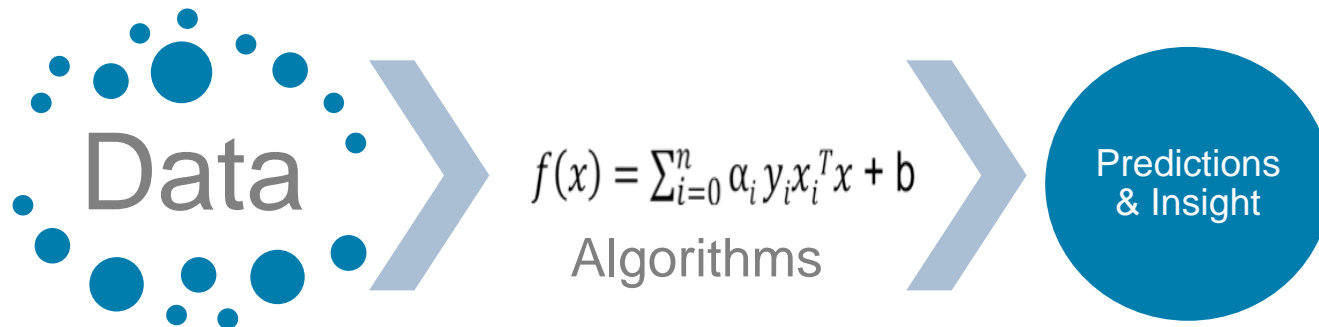
In this lab, you will use IBM's Watson Machine Learning GUI to train, evaluate, and deploy a Watson Machine Learning model based on the Titanic dataset.

Objectives:

- Upon completing the lab, you will:
 - Become familiar with the Watson Machine Learning GUI.
 - Train/Evaluate a machine learning model
 - Deploy a machine learning model.
 - Deploy an application that invokes the machine learning model service.

What is Machine Learning?

*“Computers that learn without being **explicitly programmed**”*
*“Using **algorithms** to understand patterns in data”*



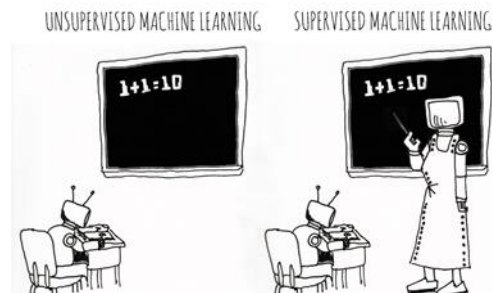
Categories of Machine Learning

■ Supervised learning

- The program is “trained” on a pre-defined set of “training examples”, which then facilitate its ability to reach an accurate conclusion when given new data
- The algorithm is presented with example inputs and their outcomes (labels)
- The goal is to learn a general rule that maps inputs to outputs

■ Unsupervised learning

- No labels are given to the learning algorithm, leaving it on its own to find structure (patterns and relationships) in its input



Categories of Machine Learning

Technique	Usage	Algorithms
Classification (or prediction)	<ul style="list-style-type: none">• Used to predict group membership (e.g., will this employee leave?) or a number (e.g., how many widgets will I sell?)	<ul style="list-style-type: none">• Decision Trees• Logistic Regression• Random Forests• Naïve Bayes• Linear Regression• Lasso Regressionetc
Segmentation	<ul style="list-style-type: none">• Used to classify data points into groups that are internally homogenous and externally heterogeneous.• Identify cases that are unusual	<ul style="list-style-type: none">• K-means• Gaussian Mixture• Latent Dirichlet allocationetc
Association	<ul style="list-style-type: none">• Used to find events that occur together or in a sequence (e.g., market basket)	<ul style="list-style-type: none">• FP Growth

Preprocessing: Matrix for Machine Learning

Known as:

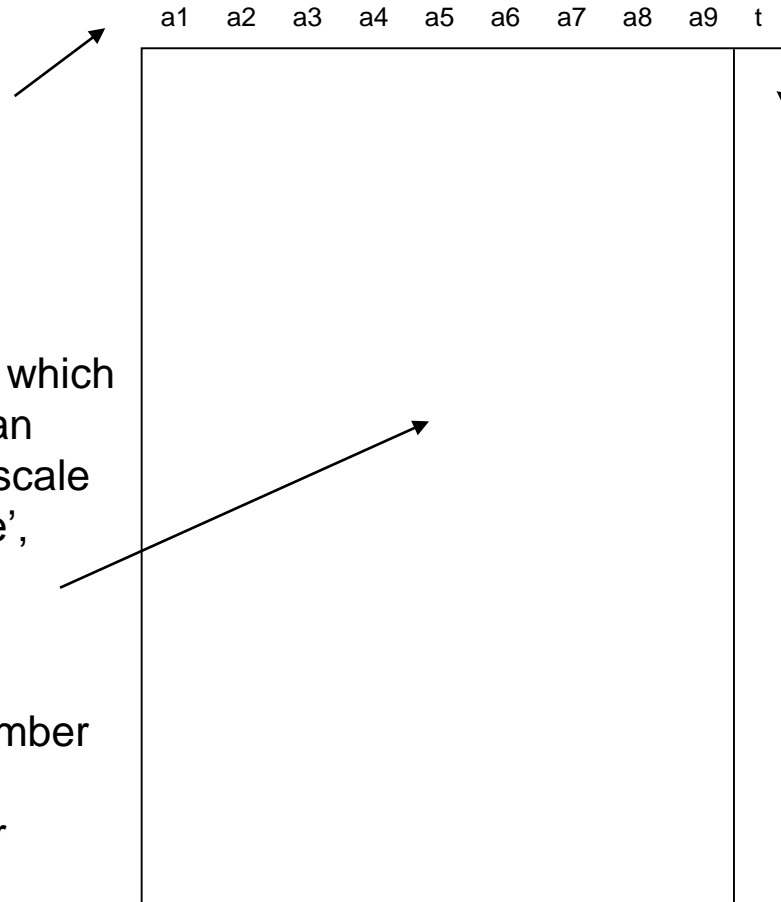
- Attributes
- Features
- Predictor variables
- Explanatory variables

Scale variables:

- Continuous variables, which can be measured on an interval scale or ratio scale
- 'Weight', 'Temperature', 'Salary', etc...

Categorical variables:

- Data with a limited number of distinct values or categories (nominal or ordinal)
- 'Hair color', 'Gender', 'Grape varieties', etc...



Known as:

- Label
 - Target variable
 - Dependent variable
- Scale or Categorical

Training, testing, & validation sets

- **During the model development process, supervised learning techniques employ **training** and **testing** sets and sometimes a **validation** set.**
 - Historical data with known outcome
 - Data is randomly split into training, testing, and/or validation sets (mutually exclusive records)
- **Why?**
 - Training set
 - Build the model
 - Tune the parameters
 - Testing set
 - Assess model quality during training/tuning process
 - Avoid overfitting the model to the training set
 - Validation set
 - Estimate accuracy or error rate of model after tuning
 - Used to compare multiple models

Demo Data - Titanic



Variable Descriptions:

survival	Survival (0 = No; 1 = Yes)
pclass	Passenger Class (1 = 1st; 2 = 2nd; 3 = 3rd)
name	Name
sex	Sex
age	Age
sibsp	Number of Siblings/Spouses Aboard
parch	Number of Parents/Children Aboard
ticket	Ticket Number
fare	Passenger Fare
cabin	Cabin
embarked	Port of Embarkation (C = Cherbourg; Q = Queenstown; S = Southampton)

PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
1	0	3	Braund, Mr. Owen Harris	male	22	1	0	A/5 21171	7.25		S
2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Thayer)	female	38	1	0	PC 17599	71.2833	C85	C
3	1	3	Heikkinen, Miss. Laina	female	26	0	0	STON/O2. 3101282	7.925		S
4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35	1	0	113803	53.1	C123	S
5	0	3	Allen, Mr. William Henry	male	35	0	0	373450	8.05		S
6	0	3	Moran, Mr. James	male		0	0	330877	8.4583		Q
7	0	1	McCarthy, Mr. Timothy J	male	54	0	0	17463	51.8625	E46	S
8	0	3	Palsson, Master. Gosta Leonard	male	2	3	1	349909	21.075		S
9	1	3	Johnson, Mrs. Oscar W (Elisabeth Vilhelmina Berg)	female	27	0	2	347742	11.1333		S
10	1	2	Nasser, Mrs. Nicholas (Adele Achem)	female	14	1	0	237736	30.0708		C

Lab 2 – SPSS Modeler

In this lab, you will use the Watson Studio SPSS Modeler capability to explore, prepare, and model passenger data from the Titanic. The SPSS Modeler is a drag and drop capability to build machine learning pipelines.

Objectives:

- Upon completing the lab, you will:
 - Become familiar with the Watson Studio SPSS Modeler capability
 - Profile the Titanic data set
 - Explore the Titanic data set with visualizations
 - Cleanse and Transform the data
 - Train/Evaluate a machine learning mode.

Lab 3 – Jupyter Notebook and SparkML

In this lab, you will use IBM's Watson Studio to create a Jupyter notebook to examine the principles of Spark Machine Learning using the Titanic dataset. You will build a model to predict who survived -- and who did not.

Objectives:

- Upon completing the lab, you will know how to:
 - Create a Jupyter notebook from a URL
 - Load data from a URL
 - Examine the data in PixieDust
 - Examine and shape the data for use in an ML model
 - Build a Pipeline for a Logistic Regression model
 - Tune the model for maximal effectiveness

Spark ML

- **Spark ML is Spark's machine learning (ML) library**
- **Goal is to make machine learning scalable and easy**
 - No need to understand the detailed math!
- **Divides into two packages:**
 - spark.mllib contains the original API built on top of RDDs
 - spark.ml provides higher-level API built on top of DataFrames for constructing ML pipelines
 - A pipeline is a series of stages where each stage either transforms, or runs through a machine learning algorithm.
- **Using spark.ml is recommended because with DataFrames the API is more versatile and flexible**
 - spark.mllib will continue to be supported

Spark ML Pipeline Terminology

Spark ML standardizes APIs for machine learning algorithms to make it easier to combine multiple algorithms into a single pipeline, or workflow

- **DataFrame**: Spark ML uses DataFrame from Spark SQL as an ML dataset, which can hold a variety of data types
- **Transformer**: A Transformer is an algorithm which can transform one DataFrame into another DataFrame
- **Estimator**: An Estimator is an algorithm which can be fit on a DataFrame to produce a Transformer
- **Pipeline**: A Pipeline chains multiple Transformers and Estimators together in a sequence to specify an ML workflow
- **Parameter**: All Transformers and Estimators share a common API for specifying parameters

Demo Flow

- **Read in Titanic dataset as a DataFrame**
 - Drop unwanted columns and rows with null or invalid data
 - Label the data (“Survived”)
- **Data Analysis**
 - Visualizations
- **Feature Engineering**
 - StringIndexer (Sex and Embarked variables)
 - Bucketizer (Age and Fare variables)
 - VectorAssembler
 - Normalizer
- **Create a Pipeline**
- **Split Ratings data into Training (80%) and Test (20%) datasets**
 - Cache the resulting DataFrames



Demo Flow (continued)

- **Fit the Pipeline to the Test data set**
 - Logistic Regression
- **Evaluate the resulting predictions**
 - Area under the ROC curve
- **Tune the model (hyperparameters)**
 - Build Parameter Grid
 - Cross-evaluate to find the best model
- **Make improved predictions using the cross-validated model**
- **Make prediction on an imaginary passenger**
- **Show how to easily reuse completed work using a different machine learning algorithm**
 - Random Forest

Lab 4 – Data Refinery

In this lab, you will use the Watson Studio Data Refinery to profile data, visualize data, and prepare data for modeling.

Objectives:

- Upon completing the lab, you will know how to:
 - Profile the data to help determine missing values
 - Visualize the data to gain a better understanding
 - Prepare the data for modeling
 - Run the sequence of data preparation operations on the entire data set.

Lab 5 – Neural Network

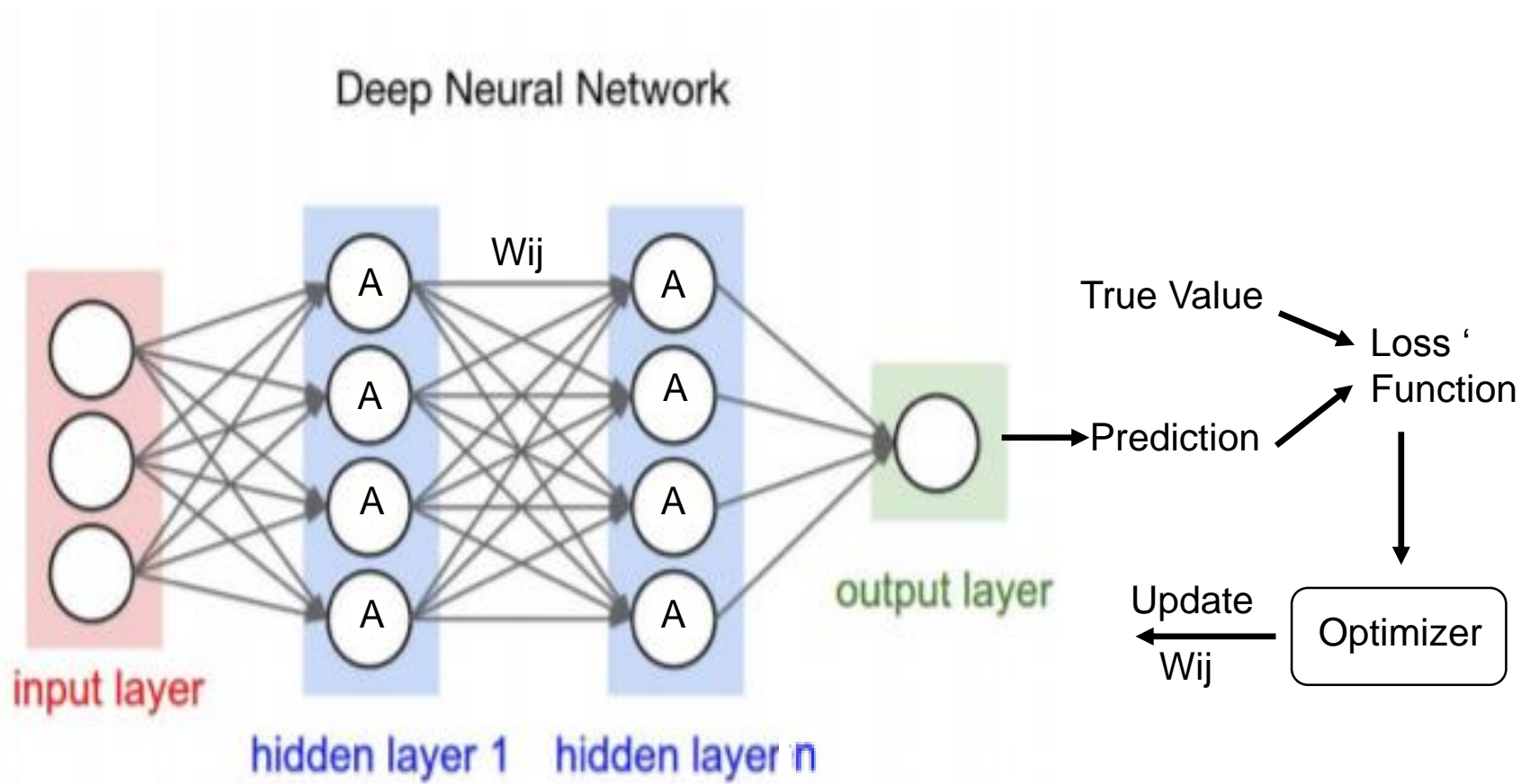
This lab will use the [MNIST](#) computer vision data set to train a convolutional neural network (CNN) model to recognize handwritten digits. The Watson Studio neural network flow editor, Watson Studio experiment builder and the Watson Machine Learning component will be used to build, train, and save the trained model.

Objectives:

- Upon completing the lab, you will know how to:
 - Create Cloud Object Storage buckets to contain the input and result files
 - Create a neural network design from an example using the flow editor
 - Use the experiment builder used to set up a training definition to train the neural network model
 - Monitor the training progress and results.
 - Save the trained model.

Neural Network

- Inspired by the way the human brain works.



W_{ij} - weights

A – Activation Function

Neural Network

Modeling

- **Originated in 1940s**
- **Became very popular this decade**
 - Hardware – GPUs, Storage
 - Availability of Large Datasets for Training
 - Better performing algorithms.
- **Especially useful for human perception type task**
 - Image Classification
 - Object Recognition
 - Speech Recognition
 - Natural Language Understanding
 - Machine Translation
 - ...

Neural Network Modeler

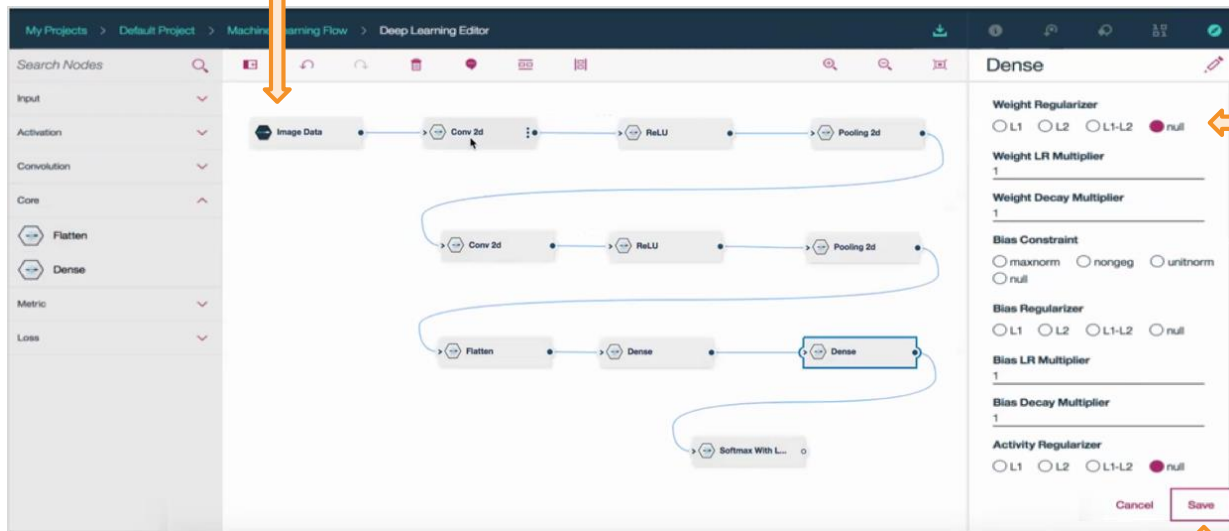
An intuitive drag-and-drop, no-code interface for designing neural network structures using the most popular deep learning frameworks. Quickly capture your network design then single click export for experimental optimization.

Supported Frameworks



Drag-and-drop
network layers

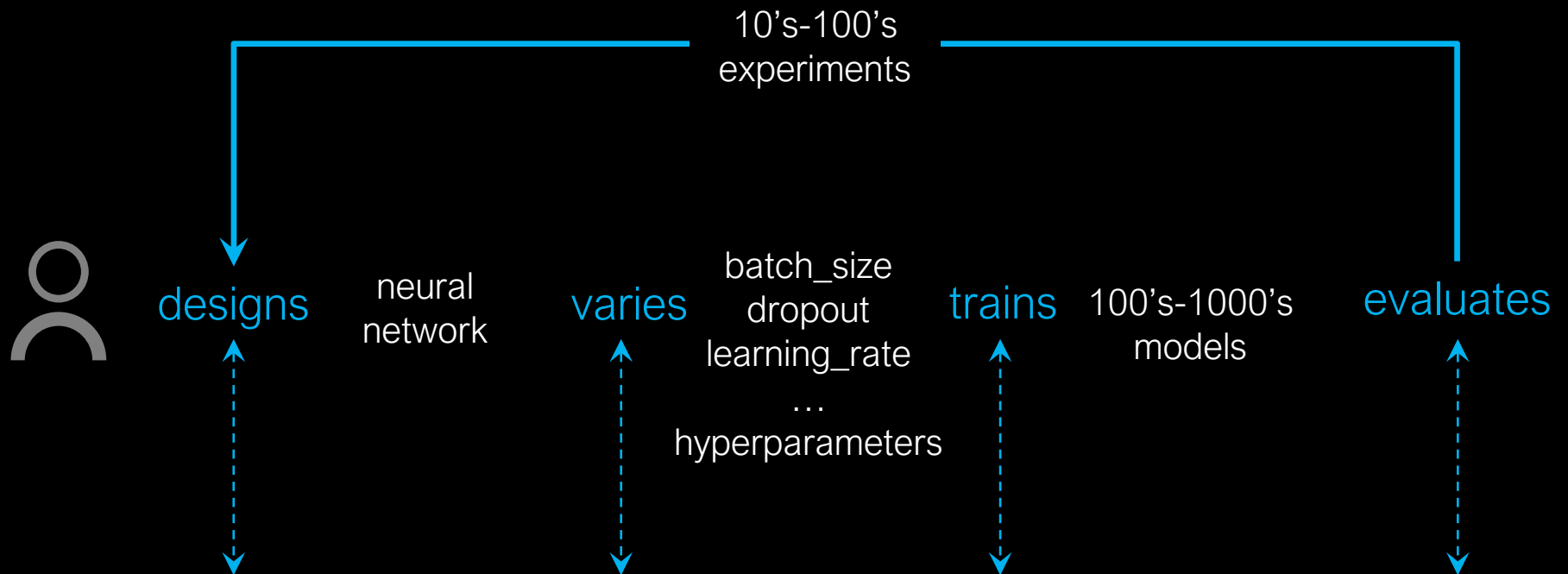
Real-time validation of network
flow



- Generate CPU or GPU compatible code

- Save as popular framework code
- Export as a python notebook
- Execute as batch experiment

Experiment Builder



Experiment Builder
supports the end-to-end workflow