

Data Science Workshop

An Experiential Journey with Data to Inspire *Your* Work

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Neeraj Madan

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Feb 2nd, 2022

Data Science Workshop

An Experiential Journey with Data to Inspire *Your* Work

Data Science Workshop

09:15 – 11:30 US CT

Data Science Workshop: An Experiential Journey with Data to Inspire Your Work

The Data Science Virtual Workshop, "An Experiential Journey with Data to Inspire Your Work", will make you think differently about data and how it can solve problems! This workshop includes surprising use cases that will make you think differently about data, sometimes laugh and hopefully inspire your own work to discover actionable insights in the mounds of data available. The use cases and introductory material will be followed by a hands-on experiential journey addressing a common challenge across industries – how to improve the customer experience. The most valuable part of this workshop is that it is designed to help you gain experience and relate it to your work – so at the end you have a plan of action on how you can make data more useful in your organization to solve a key challenge.

"Improving Customer Experiences with Real-Time Insights", will be used as an example during the workshop. This experiential session will include a step by step journey based on how data science is helping companies to predict the customer experience journey and proactively address the issues, leading to the improvement of Net Promoter Score. The session will also highlight the importance of using AI Canvas, CRISP-DM (Cross Industry Standard Process for Data Mining) and Agile in Data Science projects.

The methodology involves consuming historical NPS data; using machine learning and artificial intelligence to identify the most important features and creating an algorithm to predict the customer experience.

Pre-Workshop Setup Instructions

It is important to have the *setup done prior to the workshop* so that the time during the session is focused on the content and experiential journey. If you encounter difficulty during the set up, reach out to Neeraj Madan or George Stark with your questions at ds-workshop@opengroup.org.

Please note that while there are many tools that could be used for building predictive analytics and machine learning solutions, this workshop will be based on IBM Watson Studio for the experiential session.

Facilitators: [Neeraj Madan](#), [Maureen Norton](#), [Upkar Lidder](#)

Agenda

Section	Time (US CT)
Getting Started <i>a. Session Introduction and Expectation Setting</i> <i>b. Data Science Introduction</i> <i>c. Predictive Analytics and Machine Learning Solutions</i> <i>d. Create a Project</i> <<Break (10 mins)>>	09:15 am – 10:00 am
Hands on Experiential Journey (Net Promoter Score Example) <i>a. Business understanding: Exercise 1: Identify an opportunity in your business context and document</i> <i>b. Data understanding: Exercise 2: What data set would you gather to work the problem statement</i> <i>c. Data preparation: Exercise 3: How would you prepare the dataset and what challenges do you foresee?</i> <<Break (10 mins)>> <i>d. Modeling: Exercise 4: What modeling techniques would you attempt and why?</i> <i>e. Evaluation: Exercise 5: What metrics would you use to evaluate your model performance?</i> <i>f. Deployment: Exercise 6: How do you plan to consume the outputs of the model?</i>	10:00 am – 11:30 am

Let's talk about data

Is there a source of data that has information about

- ANY topic
- ANY where
- ANY time

Is there a source that has their finger on the pulse of what people think at any moment in time?



Twitter

Soggy Fries



Let's talk about data

What other types
of data can be
used to drive
deeper insights?

WEATHER

The image features a dramatic sky scene. The bottom half is filled with large, white, fluffy clouds. Above the clouds is a bright blue horizon line. The top half of the image is a dark blue sky with faint, wispy clouds and several thin, curved blue lines that resemble light trails or orbits. The word "WEATHER" is written in large, white, bold, sans-serif capital letters across the middle of the image, partially overlapping the clouds and the sky.

Four Common Data Science Models

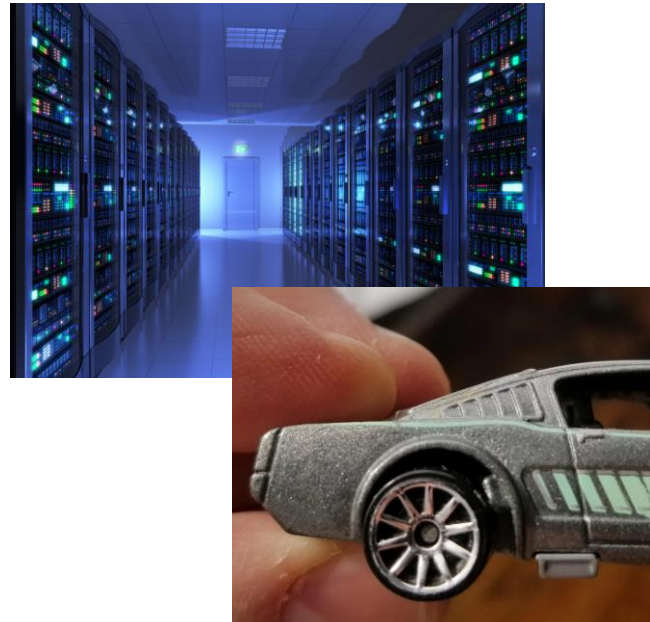
Risk Assessment

Create a “Screening Model” to identify “threats”. Threats can be any sort of fraudulent activity (e.g., credit transaction, passenger screening, ability to purchase, altered video/photo, Fake/Real news)



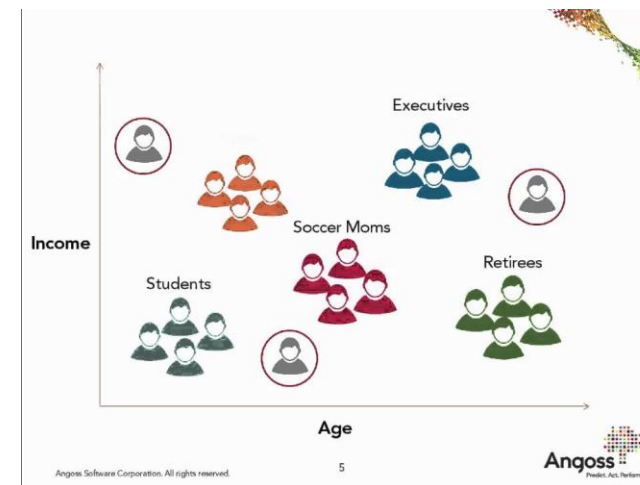
Quality/Defect

Identify problematic components, defects in a product (e.g., server failure, castings, ATM Machines)



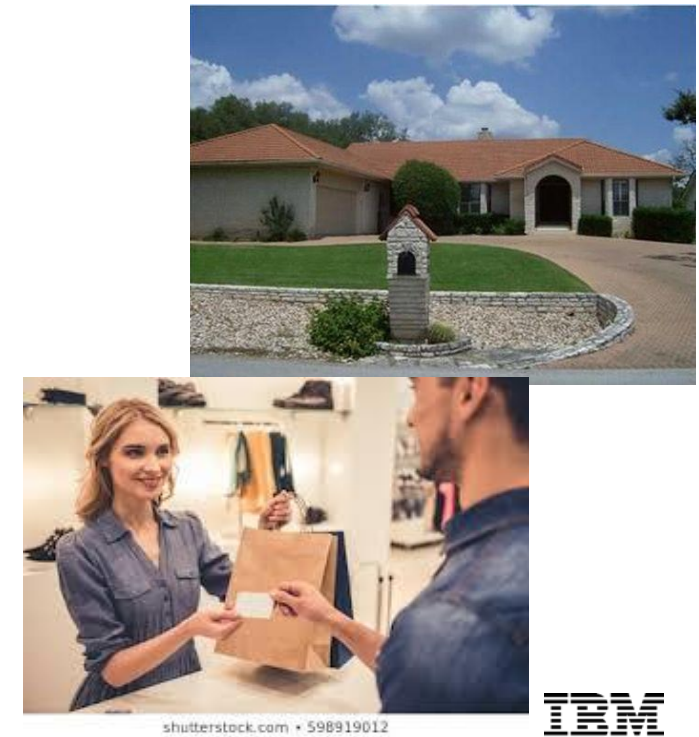
Business Value/ Customer Satisfaction

Identifies which customers have the most potential business value based on their characteristics and activities. Which customers are likely to be happy? Which will be promoters?

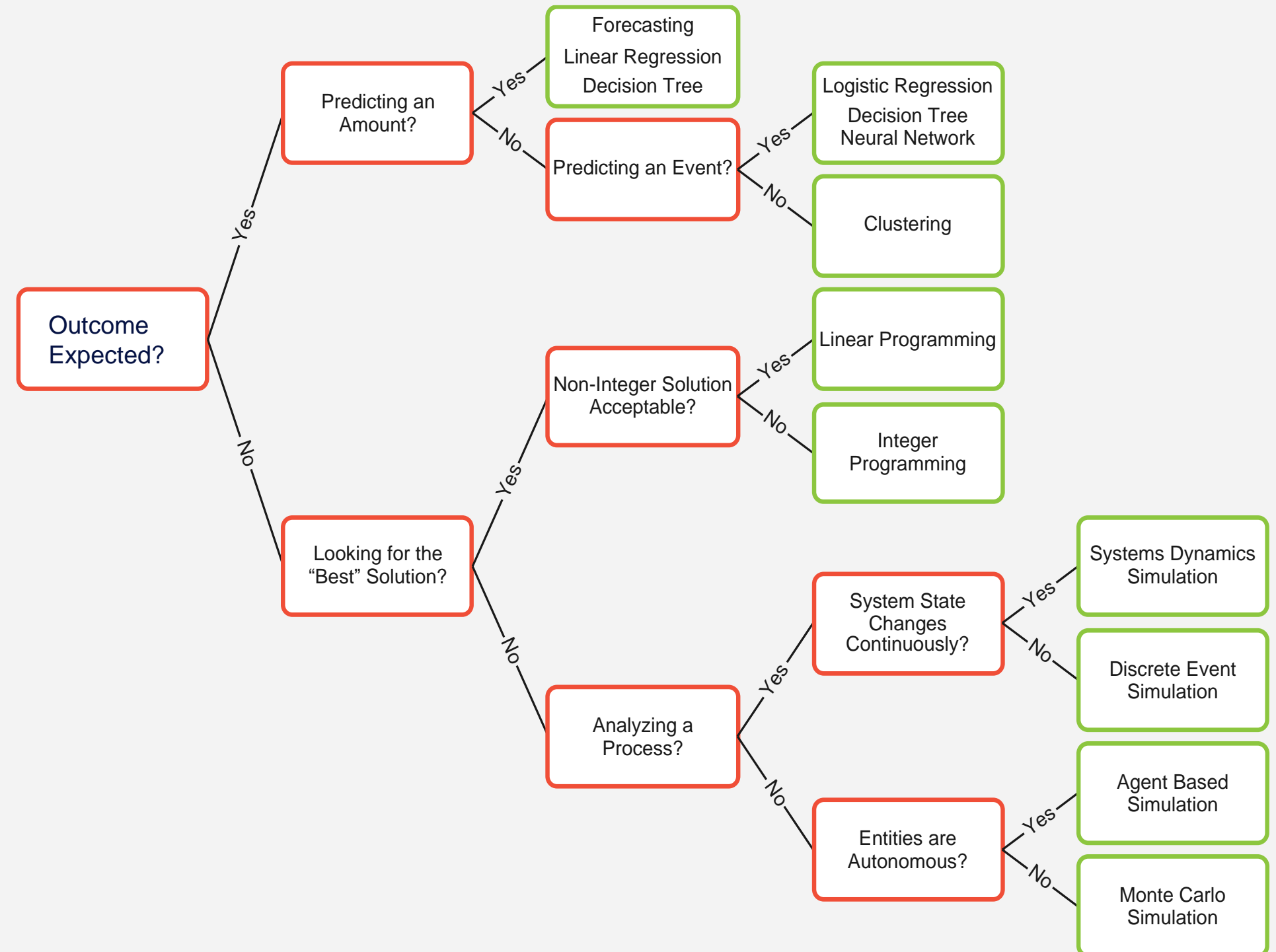


Price/Cost/Value

Predict value (e.g., home/ rental prices, value of retail transaction, number of issues, etc.)

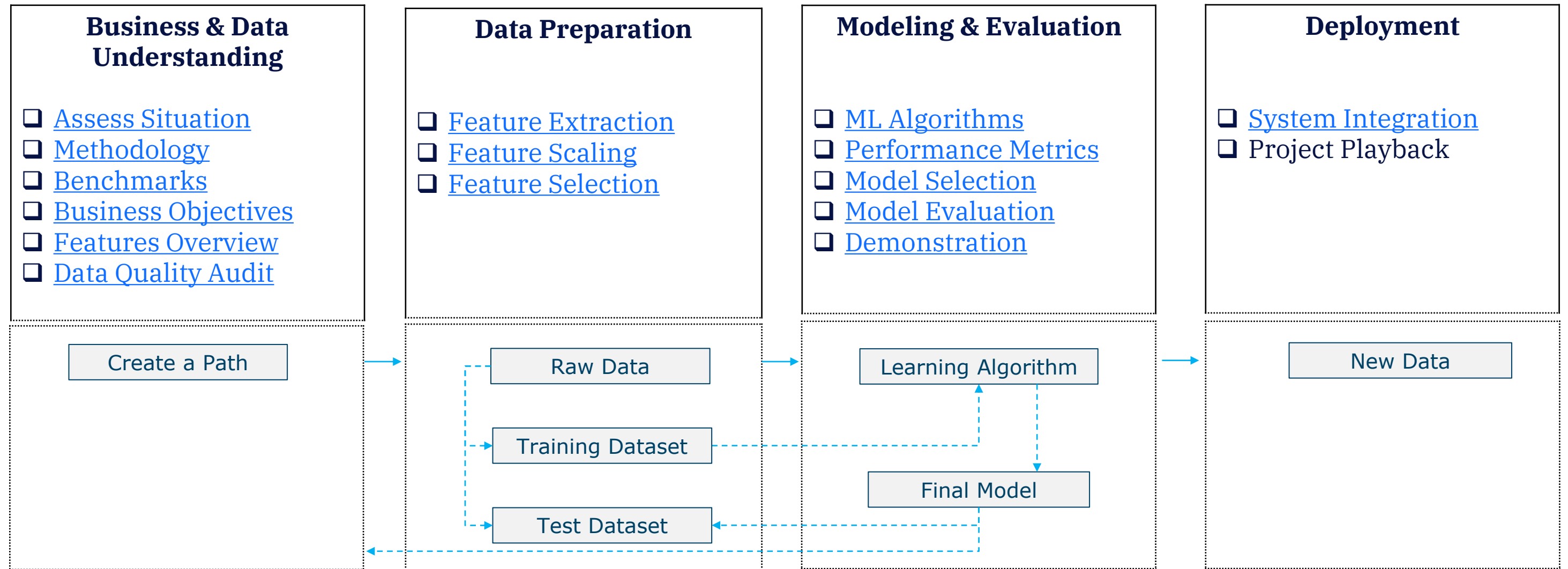


Choosing the right analytic approach



Framework: Roadmap to Building Machine Learning System

"Essentially, all models are wrong, but some are useful."--- Box, George E. P.; Norman R. Draper (1987). Empirical Model-Building and Response Surfaces, p. 424, Wiley. ISBN 0471810339.



Predictive Analytics and Machine Learning Solutions

While there are many tools that could be used for building predictive analytics and machine learning solutions (see below for examples), this workshop will be based on IBM Watson Studio for the experiential learning session.

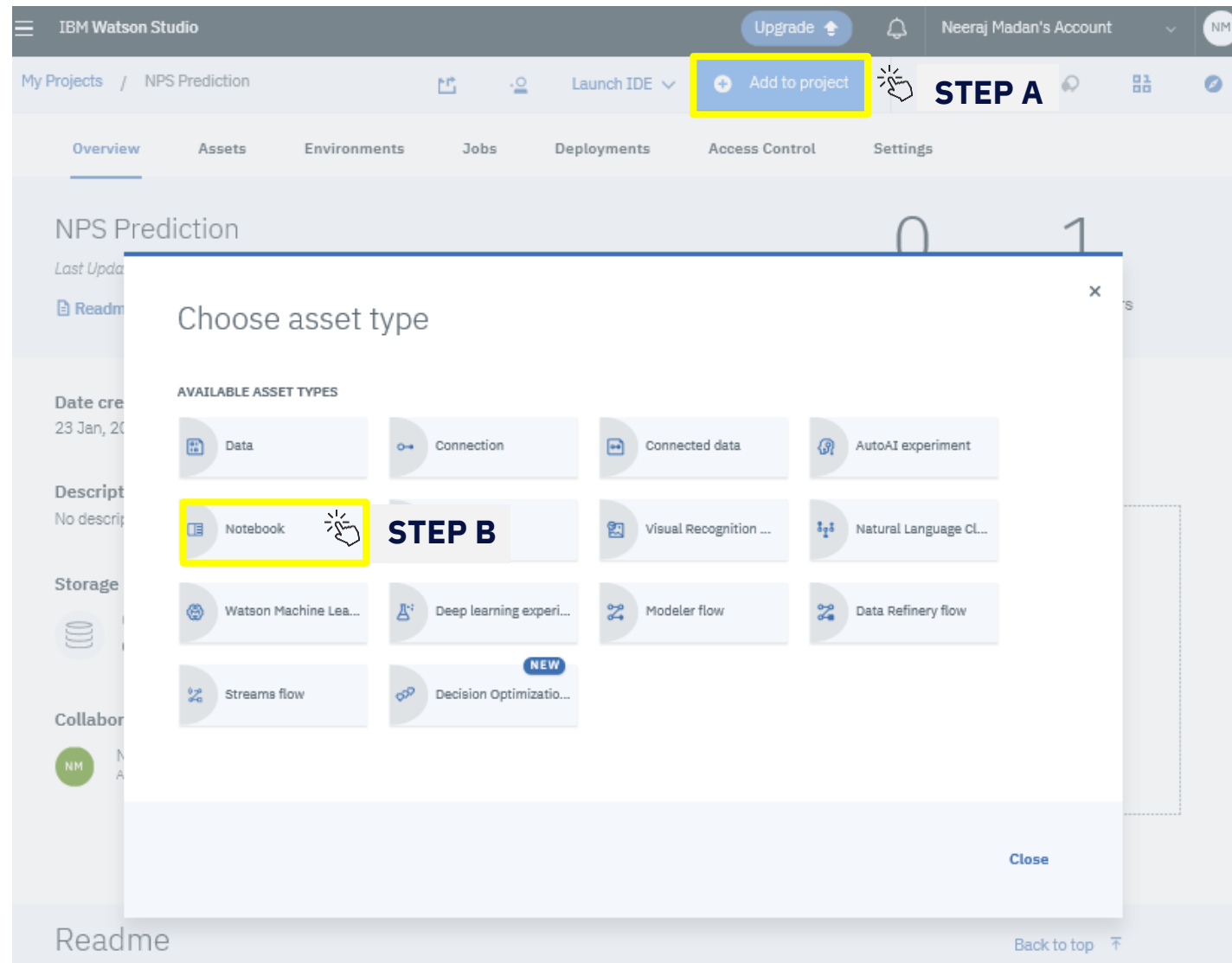
To name a few:

1. IBM Watson Studio
2. SAS Advanced Analytics
3. RapidMiner
4. Amazon SageMaker
5. Azure Machine Learning Studio (Microsoft)
6. Google Cloud AI Platform

Audience Poll

Create a Project

Add a notebook



Create a Project

Import a notebook from GitHub

The screenshot shows the 'New notebook' form in IBM Watson Studio. The form is divided into two main sections: 'Name' and 'Description (optional)' on the left, and 'Select runtime' and 'Notebook URL' on the right. Annotations with yellow boxes and lightbulb icons indicate the following steps:

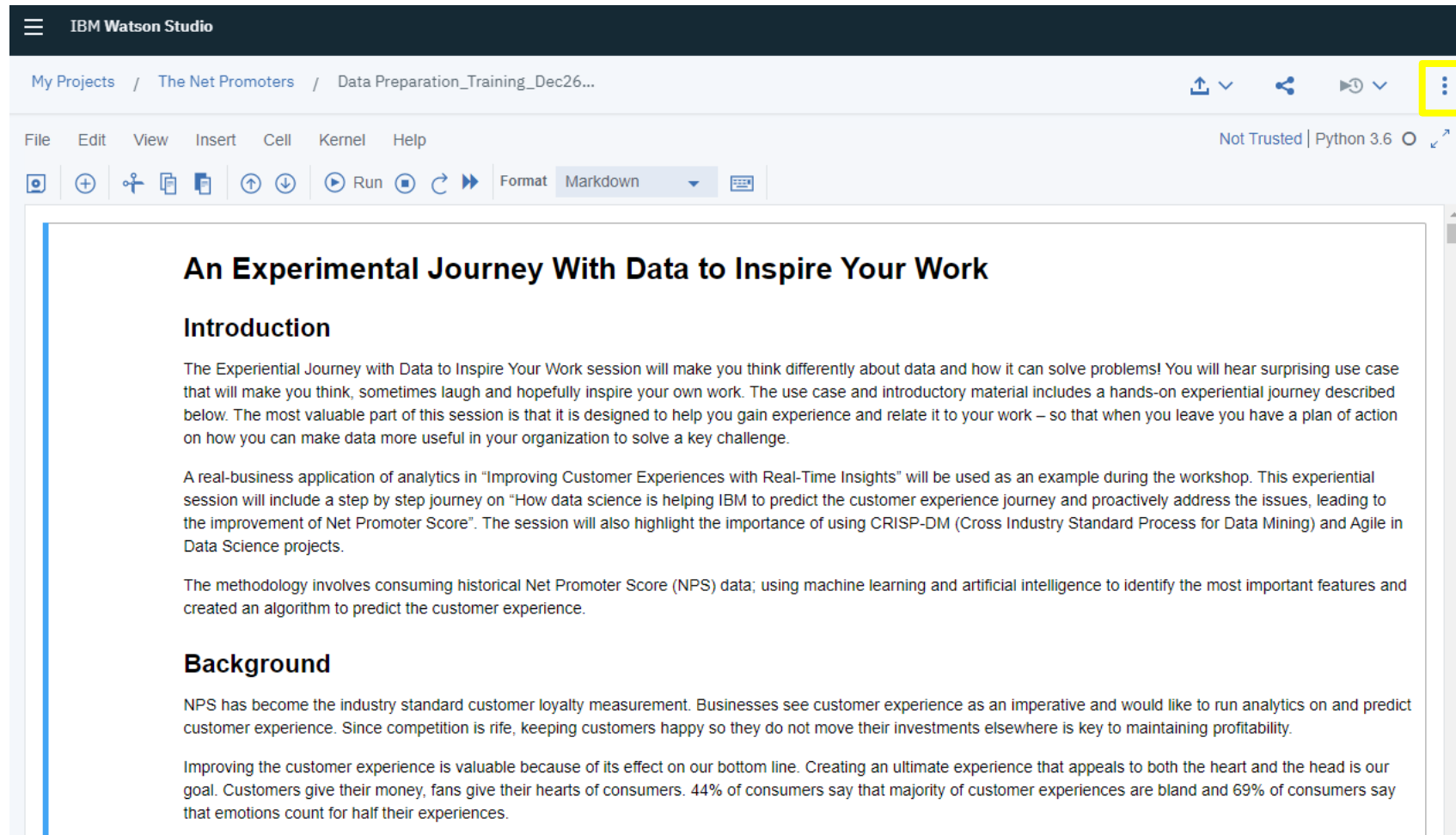
- STEP A:** The 'From URL' tab is selected in the 'New notebook' section.
- STEP B:** The 'Name' field contains the text '(Please type "NPS Prediction")'.
- STEP C:** The 'Select runtime' dropdown is set to 'Default Python 3.8 XS (2 vCPU 8 GB RAM)'.
- STEP D:** The 'Notebook URL' field contains the text 'Use the link highlighted below'.

The 'Name' field has a character count of '26 characters remaining'. The 'Description (optional)' field has a character count of '500 characters remaining'. The 'Select runtime' section includes a note: 'The selected runtime has 1 vCPU and 4 GB RAM and is free. [Learn more](#) about capacity unit hours and Watson Studio pricing plans.'

Notebook URL: <https://github.com/neemadan/An-Experiential-Journey-With-Data-to-Inspire-Your-Work/blob/master/An%20Experimental%20Journey%20With%20Data%20to%20Inspire%20Your%20Work.ipynb>

Create a Project

Let's get started!



The screenshot displays the IBM Watson Studio web interface. At the top, a dark blue header bar contains the IBM Watson Studio logo. Below this, a breadcrumb trail shows the navigation path: 'My Projects' / 'The Net Promoters' / 'Data Preparation_Training_Dec26...'. To the right of the breadcrumb, there are icons for uploading, sharing, and a menu (three vertical dots, highlighted with a yellow box). Below the header, a light blue toolbar contains various icons for file operations (new, open, save, etc.) and a 'Run' button. The main content area is a document editor with a title bar that says 'Not Trusted | Python 3.6'. The document content is titled 'An Experimental Journey With Data to Inspire Your Work' and includes an 'Introduction' section. The introduction text describes the session's purpose and the real-business application of analytics. A 'Background' section follows, discussing the importance of customer experience and the use of Net Promoter Score (NPS) data.

An Experimental Journey With Data to Inspire Your Work

Introduction

The Experiential Journey with Data to Inspire Your Work session will make you think differently about data and how it can solve problems! You will hear surprising use case that will make you think, sometimes laugh and hopefully inspire your own work. The use case and introductory material includes a hands-on experiential journey described below. The most valuable part of this session is that it is designed to help you gain experience and relate it to your work – so that when you leave you have a plan of action on how you can make data more useful in your organization to solve a key challenge.

A real-business application of analytics in “Improving Customer Experiences with Real-Time Insights” will be used as an example during the workshop. This experiential session will include a step by step journey on “How data science is helping IBM to predict the customer experience journey and proactively address the issues, leading to the improvement of Net Promoter Score”. The session will also highlight the importance of using CRISP-DM (Cross Industry Standard Process for Data Mining) and Agile in Data Science projects.

The methodology involves consuming historical Net Promoter Score (NPS) data; using machine learning and artificial intelligence to identify the most important features and created an algorithm to predict the customer experience.

Background

NPS has become the industry standard customer loyalty measurement. Businesses see customer experience as an imperative and would like to run analytics on and predict customer experience. Since competition is rife, keeping customers happy so they do not move their investments elsewhere is key to maintaining profitability.

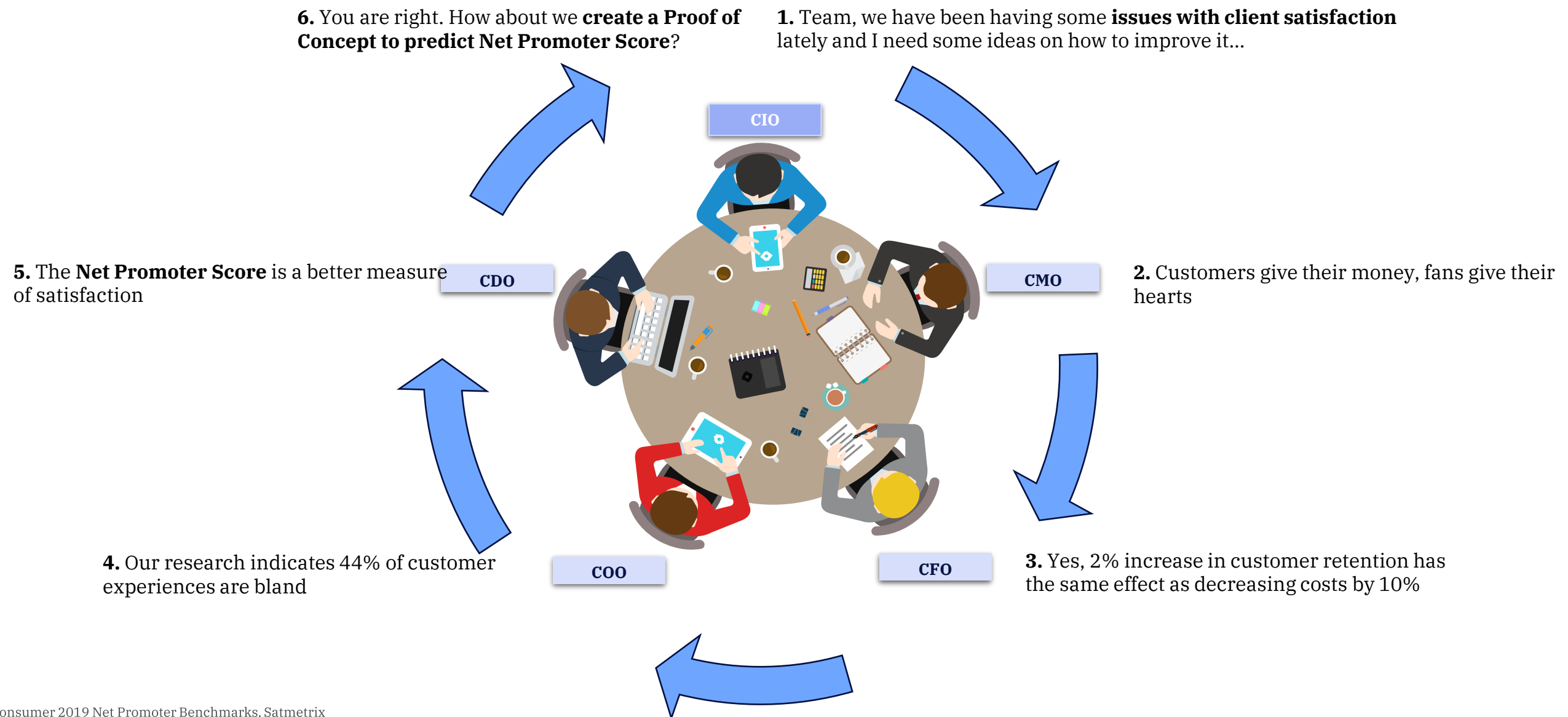
Improving the customer experience is valuable because of its effect on our bottom line. Creating an ultimate experience that appeals to both the heart and the head is our goal. Customers give their money, fans give their hearts of consumers. 44% of consumers say that majority of customer experiences are bland and 69% of consumers say that emotions count for half their experiences.

Take away

Now, I am able to

- ✓ Create/ setup the Data Science environment on IBM Cloud
- ✓ Learn the Roadmap to Building a Machine Learning System

Behind the Scenes: Let us talk about Improving Customer Experience with Real-Time Insights



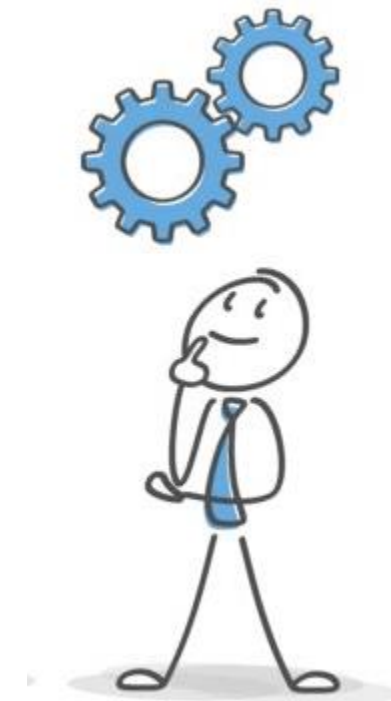
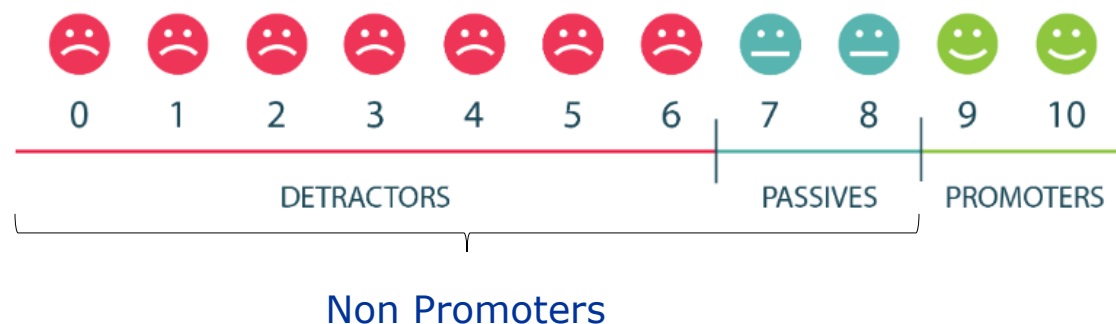
Source: U.S. Consumer 2019 Net Promoter Benchmarks, Satmetrix

Assess Situation

In year 2020, **The Company** world-wide **supported 500,000 cases** which were created in multiple platforms.

The **Net Promoter Survey (NPS)** response rate was **15%**.

60% cases were non-promoters and 40% were promoters.



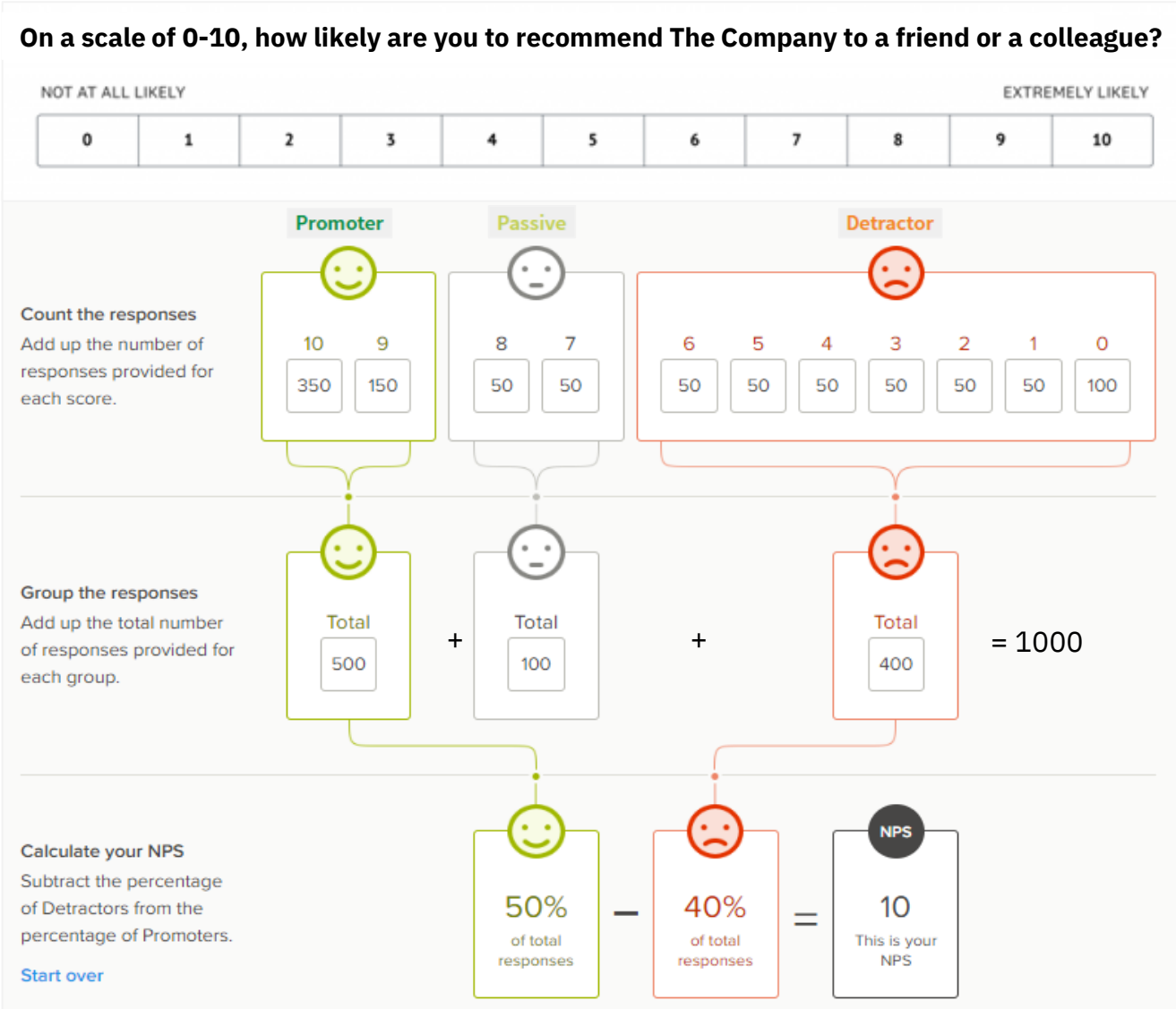
Note: The numbers highlighted above are crafted for this workshop.

Methodology

Net Promoter has become the industry standard customer loyalty measurement. Businesses see customer experience as an imperative.

On a scale of 0-10, how likely would you recommend [brand/ support] to a friend or colleague?

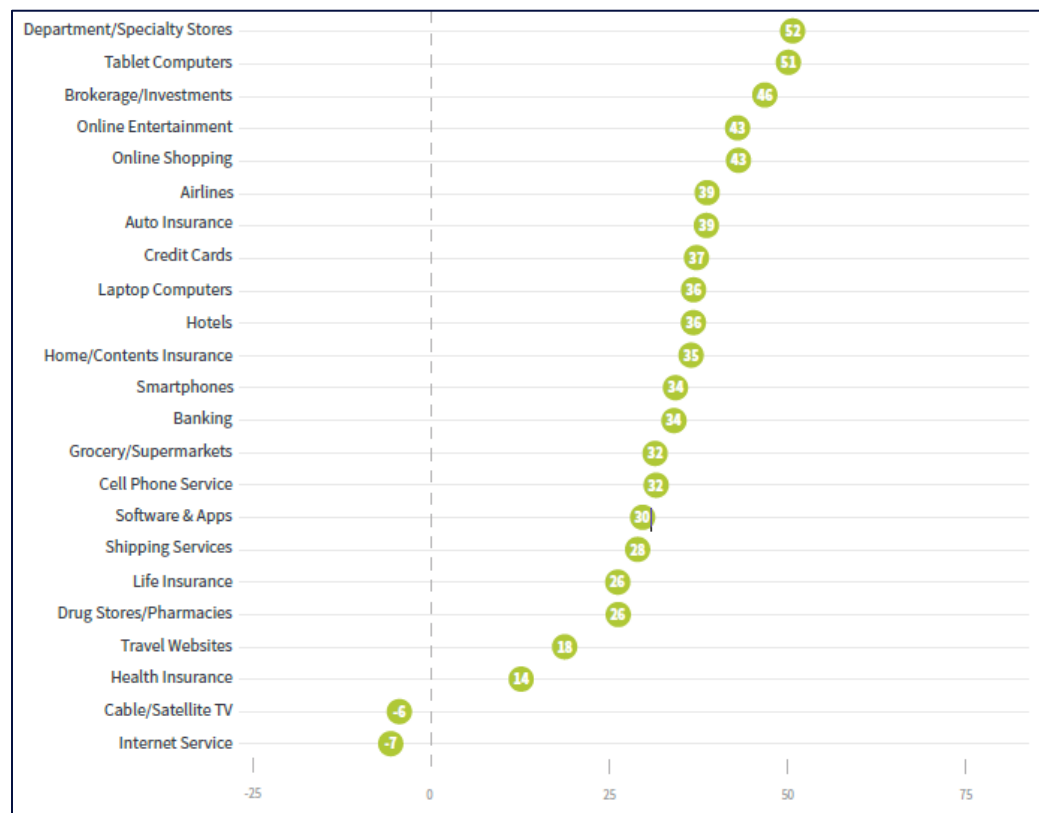
Calculating NPS score is as simple as tallying up your responses and subtracting the percentage of detractors from the percentage of promoters. The score is a whole number that ranges from -100 to 100, and indicates customer happiness with our brand experience.



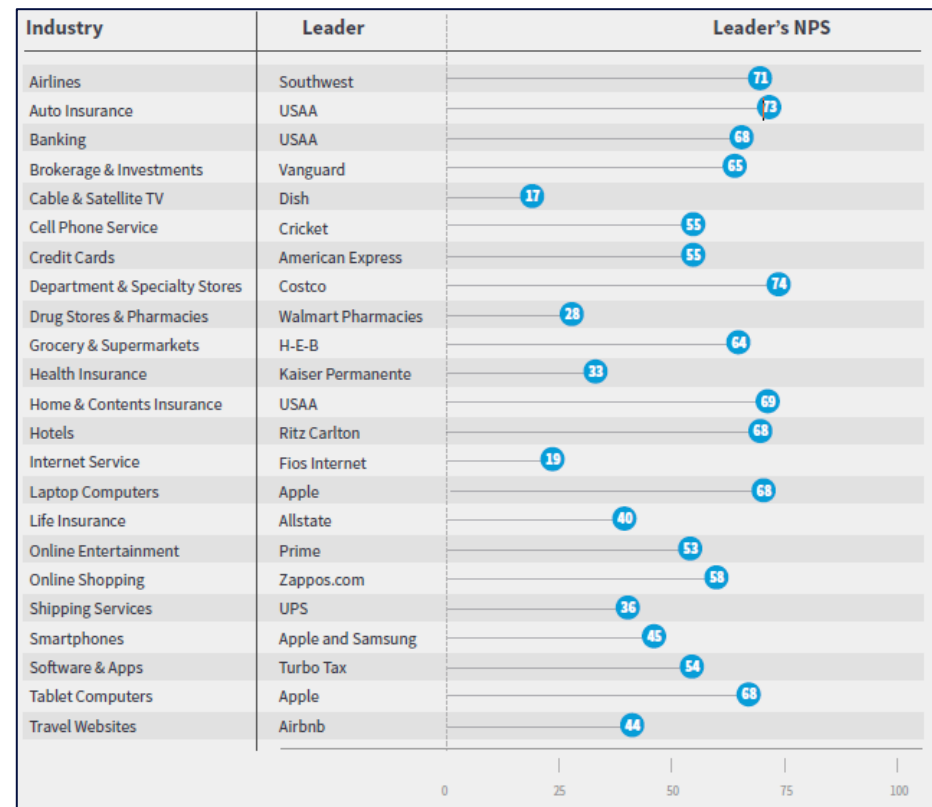
Source: How to Use Net Promoter to Drive Business Growth, Satmetrix and [NPS Calculator](#)

Benchmarks (Average NPS by Industry and Leaders)

Net Promoter Scores vary widely by industry, as you can see from the average scores for 23 industries.



Knowing what similar companies have achieved helps us to set realistic goals for improvement, and realism is key to the long-term success of your program.



Source: U.S. Consumer 2019 Net Promoter Benchmarks, Satmetrix



Business Objectives

Goal: Improve the Net Promoter Score by identifying potential non promoters ahead of time and proactively address customer issues

Approach: Consume historical NPS data; Use machine learning and artificial intelligence to identify the most important features and select an algorithm to predict the non promoters

Desired Result: Create a capability to share the top candidates for non-promoter surveys with The Company to proactively address customer issues.





Exercise

1. Identify a data science opportunity in your business context and document.

Consider using this template...

As a <role>, I would like to <direction> the <target variable> for <scope> by <amount> in <timeframe>.

Role = End User

Direction = improve/reduce or increase/decrease

Target Variable = fraud, risk, customer satisfaction, volume, effort, price, cost, availability, productivity, revenue, etc.

Scope = section of the business of interest

Amount = value or percent

Timeframe = weeks, months, years



Take-away

Now, I am able to

- ✓ Setup the Data Science environment on IBM Cloud
- ✓ Learn the Roadmap to build a Machine Learning System
- ✓ Assess the situation, understand the methodology, identify the benchmarks, and define business objectives

Features Overview

Time Emotions

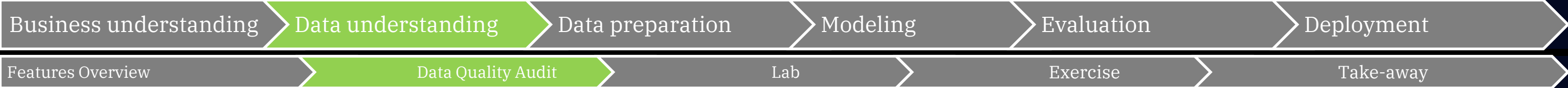
Geography

Money

Sentiment &

<u>TIME</u>	<u>LOCATION</u>	<u>MONEY</u>	<u>SENTIMENT & EMOTIONS</u>
Day of Week	Country	Lifetime Spent	Sentiment
Time Window (Prime or Non-Prime)	Geography	Monthly Recurring Spend	Emotion (Anger, Disgust, Fear, Joy, and Sadness)
Age of Account	Region		
Meaningful Update			

Other Features: Assignment Count, Support Plan, Account Type, Severity, Technology, Case Origination Source, Case Origination User Type, Tribe, Catalog



Data Quality Audit

Pandas-Profiling Report (Things to check)

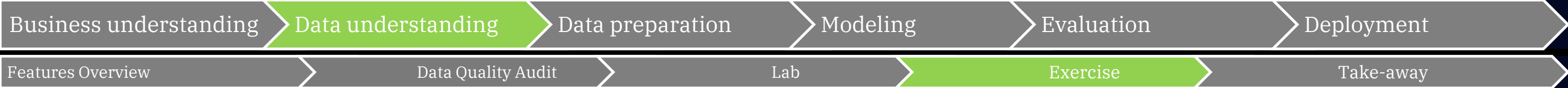
- ☐ Number of observations, features and type
- ☐ Large Number of Distinct Values (High Cardinality)
- ☐ Correlation
- ☐ Missing values
- ☐ End goal is to review the data and improve Data Quality



Lab: Instructions

Run the following section in the notebook.

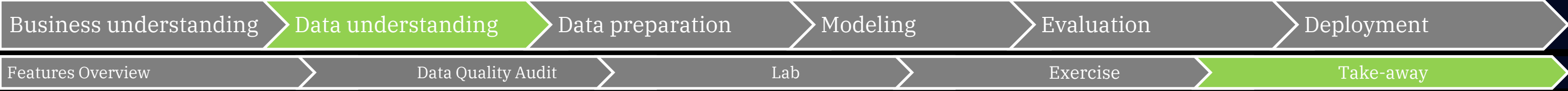
1. Introduction to Notebooks
2. Load packages and verify the version
3. Data Exploration
 - Load and read the files from GitHub
 - Explore the data and perform quality audit



Exercise

2. What data set would you gather to work on the problem statement?

Be sure to include the feature that you believe would influence the outcome, the definition, type, and range (e.g., weather, forecasted rain probability at 10 am, percent, 0.0 to 1.0) of that feature, and the data source that would provide it (weather.com).



Take away

Now, I am able to

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- ✓ Assess the situation, understand the methodology, identify the benchmarks, and define business objectives
- ✓ Introduction to Notebook, load packages, and verify the versions
- ✓ Explore the data set and perform quality audit



Extract, Scale, and Select Features (1 of 2)

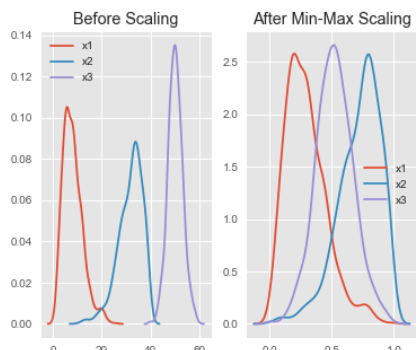
Feature Extraction

One Hot Encoding



- Other Techniques:
- OrdinalEncoding
 - LabelEncoder
 - BinaryEncoder
 - Hashing Encoder
 - Target/Mean Encoding
 - Autoencoders

Feature Scaling



- Other Techniques:
- StandardScaler
 - RobustScaler
 - Normalizer

Feature Selection

Dimensionality Reduction



- Other Techniques:
- Pairwise Correlation
 - Multi- collinearity
 - Principal Component Analysis,
 - Cluster Analysis, Correlation (with the target)
 - Forward/ Backward/ Stepwise selection
 - LASSO
 - Tree-based selection



Extract, Scale, and Select Features (2 of 3)

Why (a, b, and c) and (x & y) are commonly used as mathematical placeholders?

The letters at the end of the alphabet, viz., x, y, z, etc. are to denote unknown variables, while those at the start of the alphabet, a, b, c, etc. denote constants, was first highlighted in *La Géométrie* in year 1637.



Extract, Scale, and Select Features (3 of 3)

Why the feature variable is denoted by a upper-case 'X' whereas the target variable is a lower-case 'y'?

In terms of Linear Algebra, it is extremely common to use capital Latin letters for matrices (e.g. design matrix X) and lowercase Latin letters for vectors (response vector y).

A vector is a matrix with one row or one column. A matrix is a group of numbers(elements) that are arranged in rows and columns.



Lab: Instructions

4. Feature Extraction

5. Feature Scaling

6. Feature Selection



Exercise

3. How would you prepare the dataset and what challenges do you foresee?

Now, as you have shortlisted the input features to prepare a data model, document the transformation steps (extract, scale, and select) you would apply on features to prepare the data-set.

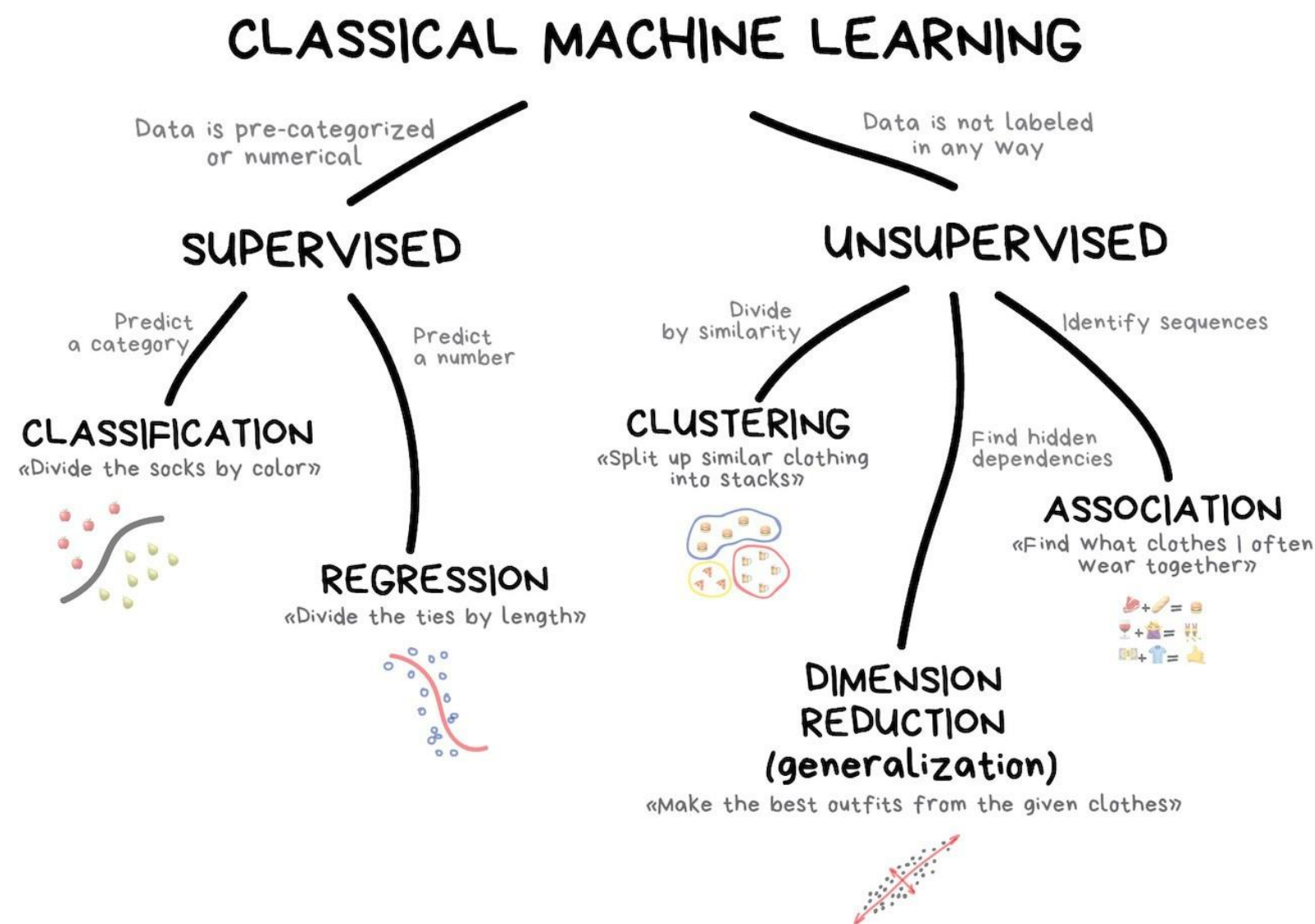


Take away

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- ✓ Setup the Data Science environment on IBM Cloud
- ✓ Learn the Roadmap to build a Machine Learning System
- ✓ Assess the situation, understand the methodology, identify the benchmarks, and define business objectives
- ✓ Introduction to Notebook, load packages and verify the versions
- ✓ Explore the data set and perform quality audit
- ✓ Extract, scale, and select features for the data model

Machine Learning Algorithm Selection (1 of 2)



Source: [Machine Learning for Everyone](#)



Machine Learning Algorithm Selection (2 of 2)

Supervised Learning		Unsupervised Learning		
Classification	Regression	Clustering	Dimensionality Reduction	Association

1. Spam filtering
2. Fraud detection
3. Customer Segmentation
4. Stock price forecasts
5. House Price
6. Topic modeling and similar document search
7. To place the products on the shelves

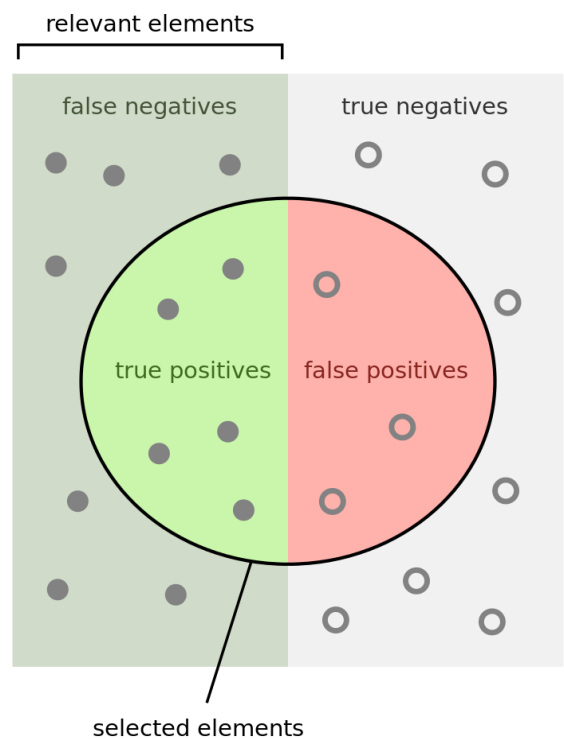




Machine Learning Algorithm Selection (2 of 2)

Supervised Learning		Unsupervised Learning		
Classification	Regression	Clustering	Dimensionality Reduction	Association
Spam filtering Fraud detection	Stock price forecasts House Price	Customer Segmentation	Topic modeling and similar document search	To place the products on the shelves

Measure Model Performance

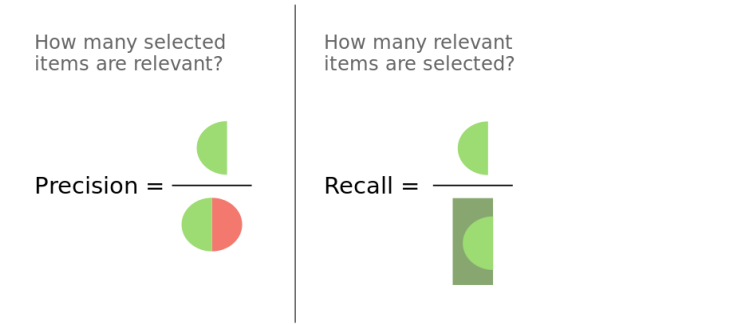


	Predicted:		
	NO	YES	
Actual: NO	TN = 50	FP = 10	60
Actual: YES	FN = 5	TP = 100	105
	55	110	

What is Confusion Matrix?

Describes the performance of a classification model on a set of test data for which true values are known.

- Accuracy:** Overall, how often is the classifier correct?
 $(TP+TN)/total = (100+50)/165 = 0.91$
- Precision:** When it predicts yes, how often is it correct?
 $(TP/predicted\ yes = 100/110 = 0.91)$
- Recall:** When it's actually yes, how often does it predict yes?
 $(TP/actual\ yes = 100/105 = 0.95)$
- F1:** harmonic mean of precision and recall: $(2 * precision * recall) / (precision + recall)$



Source: [Simple guide to confusion matrix terminology](#), [Accuracy Paradox](#)



Model Evaluation

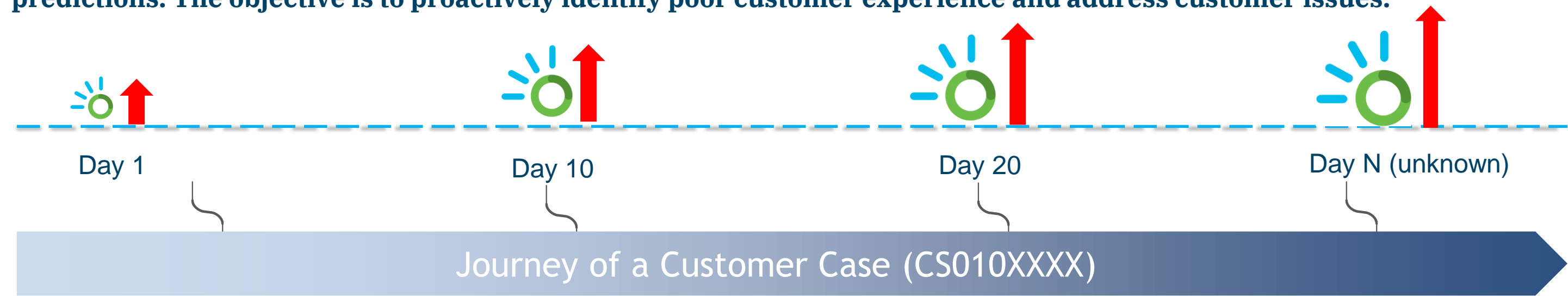
	Metrics					
Model Name	Accuracy		Precision		Recall	
LogisticRegression	↑	65%		63%		59%
SGDClassifier	→	55%		59%		58%
SVM	↗	60%		59%		60%
KNeighborsClassifier	↗	56%		53%		53%
GaussianProcessClassifier	↑	62%		57%		55%
MultinomialNB	↑	62%		59%		57%
DecisionTreeClassifier	↗	56%		54%		54%
RandomForestClassifier	↗	60%		55%		54%
GradientBoostingClassifier	↑	67%		64%		62%
VotingClassifier	↑	63%		59%		55%
MLPClassifier	↑	62%		31%		50%



Demonstration



The algorithm consumes multiple signals (time, geography, spend, and sentiments) and gives the non promoter predictions. The objective is to proactively identify poor customer experience and address customer issues.





Lab: Instructions

Run the following section in the notebook.

7. Split data into train and test sets

8. Measure Model Performance

9. Evaluate and Select Model



Exercise

4. What modeling techniques would you attempt and metrics would you use to evaluate the model performance?

Consider the type of output you are generating and choose an appropriate technique for the model. Remember, accuracy, precision, recall, F1 and think about “explainability” of the model results.



Take away

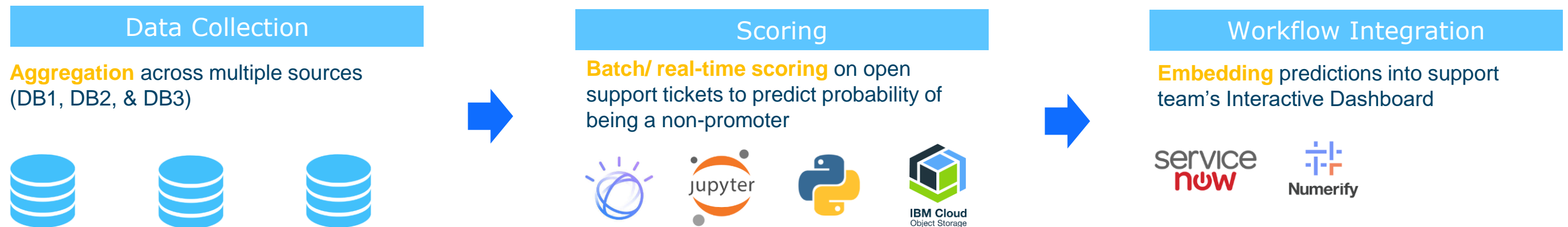
Now, I am able to

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- ✓ Introduction to Notebook, load packages and verify the versions
- ✓ Explore the data set and perform quality audit
- ✓ Extract, scale, and select features for the data model
- ✓ Split data into train & test sets, select model, evaluate performance metrics, and demonstrate

Proposed Solution

The model developed as a part of the hack uses artificial intelligence and machine learning to predict Non-Promoters on historical data pattern. Key aspects of approach include,

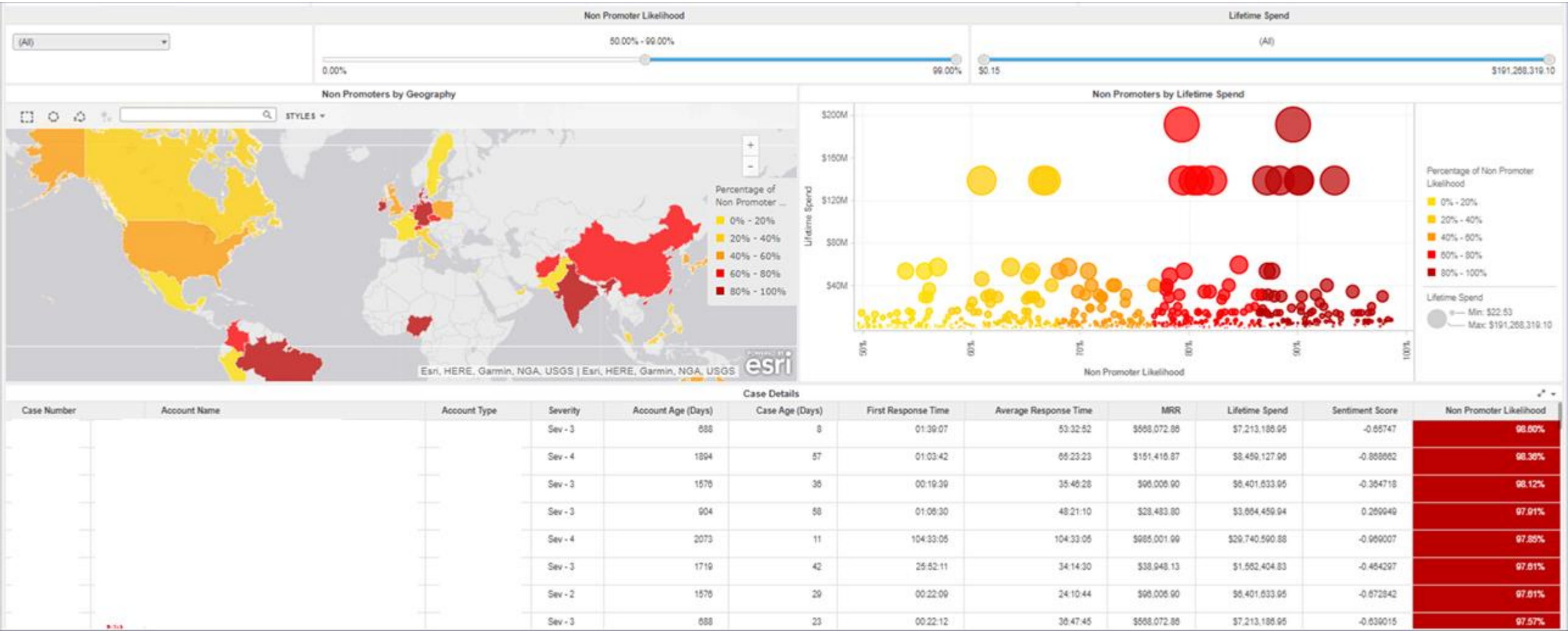
1. Watson NLP to create additional features from customer conversation logs
2. Machine Learning algorithms to produce the predictions



Solution Integration

“Ideas are easy, Execution is everything.” John Doerr

NPS Predictions Board



Note: This visual highlighted above is crafted for workshop purpose.



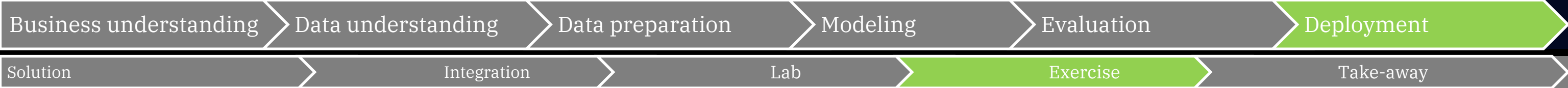
Lab: Instructions

Run the following section in the notebook.

10. Save the Model

11. Deploy the Model

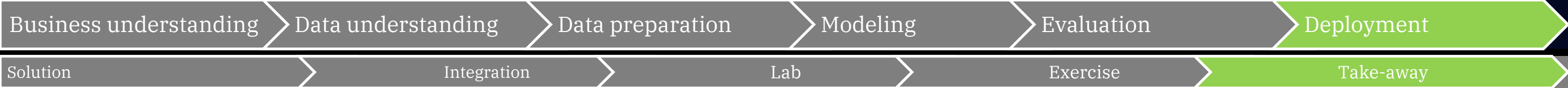
12. Predict Cases



Exercise

5. How do you plan to consume the outputs of model?

Consider dashboards, reports, visualizations or automated decisions. Describe how the end user interacts with results.

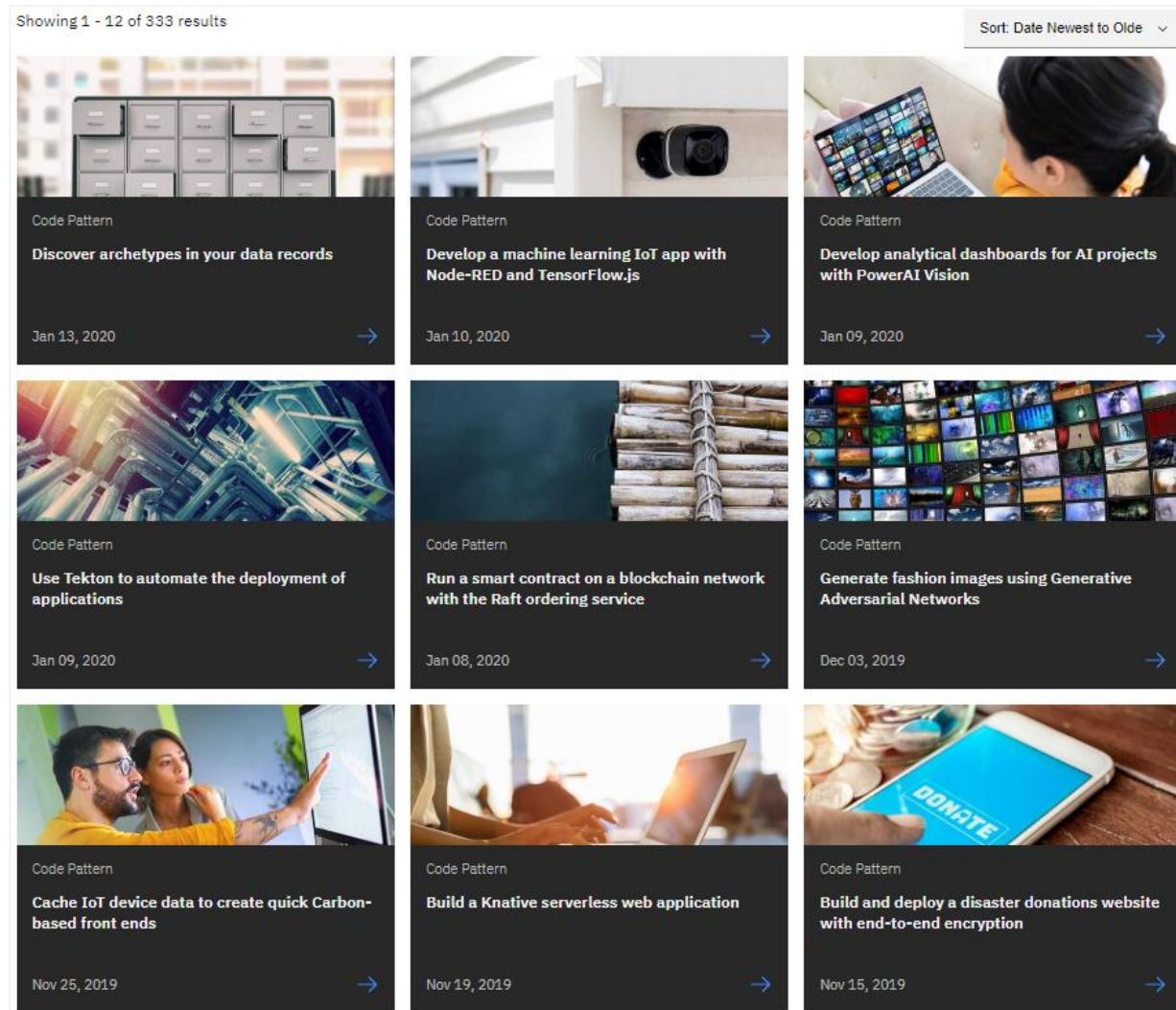


Take away

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- ✓ Explore the data set and perform quality audit
- ✓ Extract, scale, and select features for the data model
- ✓ Split data into train & test sets, select model, evaluate performance metrics, and demonstrate
- ✓ Generate ideas on how to consume the predictions and integrate solution in business systems

IBM Code Patterns



Link: <https://developer.ibm.com/patterns/>

Contact Us

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Thank you



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