

A Day in the Life of a Data Scientist: How do we train our teams to get started with AI?

Francesca Lazzeri & Jaya Mathew

 @frlazzeri

 @mathew_jaya

Strata
DATA CONFERENCE



Agenda

- What is Artificial Intelligence (AI)?
- Why AI?
- How to get started with AI
- Understanding the ML workflow
- Suggested tools for AI development
- AI usage in marketing
- Fraud management use case

What is AI?

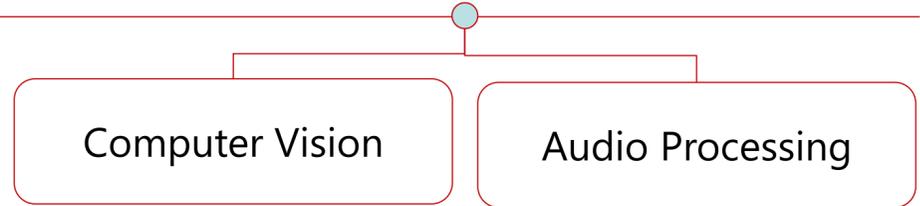
What is AI? – To sense, comprehend and act

AI Technologies
Illustrative Solutions

Sense



Computer vision and audio processing, for example are able to actively perceive the world around them by acquiring and processing images, sounds and speech. The use of facial recognition at border control kiosks is one practical example of how it can improve productivity.




Virtual Agents



Identity Analytics

Comprehend



Natural language processing and inference engines can enable AI systems to analyse and understand the information collected. This technology is used to power the language translation feature of search engine results

Emerging AI technologies




Cognitive Robotics

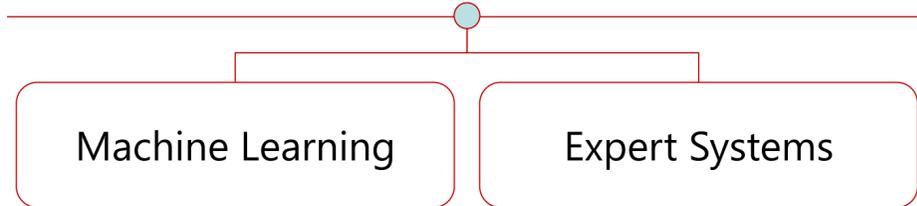


Speech Analytics

Act



An AI system can take action through technologies such as expert systems and inference engines or undertake actions in the physical world. Auto-pilot features and assisted-braking capabilities in cars are examples of this




Recommendation Systems

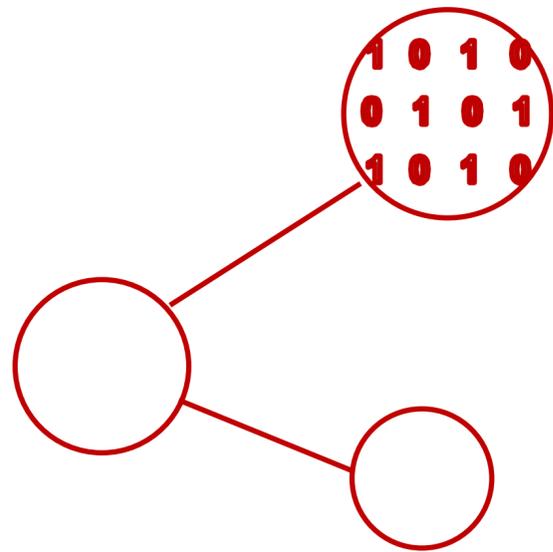


Data Visualization

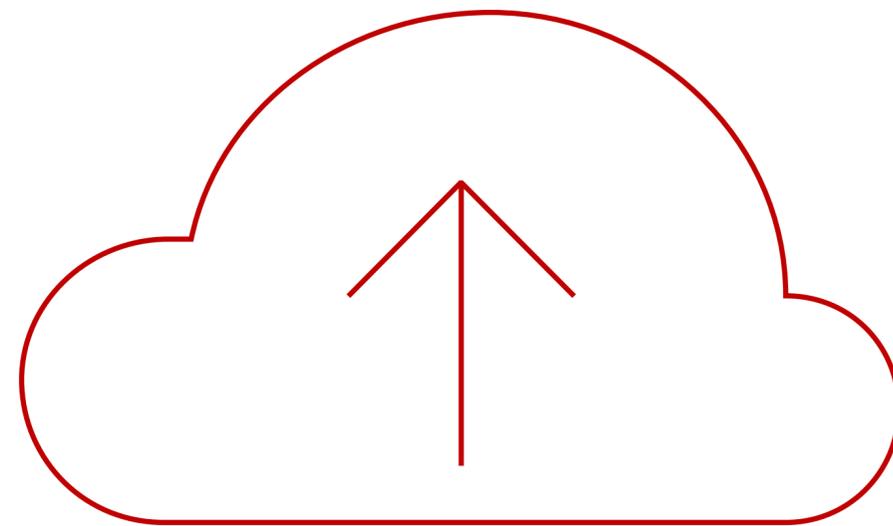
Source: [Accenture: Why artificial intelligence is the future of growth](#), April 2016

Why AI?

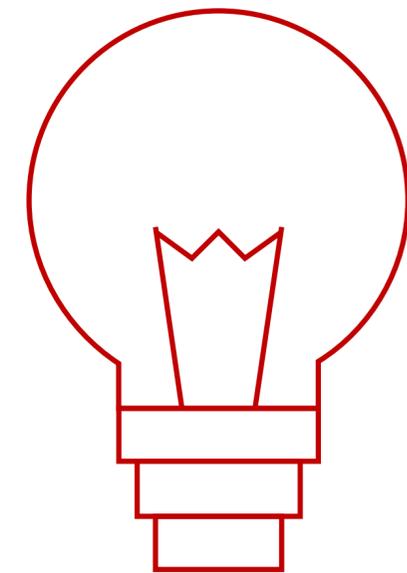
Three major trends are converging



Data

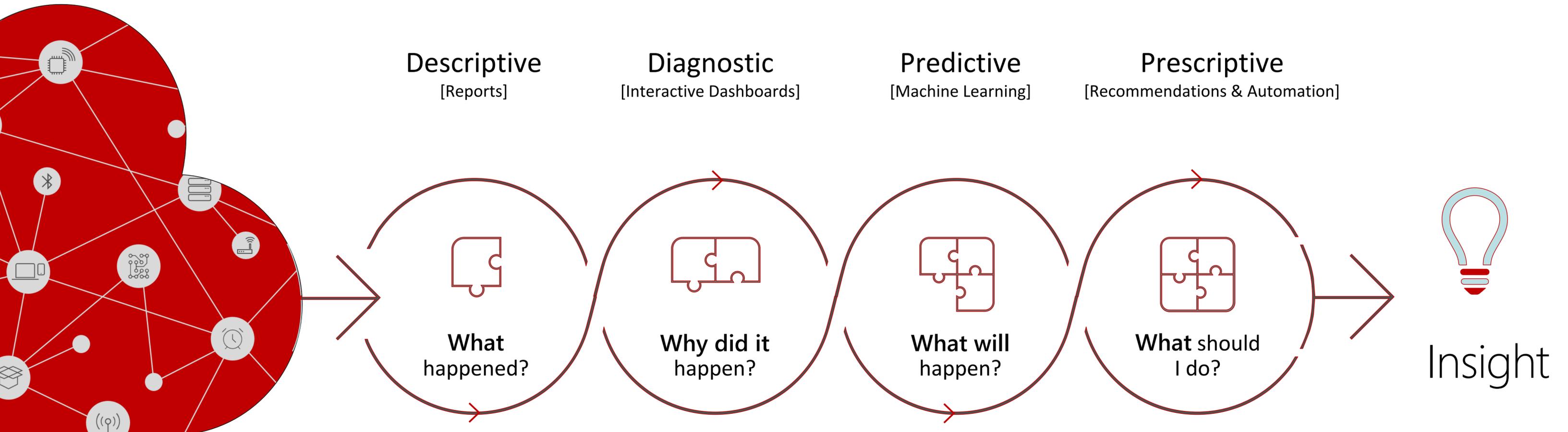


Cloud

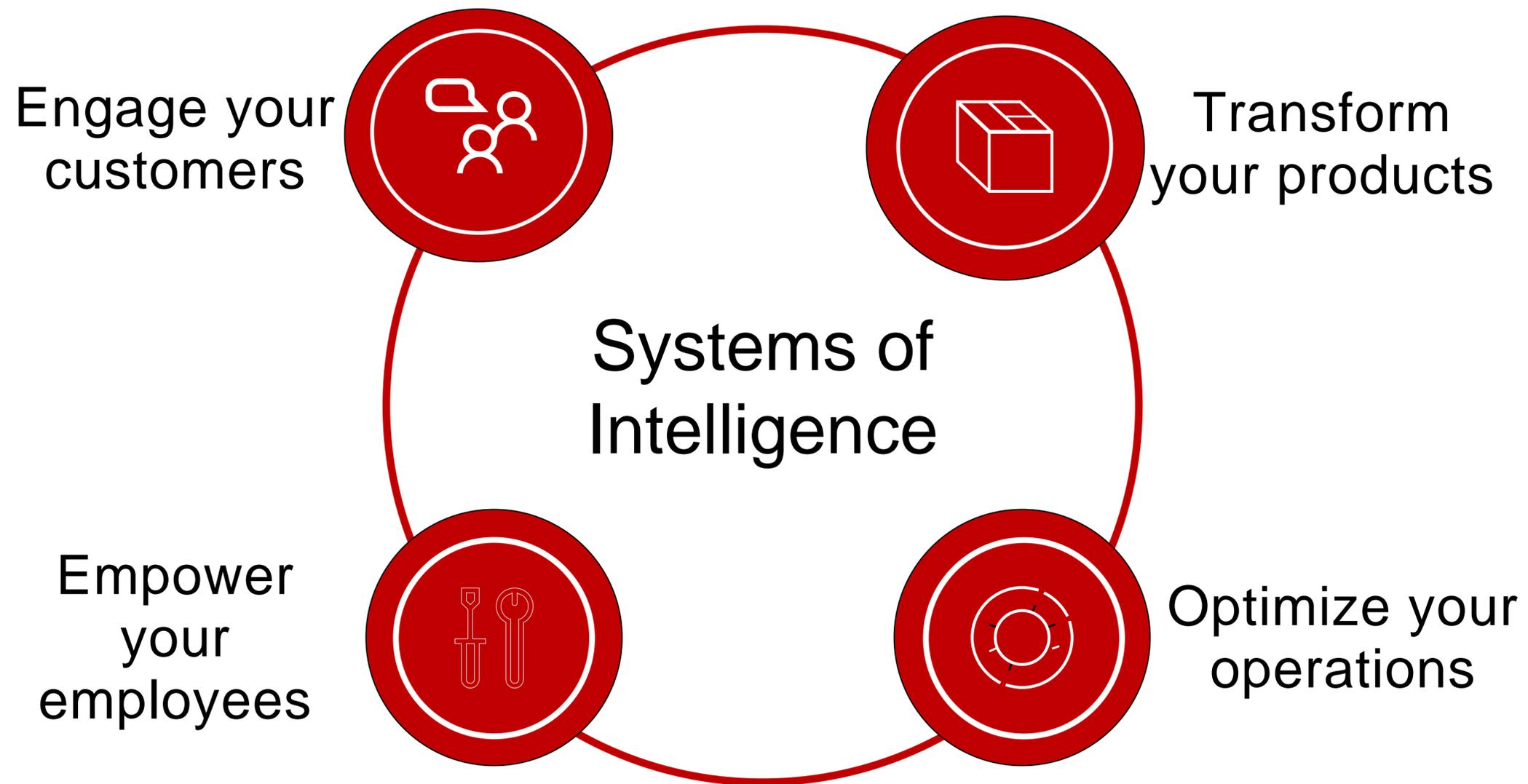


Intelligence

From data to decisions and actions



Digital transformation is driving new business value



How to get started with AI

I want to use AI – How can I get started?

aka.ms/AICognitiveServices

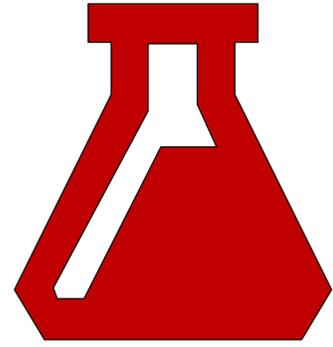
aka.ms/AICustomModels



Cognitive Services



Custom Services
Bring your Own Data



Custom AI Models



Flexibility

Being Obsessed with Data

Question
is sharp

E.g. Predict
whether
component X will
fail in the next Y
days

Data
measures
what you
care about

E.g. Identifiers at the
level you are
predicting, relevant
data collected &
feature engineering
using domain
knowledge

Data is
accurate

E.g. Failures are
really failures,
human labels on root
causes

Data is
connected

E.g. Machine
information linkable
to usage information

A lot of
data

E.g. Will be difficult to
predict failure
accurately with few
examples

Asking the right questions

Business scenario	Key decision	Data Science question
Energy forecasting	Should I buy or sell energy contracts?	What will be the long/short-term demand for energy in a region?
Customer churn	Which customers should I prioritize to reduce churn?	What is probability of churn within X days for each customer?
Personalized marketing	What product should I offer first?	What is the probability that customer will purchase each product?
Product feedback	Which service/product needs attention?	What is social media sentiment for each service/product?

Defining Performance Metrics

Establish a
**Qualitative
Objective**

Translate into
**Quantifiable
Metric**

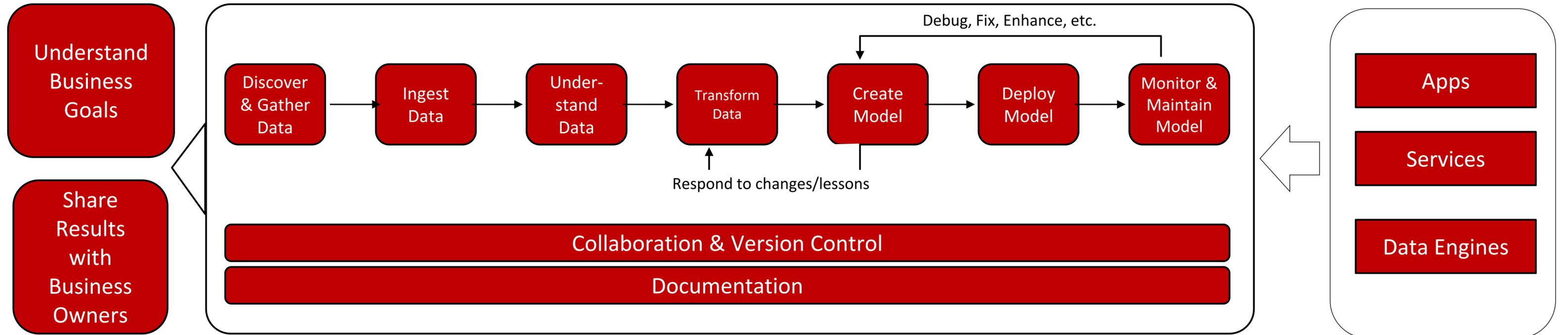
Quantify the
**Metric Value
Improvement**
useful for
customer scenario

Establish a
Baseline

Establish how to
measure the
improvement in
the **Data
Science Metric**

Understanding the ML workflow

Sample ML workflow



Build a model
Science

Publish a model
Operations

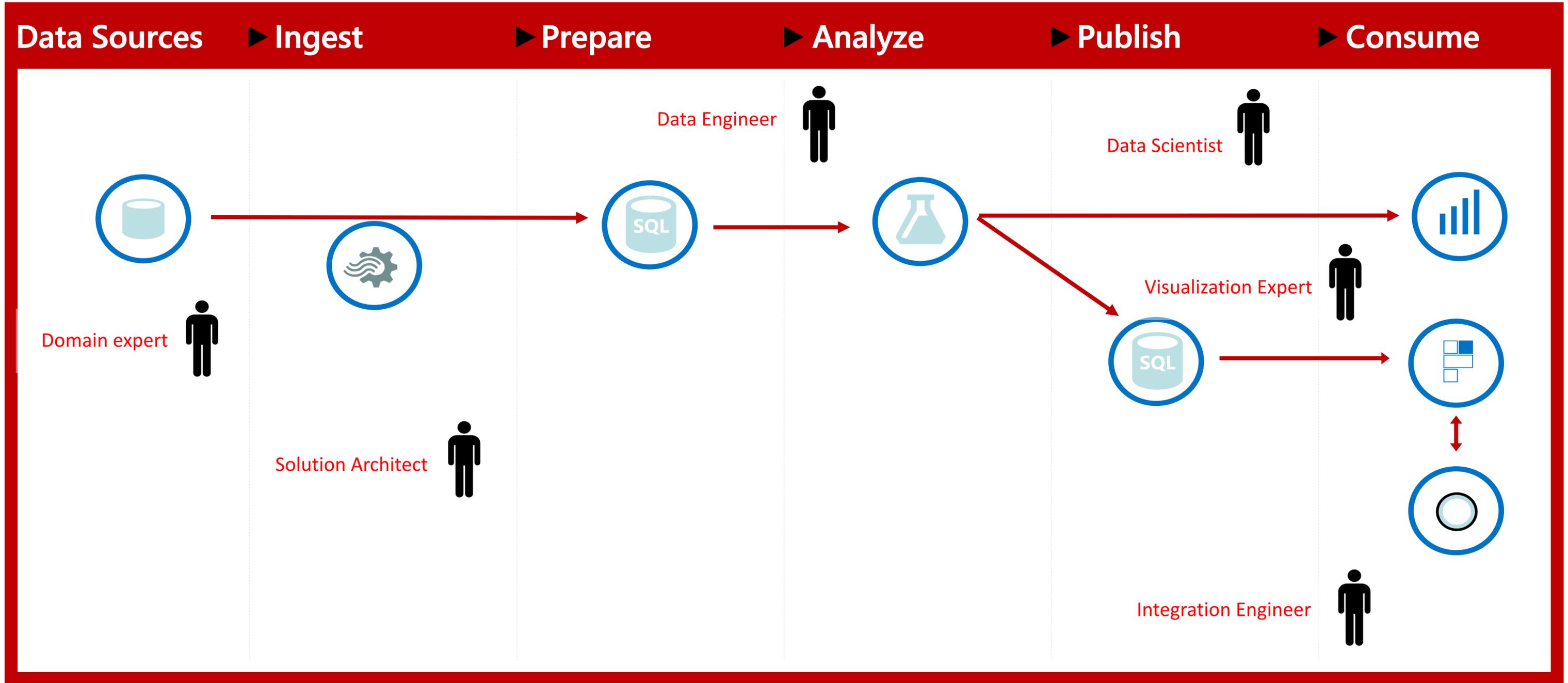
Consume a
model

Team Data Science Process

aka.ms/TeamDataScience

Project Manager 

Executive Sponsorship – IT & Business 



Suggested tools for AI development

Example Tools / Dev Environments

- [Azure Machine Learning Studio](#): Serverless collaborative drag-and-drop tool for graphical machine learning development
- [Azure Machine Learning Services](#): Visual AI powered data wrangling, experimentation, and lifecycle management
- [Visual Studio Code Tools for AI](#): Build, debug, test, and deploy AI with Visual Studio Code on Windows and Mac
- [Azure Notebooks](#): Organize your datasets and Jupyter Notebooks in a centralized library for Data Science and Analysis
- [Deep Learning Virtual Machine](#): A pre-configured environment for deep learning using GPU instances

AI Solution Templates

aka.ms/AzureAIGallery

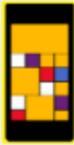
Template Title	Views	Downloads	Time Ago
Loan Credit Risk with SQL Server	1.3K	343	24 days ago
Personalized Offers	4.6K	399	3 months ago
Campaign Optimization with SQL Server	8.1K	1.2K	2 hours ago
Campaign Optimization with Azure HDInsight Spark Clust...	1.2K	157	18 days ago
Predicting Length of Stay in Hospitals	10K	1.2K	25 days ago
Demand Forecasting and Price Optimization	4.4K	670	3 months ago
Quality Assurance	1.9K	313	3 months ago
Telemetry Analytics	9.2K	1.4K	3 months ago
Demand Forecasting	9.5K	1.6K	3 months ago
Predictive Maintenance	11K	2.1K	3 months ago

Microsoft Learning

aka.ms/MicrosoftAILEarning

Free Microsoft training delivered by experts

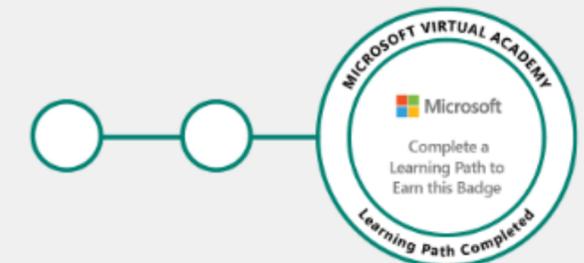
Developers IT Pros Data Pros

 Windows 10	 Cloud Development	 Game Development	 Web Development	 Database Development
 C# / XAML	 Visual Studio	 For Beginners	 Mobile App Development	Browse all developer courses 

Learning Paths

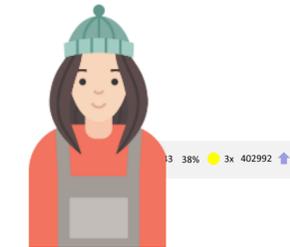
Curated collections of courses to help you build skills. Complete a learning path to earn a badge you can share with others.

[Learn more](#)



AI usage in marketing

How can AI help in Marketing & Personalization?



Input Data

- In store
- Online activity
- Social media
- Past campaign performance

ML & AI

Insight:

- Personalized content for customers
- Identify high value customers

- Learn customer omni-channel preference
- Personalize web/email experience
- Preemptive chatbots to answer customer queries
- Monitor for cart abandonment, churn, retention

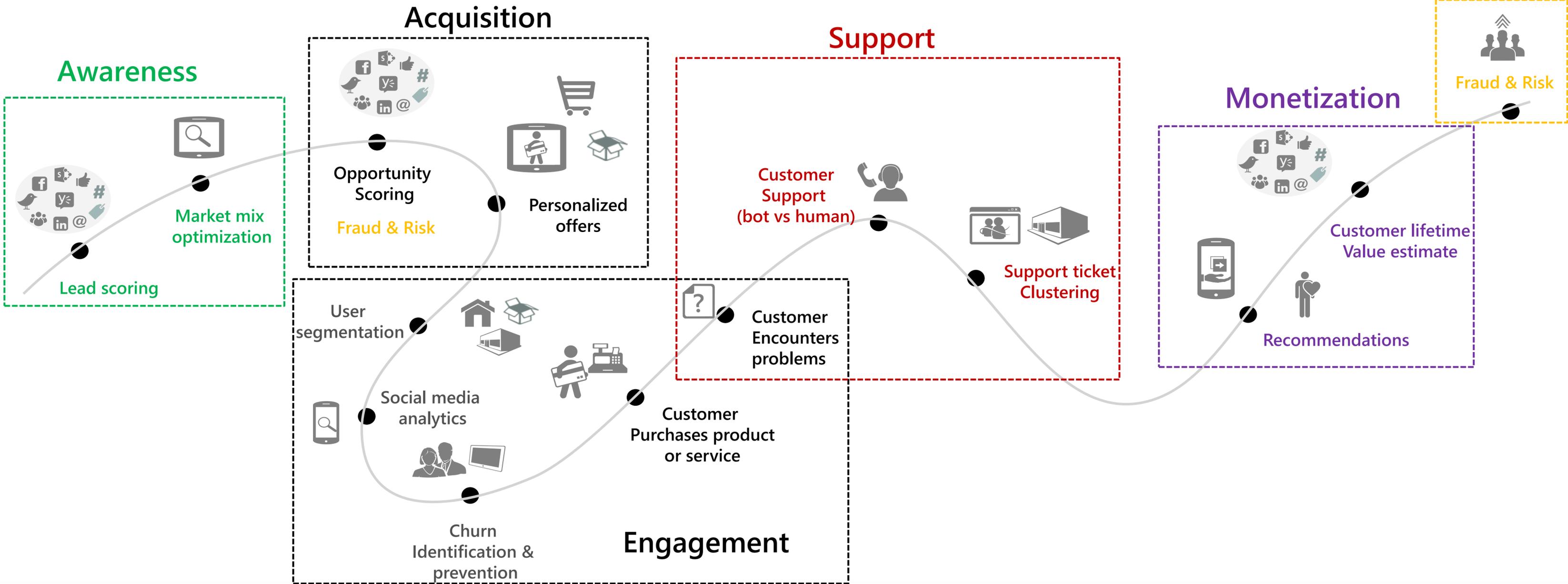
Intelligent Action

- Personalized website experience
- Adaptive product pricing, offers for cross/upsell
- Premium loyalty programs and service experience
- Predictive customer service via social media, chatbots

Marketer's AI Journey Map

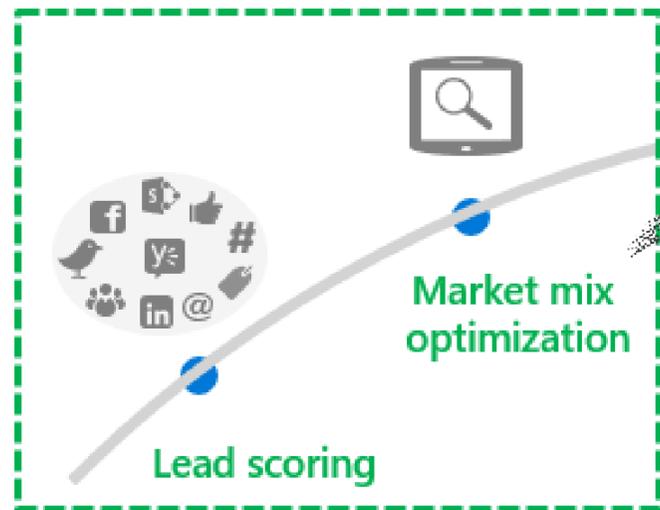
Let us map the journey based on the stages in a customer life cycle as follows:

Awareness; Acquisition; Engagement; Support; Monetization; Fraud/Risk management



Marketing cycle – Awareness & Acquisition

Awareness



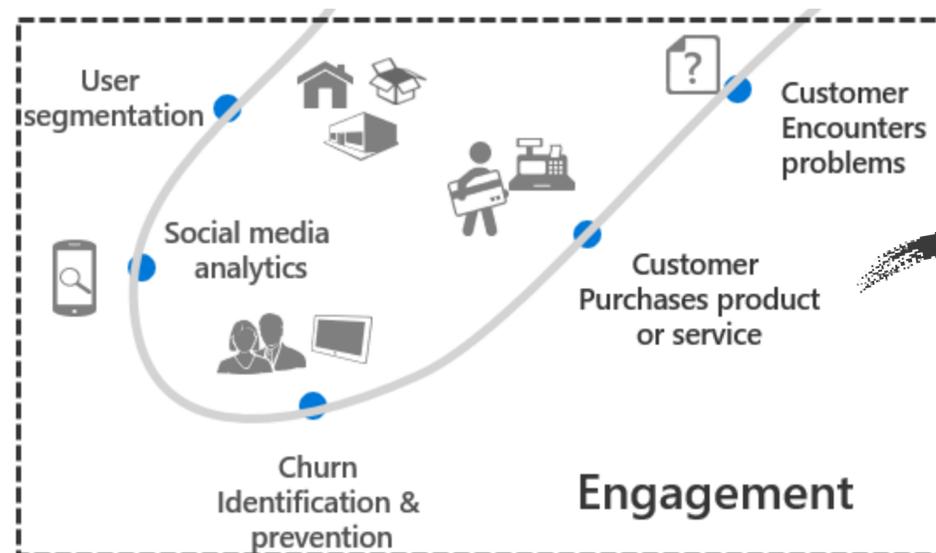
- ✓ Marketing mix optimization for omni-channel budget optimization
- ✓ ML based lead-scoring models

Acquisition

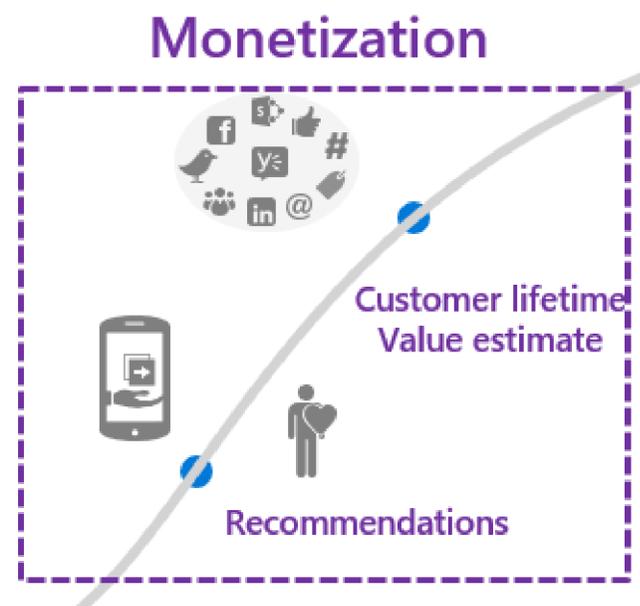


- ✓ Opportunity scoring can help target users who are most likely to make a purchase
- ✓ AI powered content creation

Marketing cycle – Engagement & Monetization



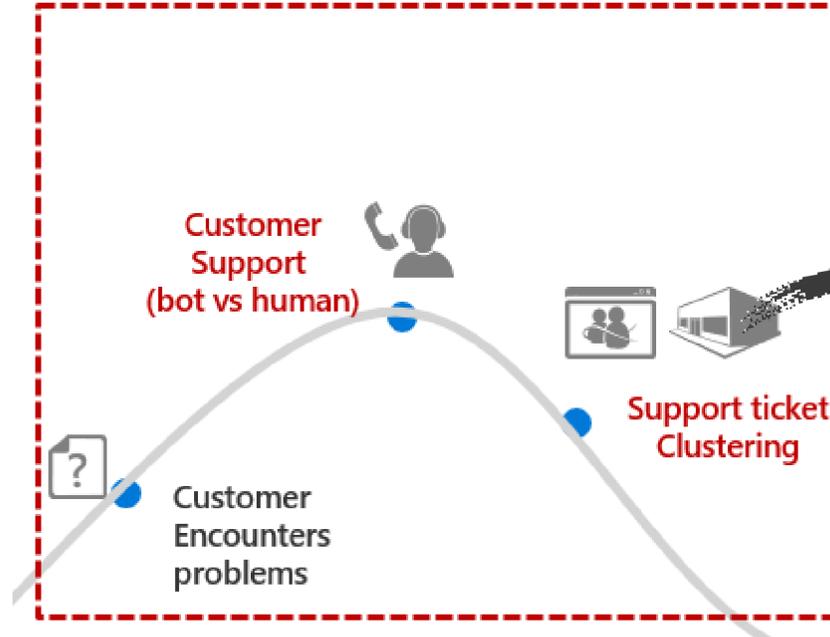
- ✓ Customer segmentation based on their browsing/purchase patterns
- ✓ ML based models can help determine customers who are most likely to churn in the near future based on their behavior



- ✓ AI powered personalized recommendations
- ✓ Using all the information available on customer's buying, browsing patterns determine the Lifetime value (LTV)

Marketing cycle – Support & Fraud

Support



- ✓ Automated answers based on a knowledge base can help decrease call center costs and predict staffing needs
- ✓ Support ticket clustering can help the team find solutions for customer complains quickly and even report similar problems back to the product teams

Fraud/Risk management



- ✓ While acquiring customers, external customer data can be used to determine customers who are likely to be risky for the business in the long run
- ✓ ML & AI models can be built to predict if a customer is likely to end up in bankruptcy, being delinquency which results in non-payment for the products/services

Fraud management use case

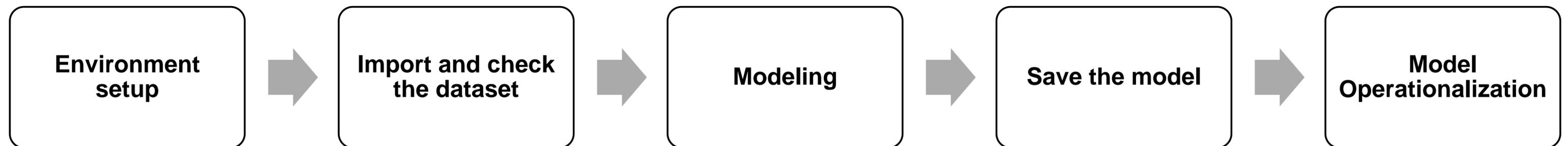
Credit Card Fraud Detection Data Set

- Dataset used:
 - contains transactions made by credit cards in September 2013 by European cardholders
 - presents transactions that occurred in two days, where we have 492 frauds out of 284,807 transactions
- Features $V1, V2, \dots, V28$: are the principal components obtained with PCA, the only features which have not been transformed with PCA are 'Time' and 'Amount'
- Feature *Time*: contains the seconds elapsed between each transaction and the first transaction in the dataset
- Feature *Amount*: is the transaction Amount
- Feature *Class*: is the response variable and it takes value 1 in case of fraud and 0 otherwise

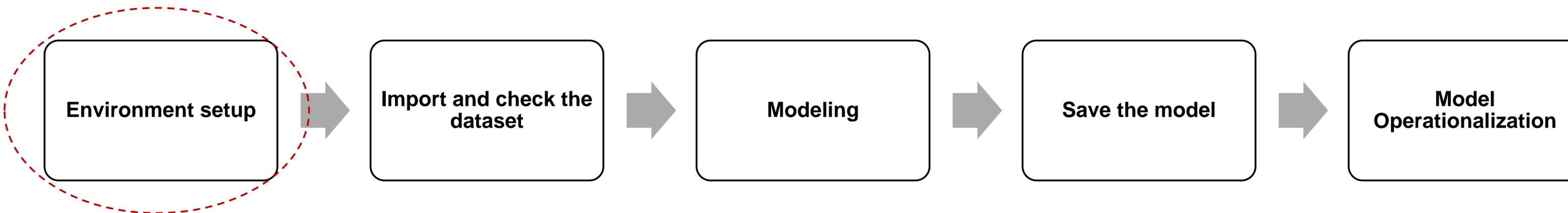
Time	V1	V2	V3	V4
0	-1.3598071336738	-0.0727811733098497	2.53634673796914	1.37815522427443
0	1.19185711131486	0.26615071205963	0.16648011335321	0.448154078460911
1	-1.35835406159823	-1.34016307473609	1.77320934263119	0.379779593034328
1	-0.966271711572087	-0.185226008082898	1.79299333957872	-0.863291275036453
2	-1.15823309349523	0.877736754848451	1.548717846511	0.403033933955121
2	-0.425965884412454	0.960523044882985	1.14110934232219	-0.168252079760302
4	1.22965763450793	0.141003507049326	0.0453707735899449	1.20261273673594
7	-0.644269442348146	1.41796354547385	1.0743803763556	-0.492199018495015

Demo

- The sample code was tested and run using the Jupyter notebook environment on a remote Azure VM (Standard F8s (8 vcpus, 16 GB memory))
- The sample code is available at the following GitHub location:
<https://github.com/jayamathew/Codebase/tree/master/conferences>
- The outline of the code is as follows:



Demo



- Import the necessary libraries and provide credentials to access the data

```
jupyter Strata_DL_Sept2018 Last Checkpoint: a minute ago (autosaved) Python [conda env:py35]
```

```
File Edit View Insert Cell Kernel Widgets Help Trusted Python [conda env:py35]
```

```
Environment setup
```

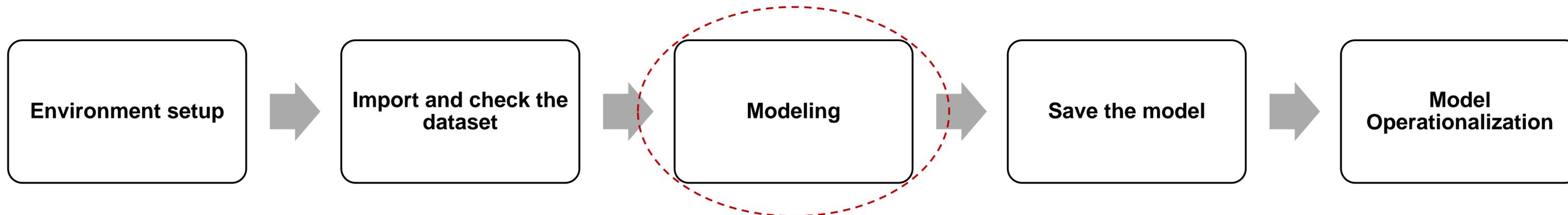
```
In [ ]: # Import necessary components
import os
import keras
import shutil
import json

In [ ]: import re
import pandas as pd
import numpy as np
import datetime

from sklearn import preprocessing
from sklearn.metrics import confusion_matrix, recall_score, precision_score
from keras.models import Sequential
from keras.layers import Dense, Dropout, LSTM, Activation
from math import ceil

In [ ]: import pickle
from scipy import stats
import tensorflow as tf
from sklearn.model_selection import train_test_split
from keras.models import Model, load_model
from keras.layers import Input, Dense
from keras.callbacks import ModelCheckpoint, TensorBoard
from keras import regularizers
```


Demo



- Build any model and tune hyper parameters (if needed)

```
jupyter Strata_DL_Sept2018 Last Checkpoint: 3 minutes ago (autosaved) Python [conda env:py35] Logout
File Edit View Insert Cell Kernel Widgets Help Trusted Python [conda env:py35]
Modeling
First exclude the variable 'Time'. Since the spread of the variable 'Amount' is large, this variable is standardized.
In [ ]: # Remove the column 'Time' and standardize the variable 'Amount'
from sklearn.preprocessing import StandardScaler
data = cc.drop(['Time'], axis=1)
data['Amount'] = StandardScaler().fit_transform(data['Amount'].values.reshape(-1, 1))

Next step is to split the data into train/test.
In [ ]: # Split the data into train/test and remove variable 'Class' and prepare for autoencoder
X_train, X_test = train_test_split(data, test_size=0.3, random_state=123)
X_train = X_train.drop(['Class'], axis=1)
y_test = X_test['Class']
X_test = X_test.drop(['Class'], axis=1)
X_train = X_train.values
X_test = X_test.values

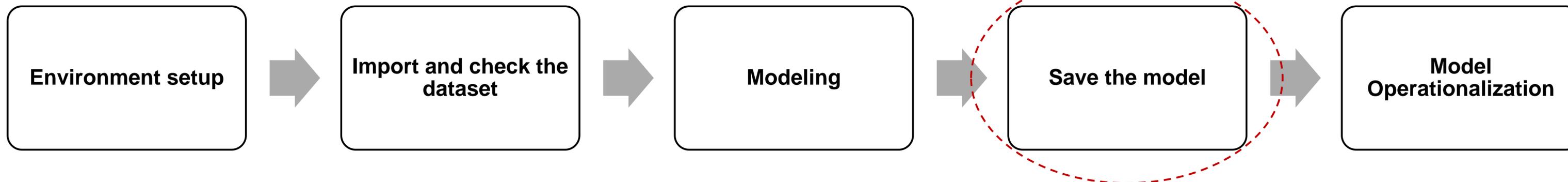
print("X_train:")
print(X_train.shape)
print("X_test:")
print(X_test.shape)

In [ ]: X_test[1]

Define the framework for the autoencoder and then compile and fit using the training data.
In [ ]: # Define the encoded/decoder framework
input_dim = X_train.shape[1]
encoding_dim = 14

input_layer = Input(shape=(input_dim,))
encoder = Dense(encoding_dim, activation="tanh", activity_regularizer=regularizers.l1(10e-5))(input_layer)
decoder = Dense(int(encoding_dim / 2), activation="tanh")(encoder)
decoder = Dense(input_dim, activations="relu")(decoder)
autoencoder = Model(inputs=input_layer, outputs=decoder)
```

Demo



- Save the best model for operationalization

jupyter Strata_DL_Sept2018 Last Checkpoint: 4 minutes ago (autosaved) Logout

File Edit View Insert Cell Kernel Widgets Help Trusted | Python [conda env:py35]

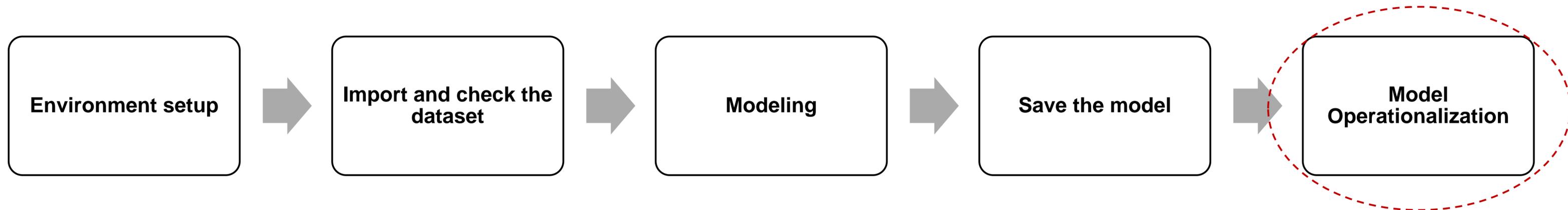
Saving the model

```
In [ ]: autoencoder

In [ ]: # Save the model for operationalization: https://machinelearningmastery.com/save-load-keras-deep-learning-models/
from keras.models import model_from_json
import os
import h5py
from sklearn import datasets

# save model
# serialize model to JSON
model_json = autoencoder.to_json()
with open("C://dsvm//notebooks/autoencoder.json", "w") as json_file:
    json_file.write(model_json)
# serialize weights to HDF5
autoencoder.save_weights("C://dsvm//notebooks/autoencoder.h5")
print("Model saved")
```

Demo



- Create the necessary functions for model operationalization using any tool of choice

jupyter Strata_DL_Sept2018 Last Checkpoint: 4 minutes ago (autosaved) Logout

File Edit View Insert Cell Kernel Widgets Help Trusted Python [conda env:py35]

Deployment

There are multiple options to operationalize a model, this is entirely dependent on the tools used.

Once the assets (model, schema file, scoring script etc.) are stored, we can download them into a deployment compute context for operationalization on an Azure web service. For this scenario, we will deploy this on our local context. We demonstrate how to setup this web service this through a CLI window opened in the AML.

Create a model management endpoint

Create a modelmanagement under your account. We will call this automodelmanagement. The remaining defaults are acceptable. `az ml account modelmanagement create --location <ACCOUNT_REGION> --resource-group <RESOURCE_GROUP> --name automodelmanagement` You can find the subscription name or subscription id through the (<https://portal.azure.com>) under the resource group you'd like to use.

Check environment settings

Show what environment is currently active: `az ml env show`

If nothing is set, we setup the environment with the existing model management context first: `az ml env setup --location <ACCOUNT_REGION> --resource-group <RESOURCE_GROUP> --name automodelmanagement` using the same and in the previous section.

Then set the current environment: `az ml env set --resource-group <RESOURCE_GROUP> --cluster-name automodelmanagement`

Check that the environment is now set: `az ml env show`

Links to get started with AI

- Cognitive Services: <https://aka.ms/AICognitiveServices>
- Azure ML: <https://aka.ms/AICustomModels>
- Azure Machine Learning Studio: <https://aka.ms/AzureStudio>
- Azure Machine Learning Services: <https://aka.ms/AMLServices>
- Visual Studio Code Tools for AI: <https://aka.ms/VSCodeToolsAI>
- Azure Notebooks: <https://aka.ms/AzureJNotebooks>
- Preconfigured Virtual Machines: <https://aka.ms/AzureVirtualMachines>
- Deep Learning Virtual Machine: <https://aka.ms/AzureDSVM>
- Team Data Science Process: <https://aka.ms/TeamDataScience>
- Data Source for demo: <https://www.kaggle.com/mlg-ulb/creditcardfraud>
- Blog Post by Venelin Valkov: <https://medium.com/@curiously/credit-card-fraud-detection-using-autoencoders-in-keras-tensorflow-for-hackers-part-vii-20e0c85301bd>
- GitHub location for demo: <https://github.com/jayamathew/Codebase/tree/master/conferences>
- Deep Learning Book by Ian Goodfellow, Yoshua Bengio, Aaron Courville: <http://www.deeplearningbook.org/>

Thank You!

Francesca Lazzeri & Jaya Mathew

 @frlazzeri

 @mathew_jaya

Strata
DATA CONFERENCE