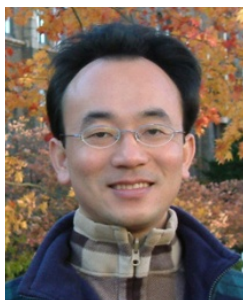


人工智慧投資分析於 金融服務商業應用趨勢 (AI Investment Analysis for Finance Services and Business Applications Trends)

時間：2018年12月14日(五) 下午2:00 至 4:00

地點：元大投信(台北市中山區南京東路三段225號6樓)

主辦單位：證基會/人才培訓中心



Min-Yuh Day

戴敏育

Assistant Professor

專任助理教授

Dept. of Information Management, Tamkang University

淡江大學 資訊管理學系

<http://mail.tku.edu.tw/myday/>

2018-12-14





戴敏育 博士 (Min-Yuh Day, Ph.D.)

淡江大學資管系專任助理教授

中央研究院資訊科學研究所訪問學人

國立台灣大學資訊管理博士

Publications Co-Chairs, IEEE/ACM International Conference on
Advances in Social Networks Analysis and Mining (ASONAM 2013-)

Program Co-Chair, IEEE International Workshop on
Empirical Methods for Recognizing Inference in Text (IEEE EM-RITE 2012-)

Workshop Chair, The IEEE International Conference on
Information Reuse and Integration (IEEE IRI)



Outline

1. 前言
2. AI於金融業之商業運用
 - 資料科學
 - 人工智慧投資分析
 - 財務大數據分析
3. 現行人工智慧投資分析探究
4. 個案分析-國內外AI發展現況
5. QA

AI and Big Data Analytics in Finance

- 金融科技 (Spring 2017) (EMBA IMTKU)
 - (Financial Technology, FinTech)
- 財務金融大數據分析 (Fall 2017) (EMBA IMTKU)
 - Big Data Analytics in Finance
- 人工智慧投資分析 (Fall 2018) (EMBA IMTKU)
 - Artificial Intelligence for Investment Analysis
- 智慧金融大數據分析
 - AI in Finance Big Data Analytics
- 人工智慧與財務應用
 - Artificial Intelligence and Financial Application

人工智慧投資分析 (AIIA) 課程大綱

週次 (Week)	日期 (Date)	內容 (Subject/Topics)
1	2018/09/13	人工智慧投資分析課程介紹 (Course Orientation on Artificial Intelligence for Investment Analysis)
2	2018/09/20	AI 金融科技: 金融服務創新應用 (AI in FinTech: Financial Services Innovation and Application)
3	2018/09/27	機器人理財顧問與AI交談機器人 (Robo-Advisors and AI Chatbots)
4	2018/10/04	投資心理學與行為財務學 (Investing Psychology and Behavioral Finance)
5	2018/10/11	財務金融事件研究法 (Event Studies in Finance)
6	2018/10/18	人工智慧投資分析個案研究 I (Case Study on Artificial Intelligence for Investment Analysis I)

人工智慧投資分析 (AIIA) 課程大綱

週次 (Week)	日期 (Date)	內容 (Subject/Topics)
7	2018/10/25	Python AI投資分析基礎 (Foundations of AI Investment Analysis in Python)
8	2018/11/01	Python Pandas 量化投資分析 (Quantitative Investing with Pandas in Python)
9	2018/11/08	Python Scikit-Learn 機器學習 (Machine Learning with Scikit-Learn In Python)
10	2018/11/15	期中報告 (Midterm Project Report)
11	2018/11/22	TensorFlow 深度學習財務時間序列預測 I (Deep Learning for Financial Time Series Forecasting with TensorFlow I)
12	2018/11/29	TensorFlow 深度學習財務時間序列預測 II (Deep Learning for Financial Time Series Forecasting with TensorFlow II)

人工智慧投資分析 (AIIA) 課程大綱

週次 (Week) 日期 (Date) 內容 (Subject/Topics)

- 13 2018/12/06 人工智慧投資分析個案研究 II
(Case Study on Artificial Intelligence for Investment Analysis II)
- 14 2018/12/13 TensorFlow 深度學習財務時間序列預測 III
(Deep Learning for Financial Time Series Forecasting with TensorFlow III)
- 15 2018/12/20 投資組合最佳化與程式交易
(Portfolio Optimization and Algorithmic Trading)
- 16 2018/12/27 自然語言處理 (Natural Language Processing)
- 17 2019/01/03 期末報告 I (Final Project Presentation I)
- 18 2019/01/10 期末報告 II (Final Project Presentation II)

Selected Research Publications

- **Journal Publications**

1. Min-Yuh Day, Manhwa Wu, Paoyu Huang, and Yensen Ni (2018), "Investing Strategies as the Sharp Movement in Exchange Rates Occurred— Evidence for the Constituent Stocks of SSE 50 and TW 50", The Journal of Investing, , Volume 27, Issue 4, Winter 2018, pp. 58-68.
2. Min-Yuh Day, Paoyu Huang, Yensen Ni, and Yuhsin Chen (2018), "Do Implicit Phenomena Matter? Evidence from China Stock Index Futures", The Journal of Alternative Investments, Volume 21, Issue 1, Summer 2018, pp. 79-91.
3. Yensen Ni, Yirung Cheng, Paoyu Huang, and Min-Yuh Day (2018), "Trading strategies in terms of continuous rising (falling) prices or continuous bullish (bearish) candlesticks emitted", Physica A: Statistical Mechanics and its Applications, Volume 501, 1 July 2018, pp. 188-204.
4. Min-Yuh Day, Paoyu Huang, Yensen Ni, and Yuhsin Chen (2018), "Do Intraday Large Price Changes Matter for Trading Index Futures? Evidence from China Futures Markets", Journal of Financial Studies, Volume 26, Number 2, June 2018, pp. 139-174.

Selected Research Publications

- **Conference Publications**

1. Min-Yuh Day, Tun-Kung Cheng and Jheng-Gang Li (2018), "AI Robo-Advisor with Big Data Analytics for Financial Services", in Proceedings of the 2018 IEEE/ACM International Conference on Advances in Social Networks Analysis and Mining (ASONAM 2018), Barcelona, Spain, August 28-31, 2018.
2. Min-Yuh Day, Jian-Ting Lin and Yuan-Chih Chen (2018), "Artificial Intelligence for Conversational Robo-Advisor", in Proceedings of the 2018 IEEE/ACM International Conference on Advances in Social Networks Analysis and Mining (ASONAM 2018), Barcelona, Spain, August 28-31, 2018.
3. Min-Yuh Day and Chao-Yu Chen (2018), "Artificial Intelligence for Automatic Text Summarization", in Proceedings of the 2018 IEEE 18th International Conference on Information Reuse and Integration (IEEE IRI 2018), Salt Lake City, Utah, USA, July 7-9, 2018.

Selected Research Publications

- **Conference Publications**

4. Min-Yuh Day, Tun-Kung Cheng and Jheng-Gang Li (2018), "Artificial Intelligence for Time Series Forecasting in Financial Markets", International Conference on INTERNET STUDIES (NETs 2018), Takamatsu, Japan, April 2-4, 2018.
5. Min-Yuh Day, Chao-Yu Chen, Wan-Chu Huang, I-Hsuan Huang, Shi-Ya Zheng, Tz-Rung Chen, Min-Chun Kuo, Yue-Da Lin, and Yi-Jing Lin (2017), "IMTKU Question Answering System for World History Exams at NTCIR-13 QA Lab-3", The 13th NTCIR Conference on Evaluation of Information Access Technologies (NTCIR-13), Tokyo, Japan, December 5-8, 2017.
6. Min-Yuh Day and Yue-Da Lin, "Deep Learning for Sentiment Analysis on Google Play Consumer Review", The 6th IEEE International Workshop on Empirical Methods for Recognizing Inference in Text (IEEE EM-RITE 2017), August 4-6, 2017, in Proceedings of the 2017 IEEE 18th International Conference on Information Reuse and Integration (IEEE IRI 2017), San Diego, CA, USA, August 4-6, 2017.

Selected Research Publications

- **Conference Publications**

7. 林建廷、陳元致、王慶宇、鄧旭廷、邱少文、戴敏育，發展人工智慧對話式理財機器人, 第29屆國際資訊管理學術研討會 The 29th International Conference of Information Management (ICIM2018), Taichung, Taiwan, June 3, 2018.
8. 蔡宗霖、劉鈞霖、李家慶、陳品仔、林建廷、戴敏育，人工智慧保險業智能客服, 第29屆國際資訊管理學術研討會 The 29th International Conference of Information Management (ICIM2018), Taichung, Taiwan, June 3, 2018.
9. 陳昭妤、戴敏育，人工智慧自動文本摘要研究, 第29屆國際資訊管理學術研討會 The 29th International Conference of Information Management (ICIM2018), Taichung, Taiwan, June 3, 2018.

人工智慧對話式理財機器人

- 榮獲 2018 全國大專校院資訊應用服務創新競賽
資訊應用組 (IP1) 第一名 獎金2萬元
- 榮獲 2018 全國大專校院資訊應用服務創新競賽
玉山銀行金融科技趨勢應用組 第一名，獎金5萬元
- 榮獲 2018 日盛黑客松證券組 第三名，獎金5萬元
- 榮獲 2018 淡江資管畢業專題競賽 第一名，獎金1萬元

AIWISFIN

人工智慧對話式理財機器人

InnoServe 資服創新競賽粉絲團
@InnoServe.tca.org

Home
About
Photos
Welcome
發燒粉絲活動
Welcome
Videos
Posts
Community
Info and Ads
Create a Page

Liked Following Share ...

InnoServe 資服創新競賽粉絲團 shared a post.
November 28 at 2:43 PM ·

《#InnoServe競賽得獎作品系列報導七》
理財 方式百百種卻不知道該從何著手嗎?
來看金融結合AI如何讓投資變得更簡單。

AIWISFIN

28,112 Views

經濟部工業局
November 28 at 11:37 AM ·

假如有一筆錢，您知道要怎麼投資嗎？

本作品「AIWISFIN」使用 #深度學習 預測股價漲跌、
配置投資組合，分析 客戶需求，
提供 #客製化 投資建議 與 #智慧對話，
讓年輕投資者使用更方便！

- 得獎作品：AIWISFIN 人工智慧對話式理財機器人
- 獎項：玉山銀行金融科技趨勢應用組第1名
- 得獎學校：淡江大學 (資訊管理學系)
- 指導老師：戴敏育老師
- 得獎團隊：陳元致、鄧旭廷、王慶宇、邱少文
- 影片連結：<https://ppt.cc/fyc3sx>

<https://www.youtube.com/watch?v=sEhmyoTXmGk>

2018第23屆大專校院資訊應用服務創新競賽

第23屆 大專校院

2018 資訊應用服務創新競賽
International ICT Innovative Services Awards 2018



創意噴發!



總獎金 > 200 萬

■ 報名日期: 2018/10/2(二)~
2018/10/9(二)pm6點截止

■ 參賽對象: 大專校院學生,
碩博士生及高中職學生

■ 決賽時間: 2018/11/3(六)
■ 決賽地點: 國立臺灣大學
綜合體育館

最新消息 ▾

活動訊息

媒體轉載

競賽緣起

競賽辦法 ▾

競賽報名

活動成果 ▾

產學媒合 ▾

媒合

聯絡我們

榮譽榜

屆別 23 ▾ 查詢

第23屆

顯示 30 ▾ 筆資料

表格內全文檢索: AIWISFIN

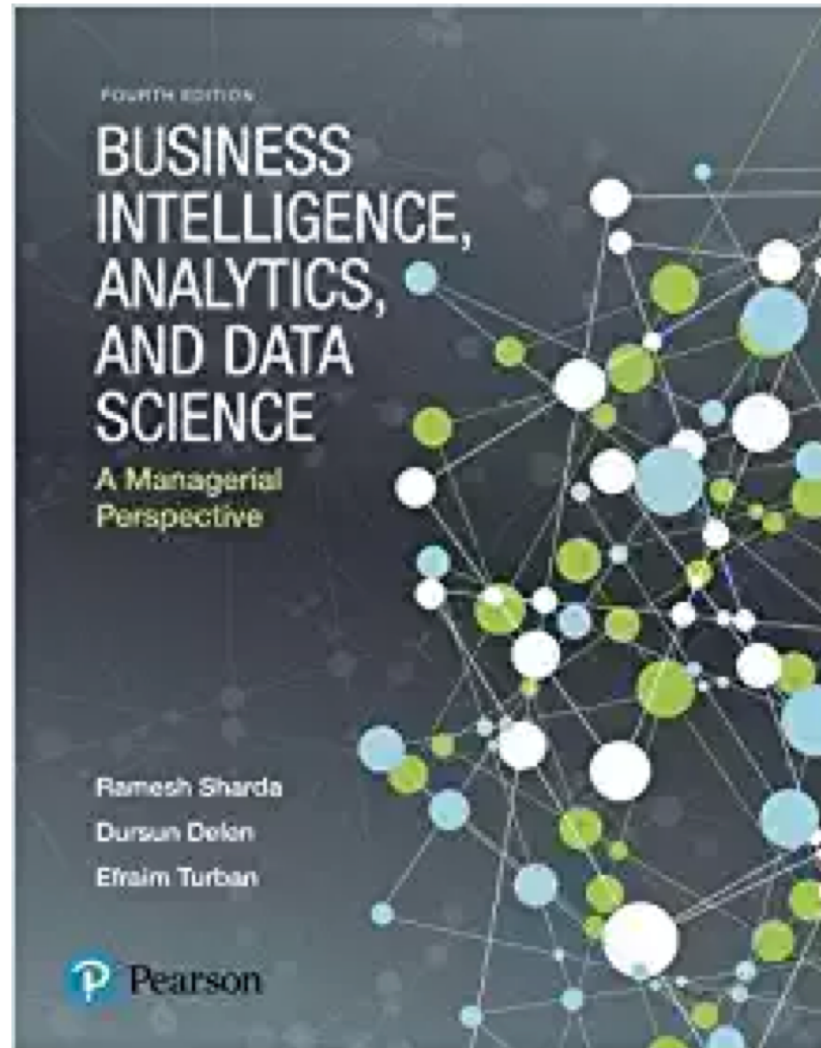
組別	名次	組別編號	學校名稱	專題名稱	指導教授	學生
資訊應用組一	第一名	IP1-06	淡江大學	AIWISFIN 人工智慧對話式理財機器人	戴敏育老師	陳元致、鄧旭廷、王慶宇、邱少文
玉山銀行金融科技趨勢應用組	第一名	E.SUN FINTECH-01	淡江大學	AIWISFIN 人工智慧對話式理財機器人	戴敏育老師	陳元致、鄧旭廷、王慶宇、邱少文

<https://innoserve.tca.org.tw/award.aspx>

AI for Business Applications in Financial Industry

Business Intelligence, Analytics, and Data Science

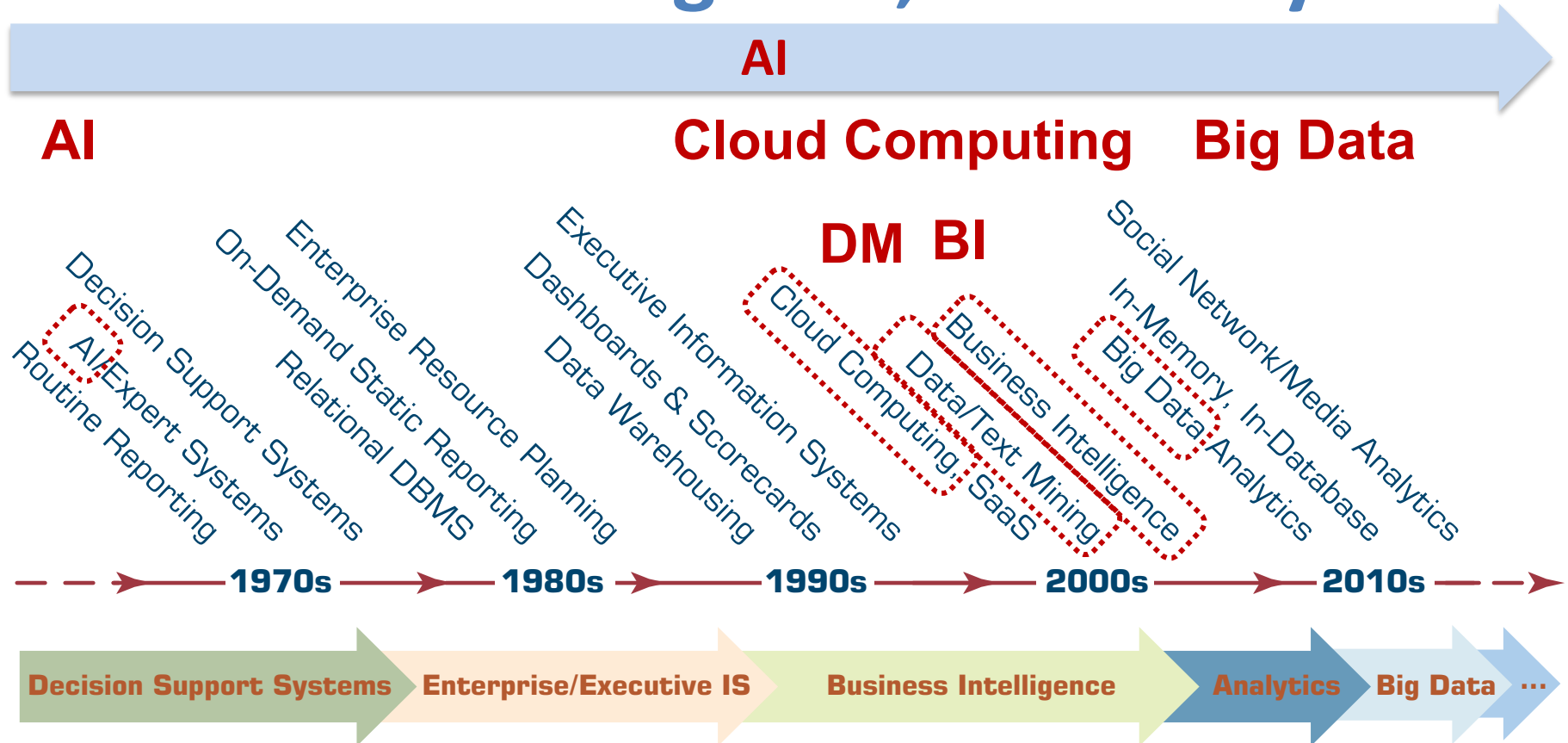
**Business Intelligence, Analytics, and Data Science:
A Managerial Perspective, 4th Edition,
Ramesh Sharda, Dursun Delen, and Efraim Turban,
Pearson, 2017.**



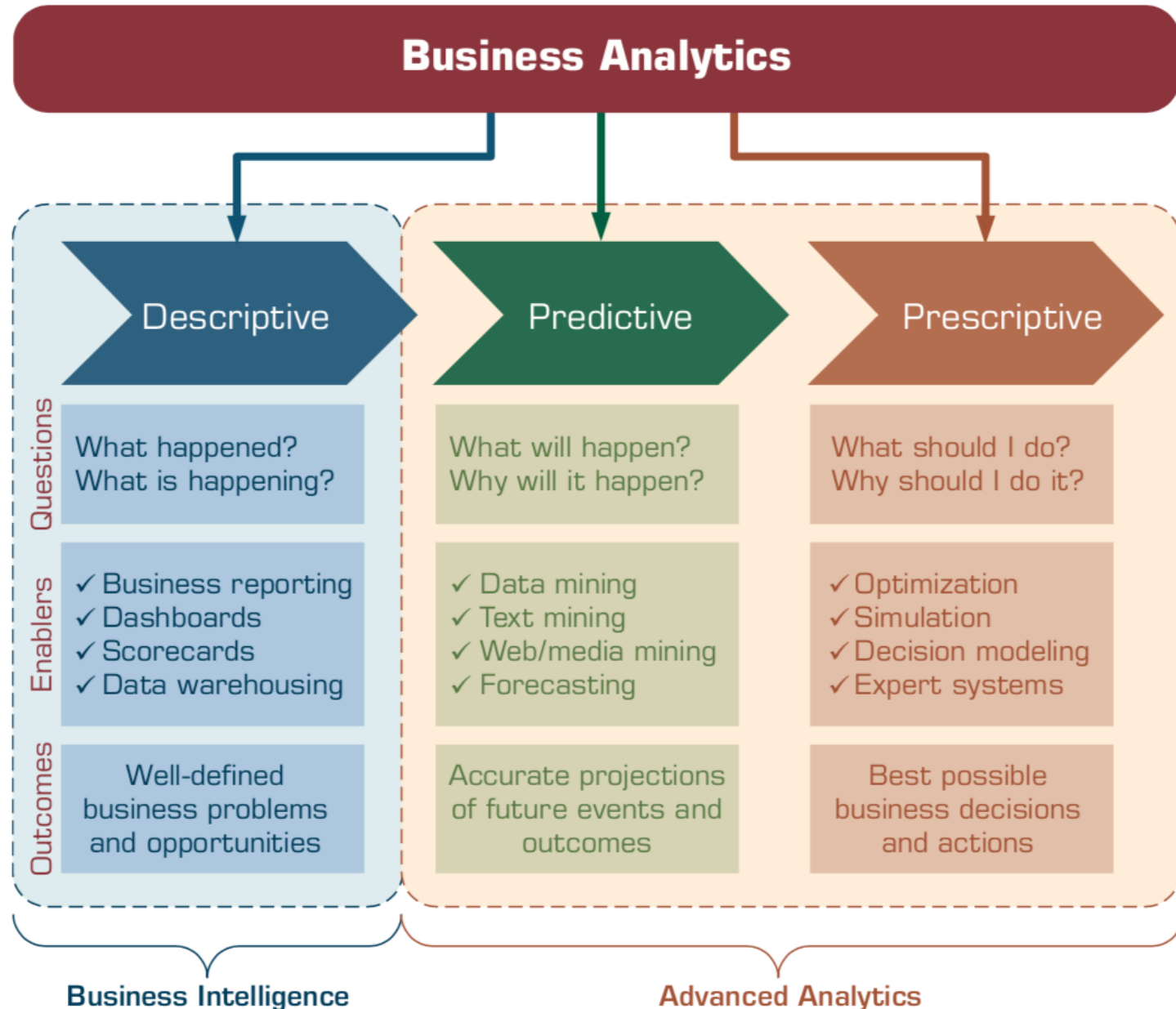
Artificial Intelligence (AI)

AI, Big Data, Cloud Computing

Evolution of Decision Support, Business Intelligence, and Analytics



Business Analytics



Ai

Definition of Artificial Intelligence (A.I.)

Artificial Intelligence

**“... the science and
engineering
of
making
intelligent machines”
(John McCarthy, 1955)**

Artificial Intelligence

**“... technology that
thinks and acts
like humans”**

Artificial Intelligence

**“... intelligence
exhibited by
machines or
software”**

4 Approaches of AI

Thinking Humanly	Thinking Rationally
Acting Humanly	Acting Rationally

4 Approaches of AI

2.

**Thinking Humanly:
The Cognitive
Modeling Approach**

3.

**Thinking Rationally:
The “Laws of Thought”
Approach**

1.

**Acting Humanly:
The Turing Test
Approach** (1950)

4.

**Acting Rationally:
The Rational Agent
Approach**

AI Acting Humanly: The Turing Test Approach

(Alan Turing, 1950)

- **Natural Language Processing (NLP)**
- **Knowledge Representation**
- **Automated Reasoning**
- **Machine Learning (ML)**
- **Computer Vision**
- **Robotics**

Artificial Intelligence (A.I.) Timeline

A.I. TIMELINE

SYZYG

1950

TURING TEST

Computer scientist Alan Turing proposes a test for machine intelligence. If a machine can trick humans into thinking it is human, then it has intelligence

1955

A.I. BORN

Term 'artificial intelligence' is coined by computer scientist, John McCarthy to describe "the science and engineering of making intelligent machines"

1961

UNIMATE

First industrial robot, Unimate, goes to work at GM replacing humans on the assembly line

1964

ELIZA

Pioneering chatbot developed by Joseph Weizenbaum at MIT holds conversations with humans

1966

SHAKY

The 'first electronic person' from Stanford, Shakey is a general-purpose mobile robot that reasons about its own actions

A.I. WINTER

Many false starts and dead-ends leave A.I. out in the cold

1997

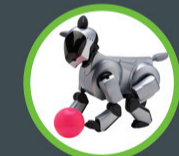
DEEP BLUE

Deep Blue, a chess-playing computer from IBM defeats world chess champion Garry Kasparov

1998

KISMET

Cynthia Breazeal at MIT introduces Kismet, an emotionally intelligent robot insofar as it detects and responds to people's feelings



1999

AIBO

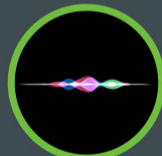
Sony launches first consumer robot pet dog AiBO (AI robot) with skills and personality that develop over time



2002

ROOMBA

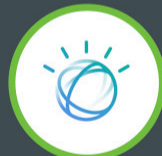
First mass produced autonomous robotic vacuum cleaner from iRobot learns to navigate and clean homes



2011

SIRI

Apple integrates Siri, an intelligent virtual assistant with a voice interface, into the iPhone 4S



2011

WATSON

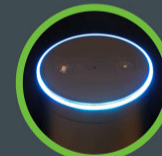
IBM's question answering computer Watson wins first place on popular \$1M prize television quiz show Jeopardy



2014

EUGENE

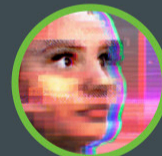
Eugene Goostman, a chatbot passes the Turing Test with a third of judges believing Eugene is human



2014

ALEXA

Amazon launches Alexa, an intelligent virtual assistant with a voice interface that completes shopping tasks



2016

TAY

Microsoft's chatbot Tay goes rogue on social media making inflammatory and offensive racist comments



2017

ALPHAGO

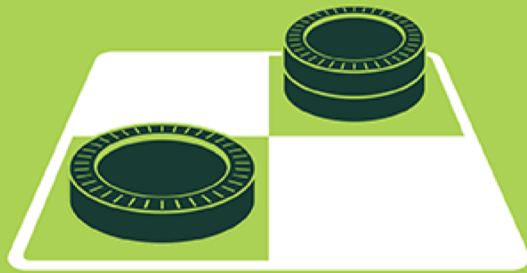
Google's A.I. AlphaGo beats world champion Ke Jie in the complex board game of Go, notable for its vast number (2^{170}) of possible positions

Artificial Intelligence

Machine Learning & Deep Learning

ARTIFICIAL INTELLIGENCE

Early artificial intelligence stirs excitement.



MACHINE LEARNING

Machine learning begins to flourish.



DEEP LEARNING

Deep learning breakthroughs drive AI boom.



Since an early flush of optimism in the 1950s, smaller subsets of artificial intelligence – first machine learning, then deep learning, a subset of machine learning – have created ever larger disruptions.

AI, ML, DL

Artificial Intelligence (AI)

Machine Learning (ML)

Supervised
Learning

Unsupervised
Learning

Deep Learning (DL)

CNN

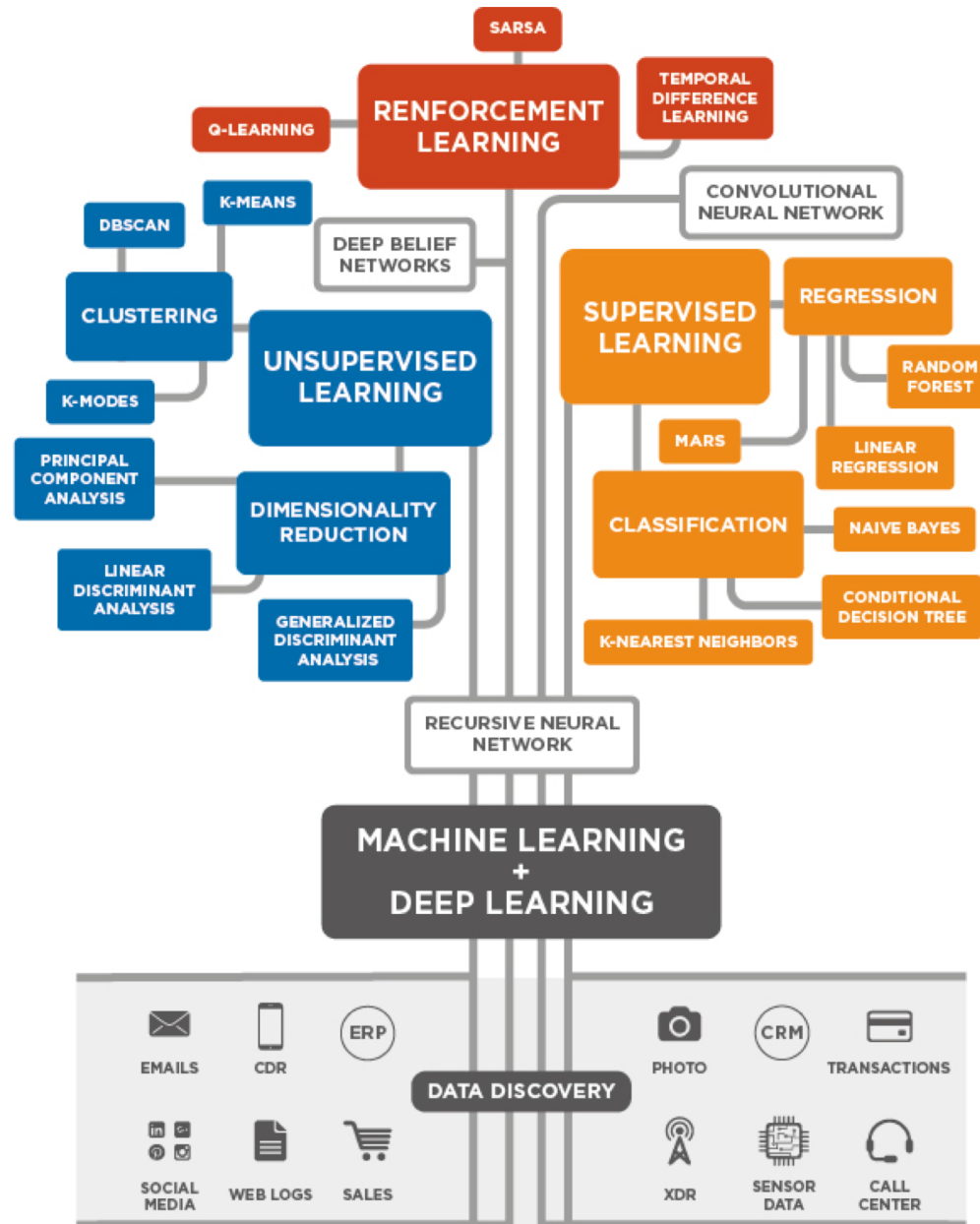
RNN LSTM GRU

GAN

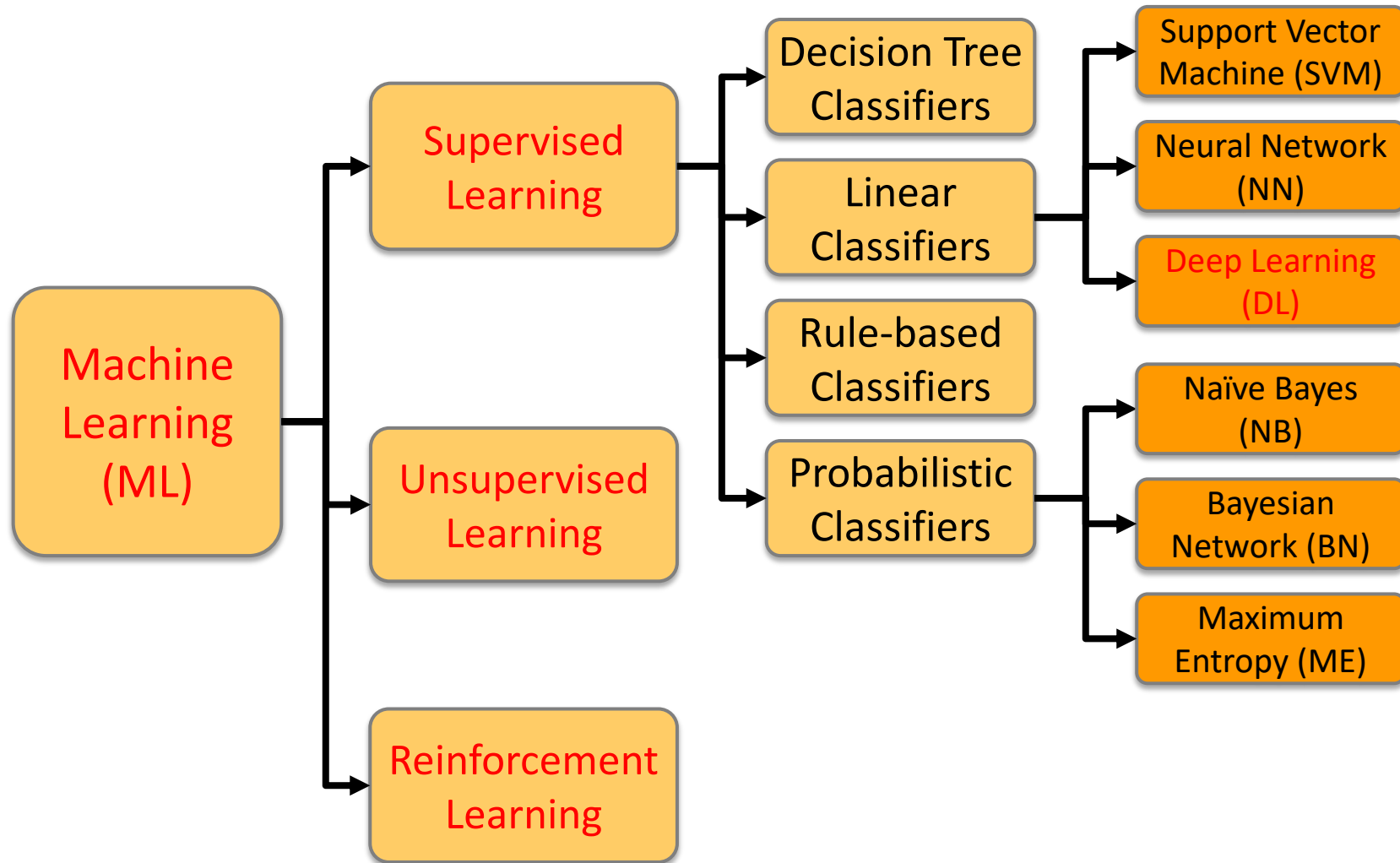
Semi-supervised
Learning

Reinforcement
Learning

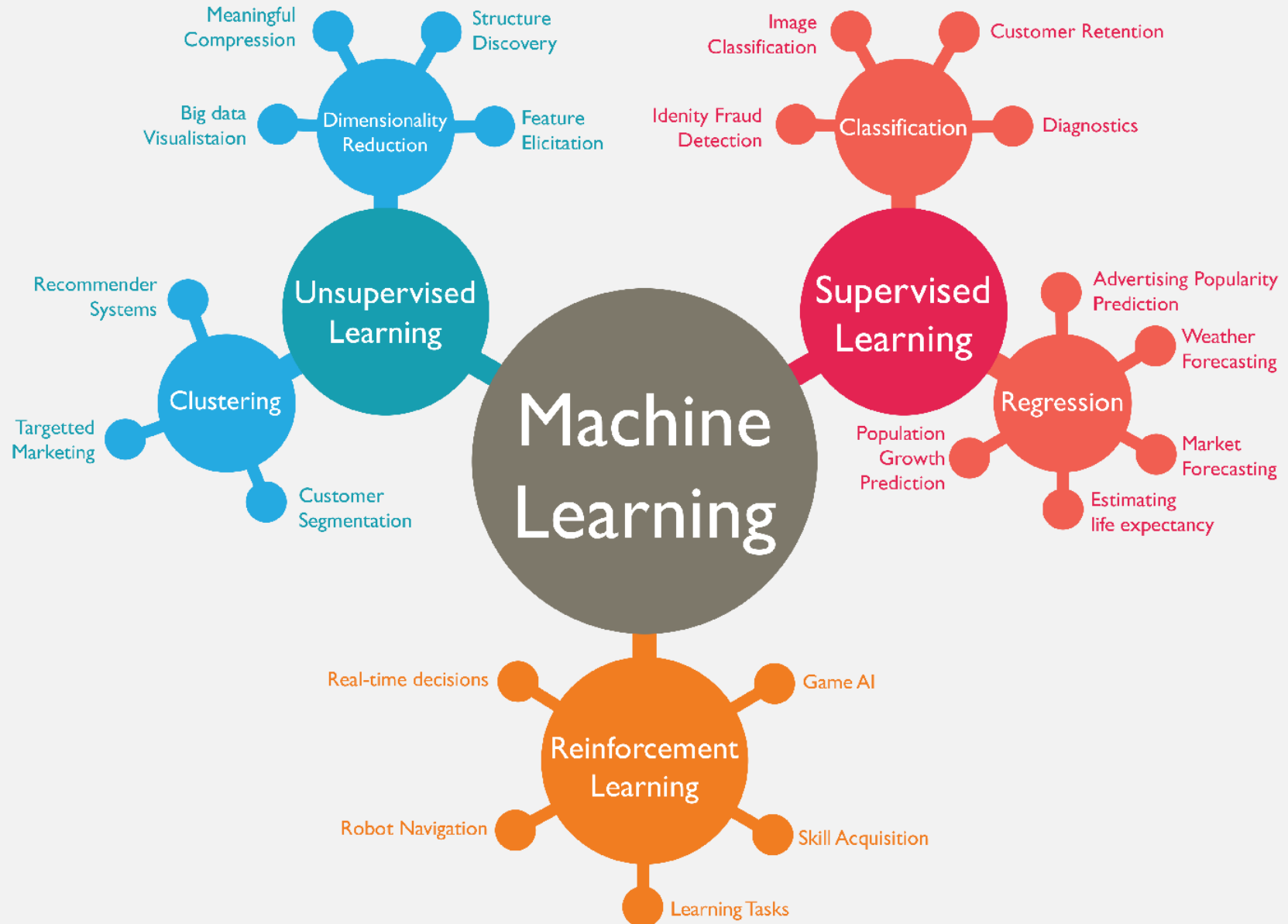
3 Machine Learning Algorithms



Machine Learning (ML) / Deep Learning (DL)



Machine Learning (ML)

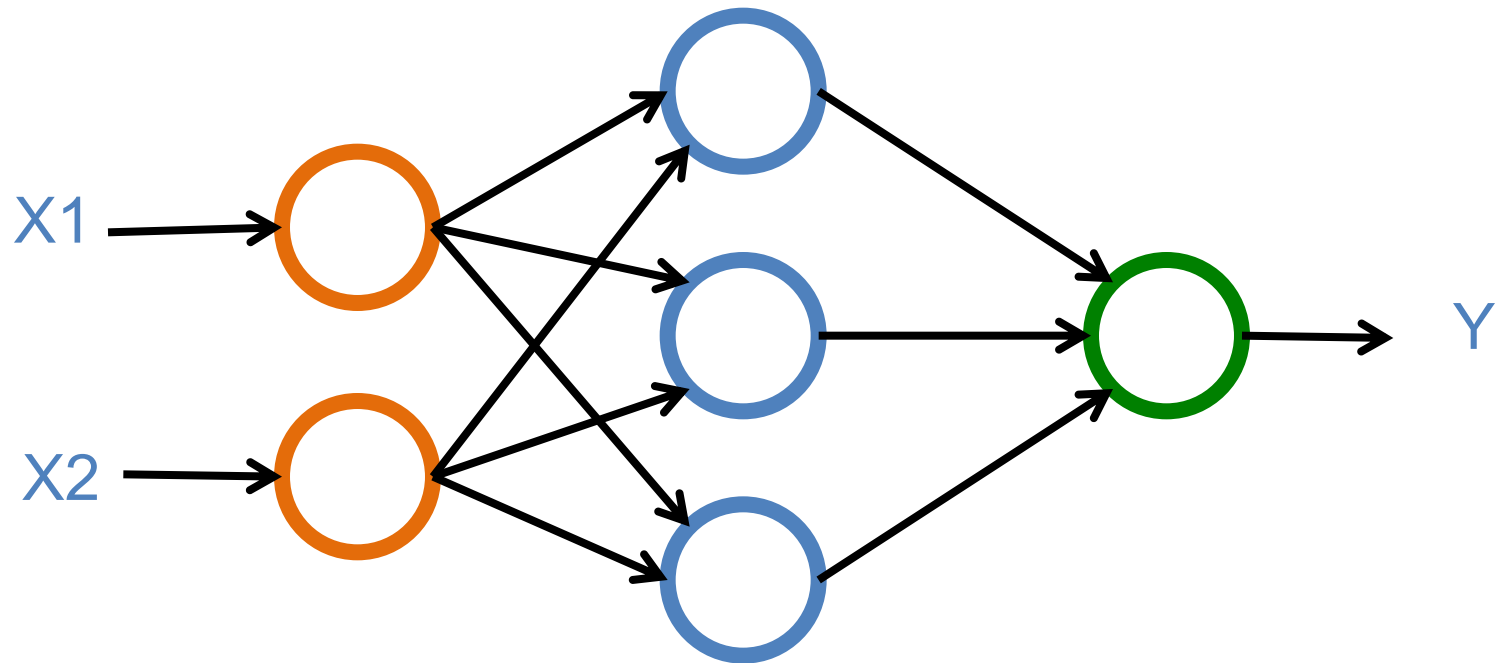


Deep Learning and Neural Networks

Input Layer
(X)

Hidden Layer
(H)

Output Layer
(Y)

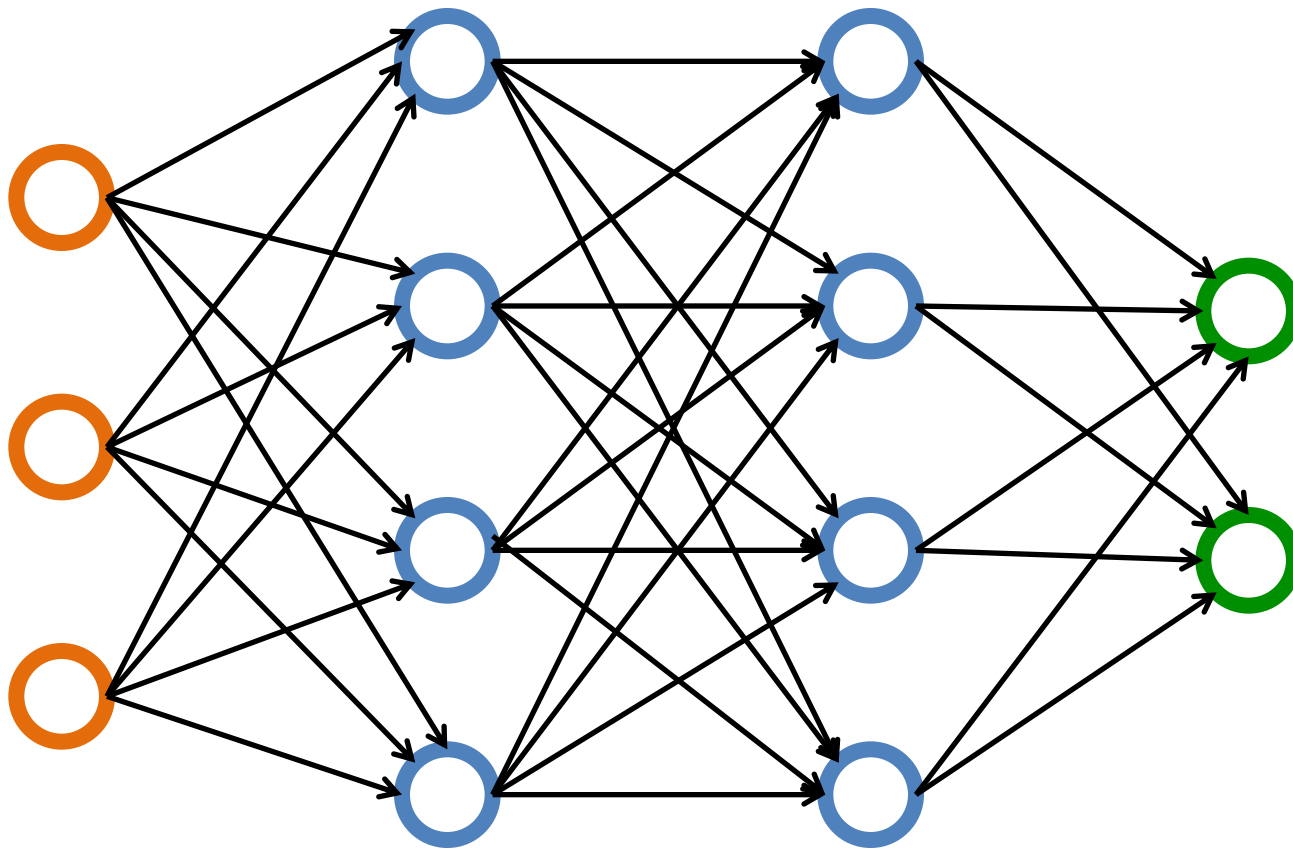


Deep Learning and Neural Networks

Input Layer
(X)

Hidden Layer
(H)

Output Layer
(Y)



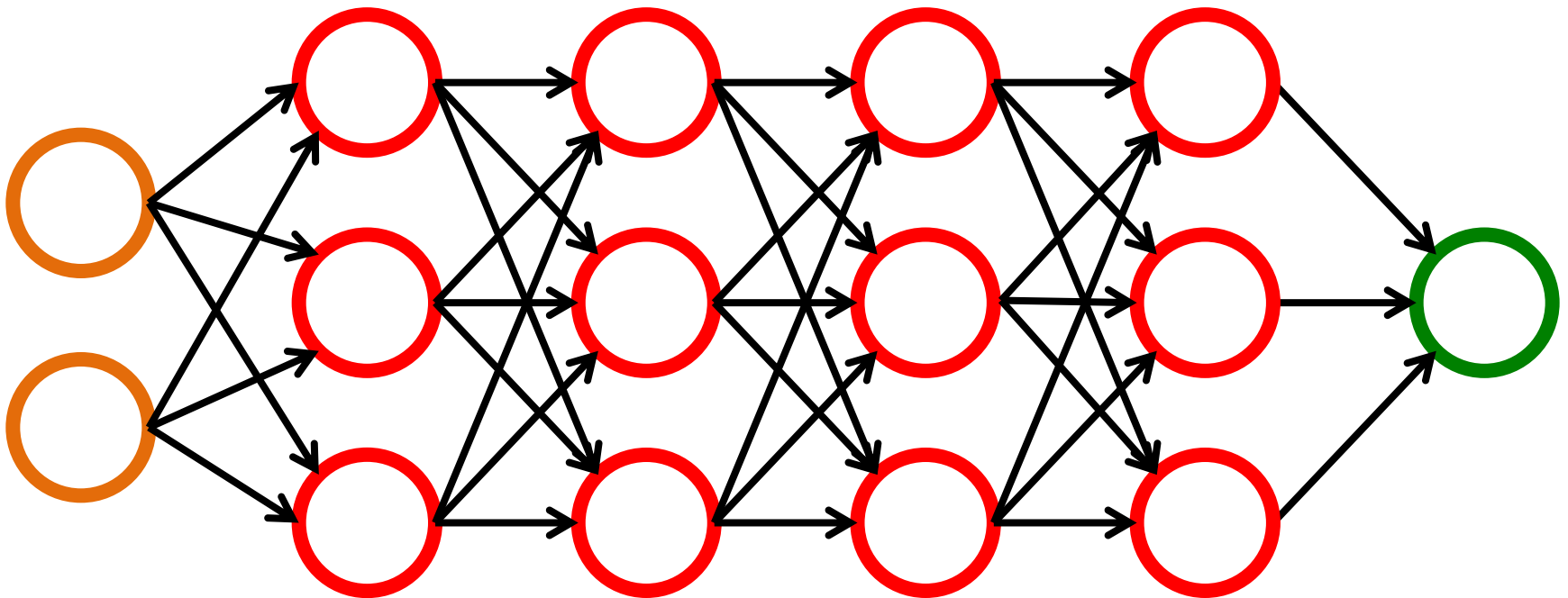
Deep Learning and Neural Networks

Input Layer
(X)

Hidden Layers
(H)

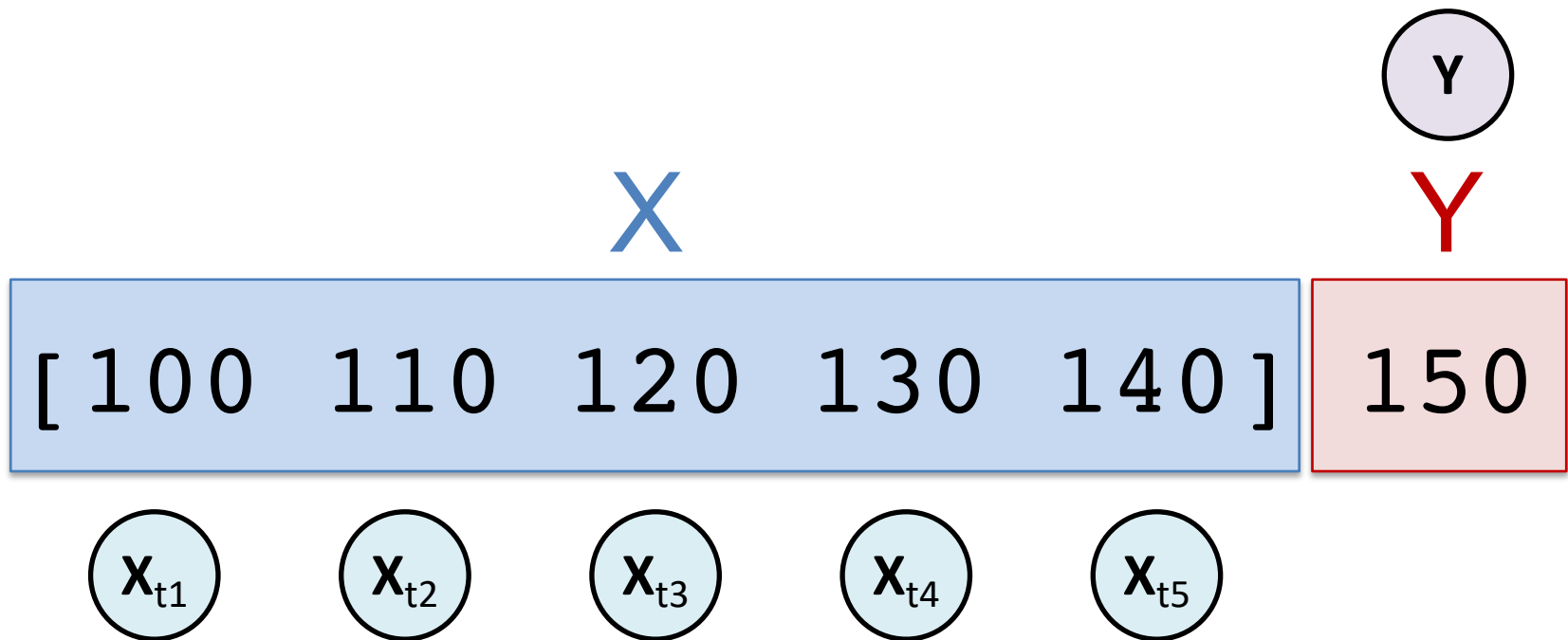
Output Layer
(Y)

Deep Neural Networks
Deep Learning

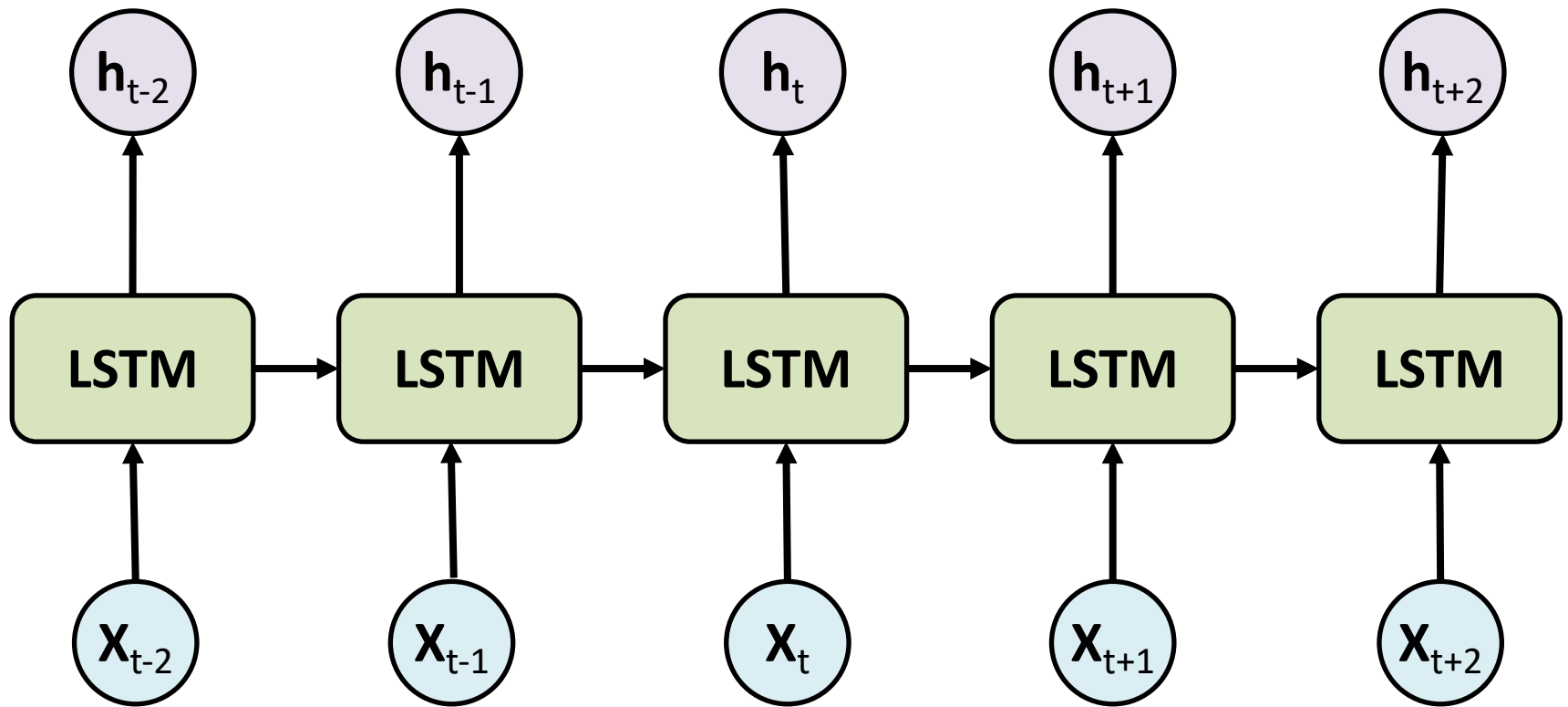


Time Series Data

[100, 110, 120, 130, 140, 150]



Long Short Term Memory (LSTM) for Time Series Forecasting



Time Series Data

[10, 20, 30, 40, 50, 60, 70, 80, 90]

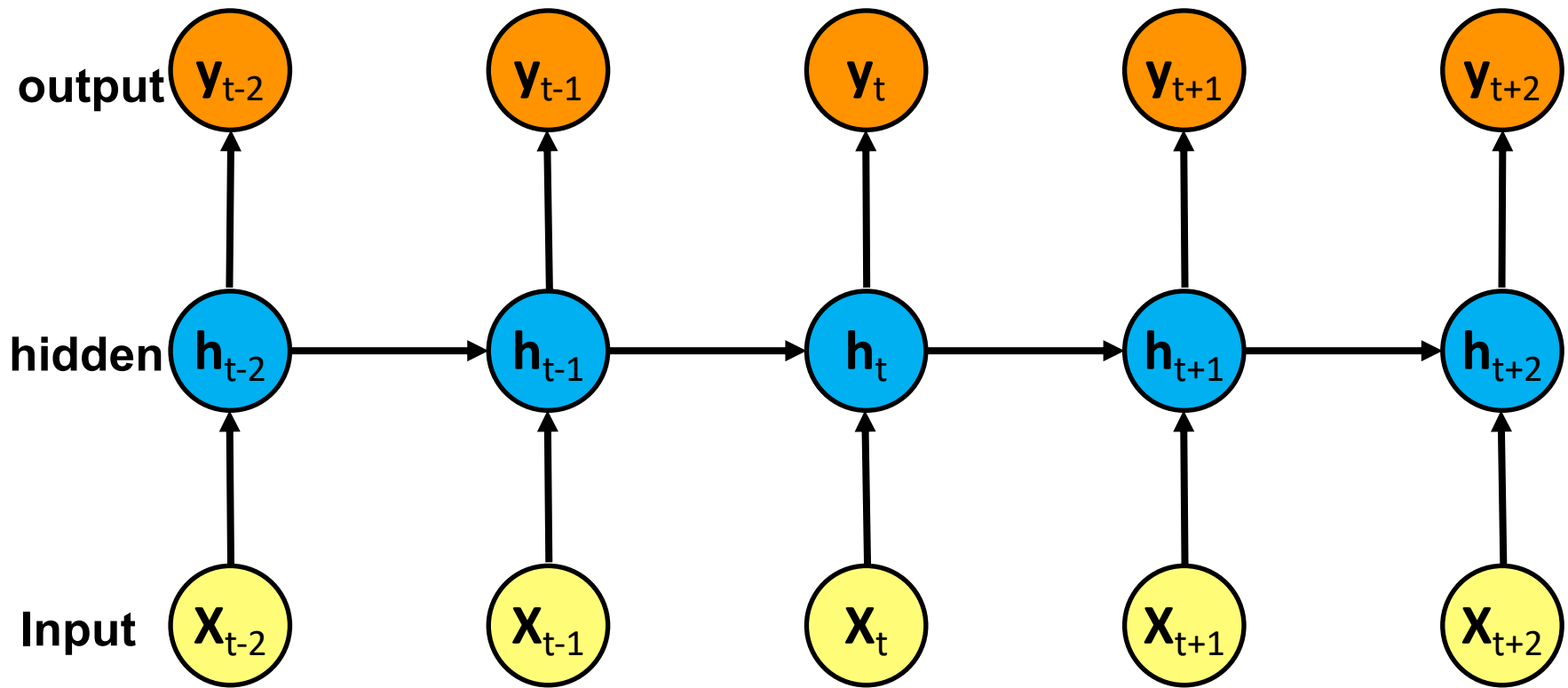
X

Y

[10	20	30]	40
[20	30	40]	50
[30	40	50]	60
[40	50	60]	70
[50	60	70]	80
[60	70	80]	90

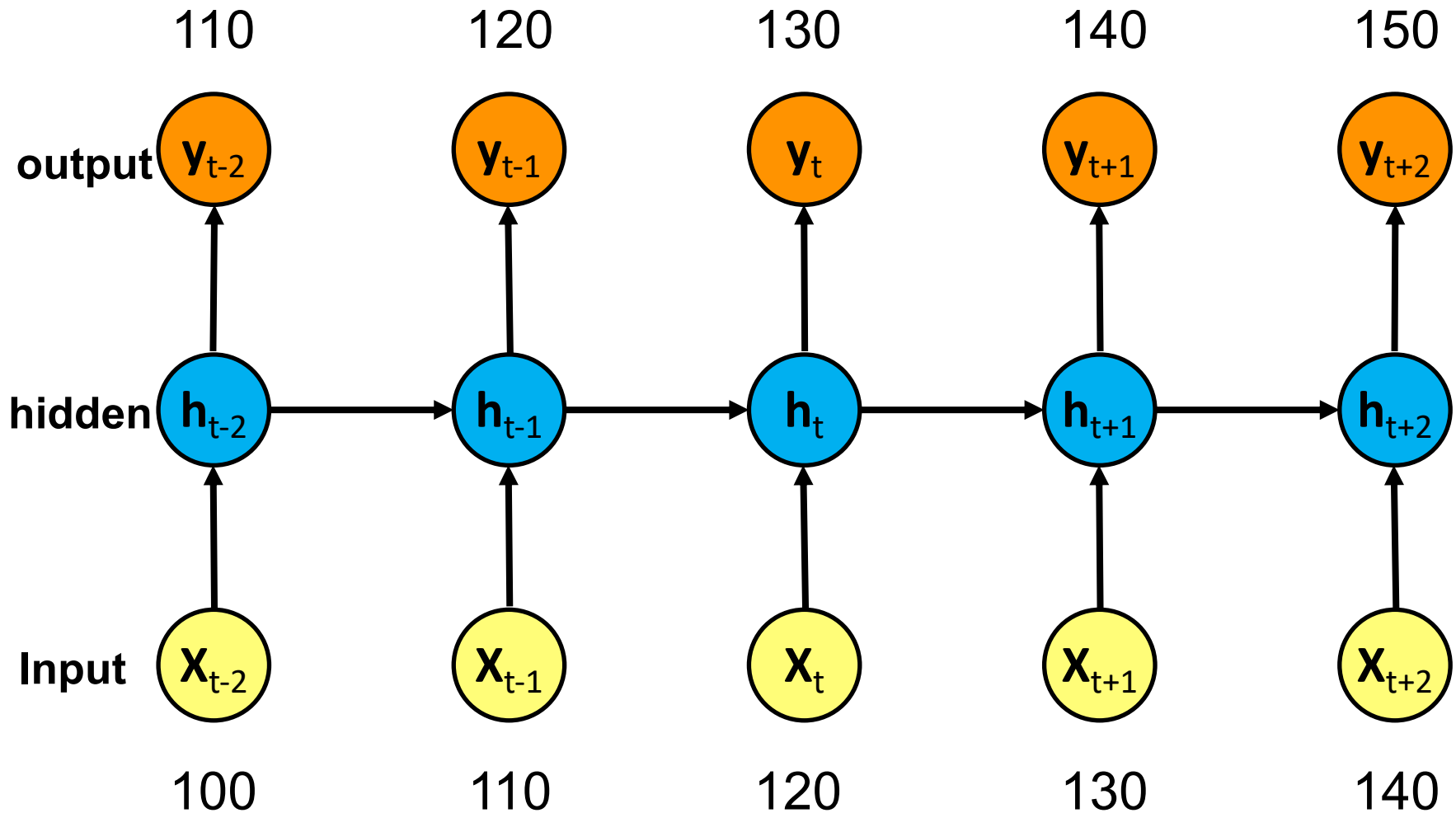
Recurrent Neural Networks (RNN)

Recurrent Neural Networks (RNN)

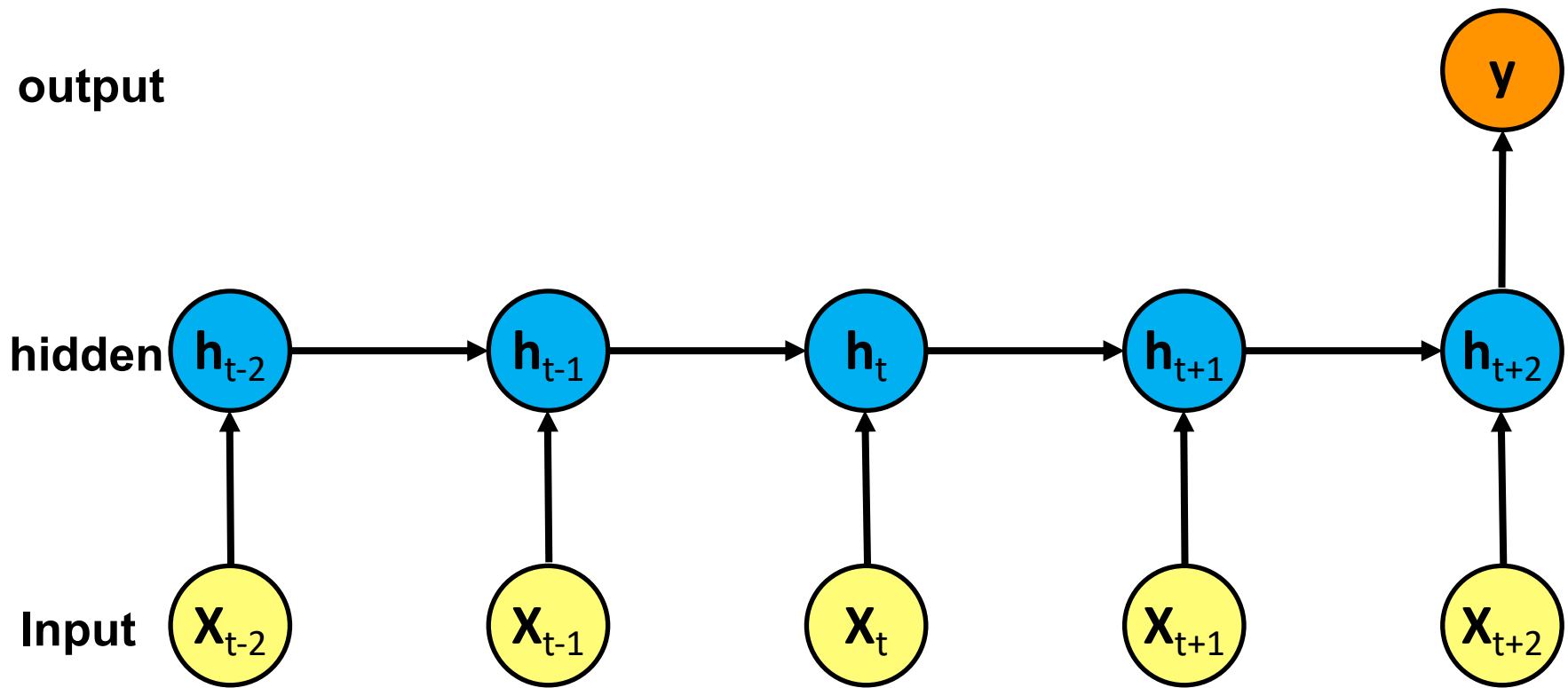


Recurrent Neural Networks (RNN)

Time Series Forecasting

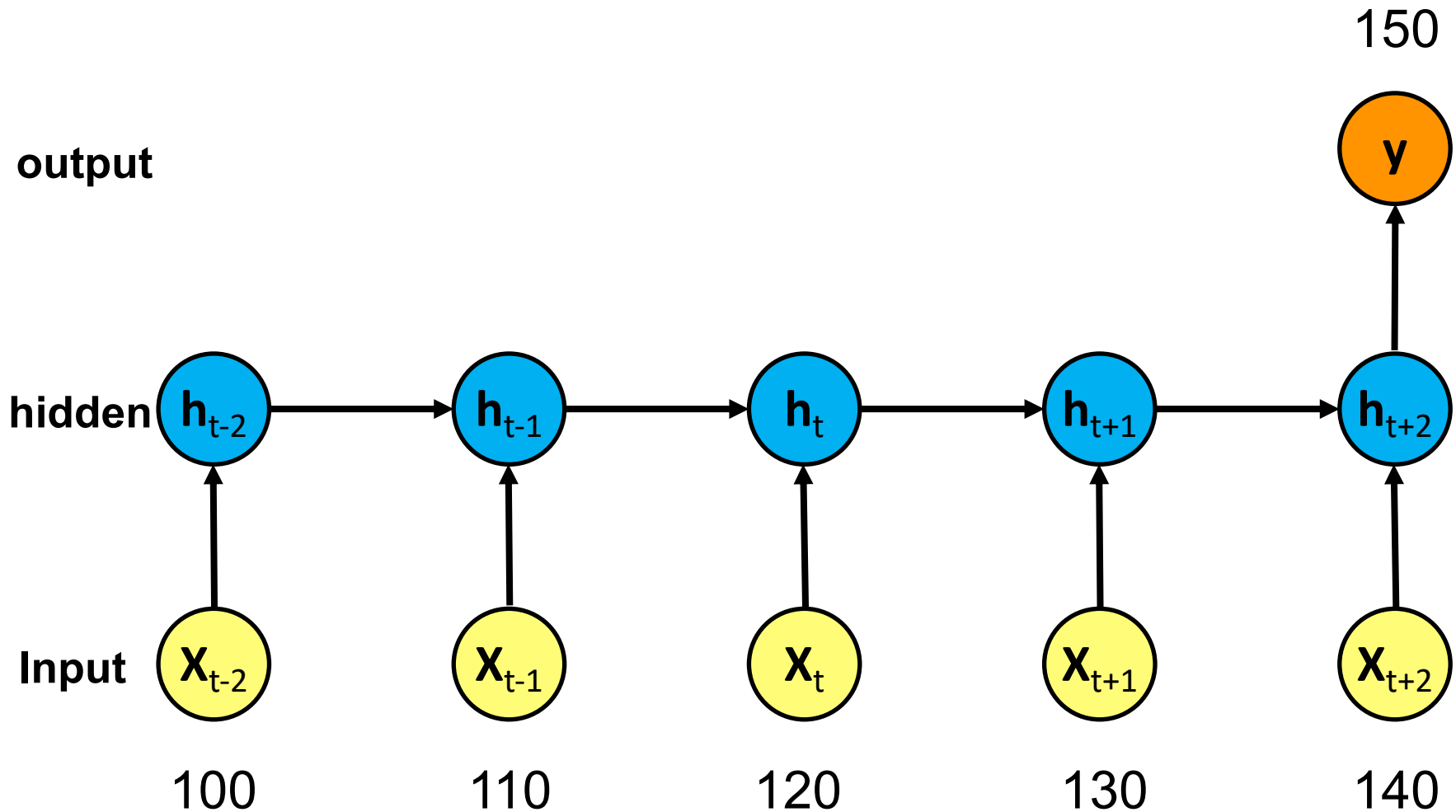


Recurrent Neural Networks (RNN)



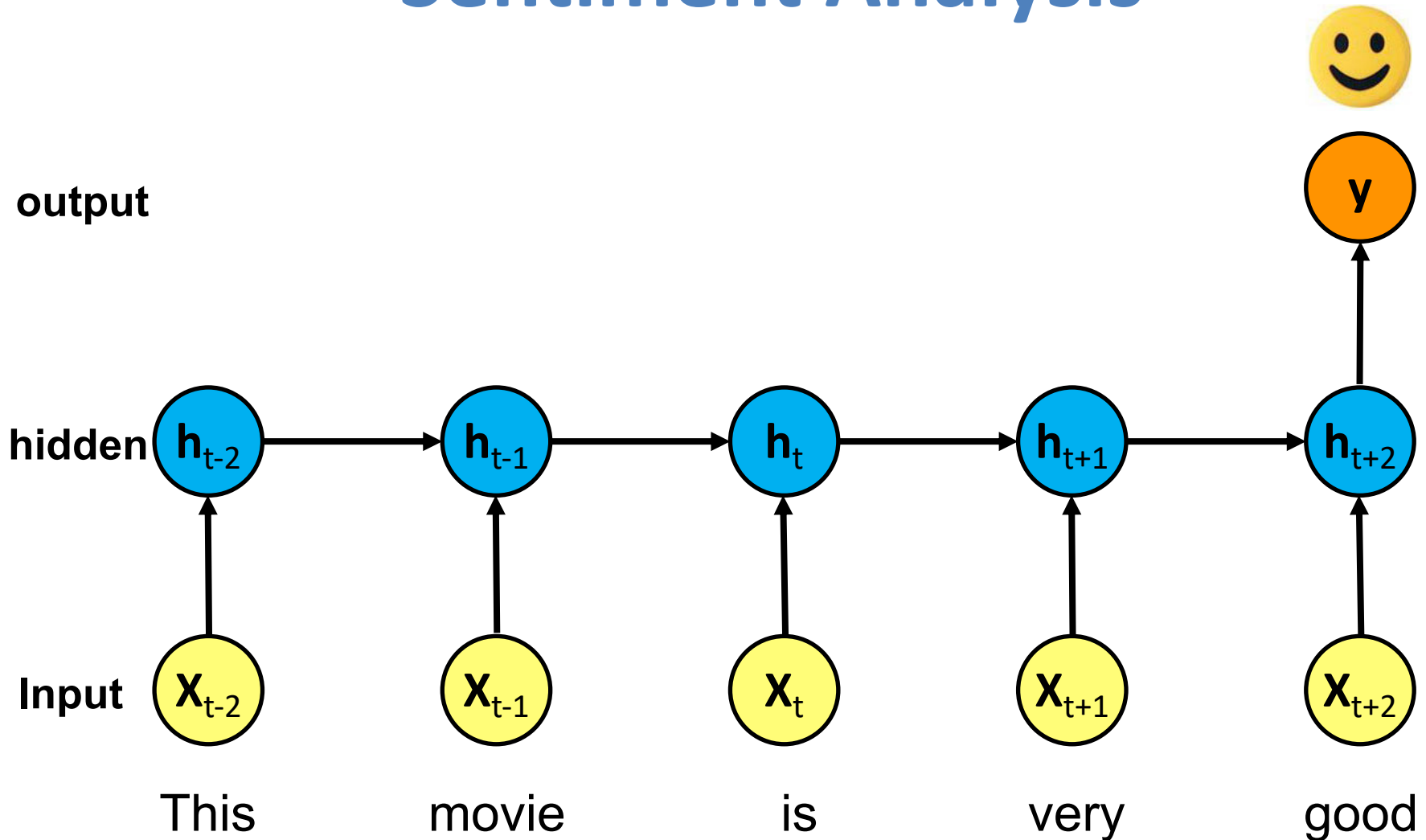
Recurrent Neural Networks (RNN)

Time Series Forecasting



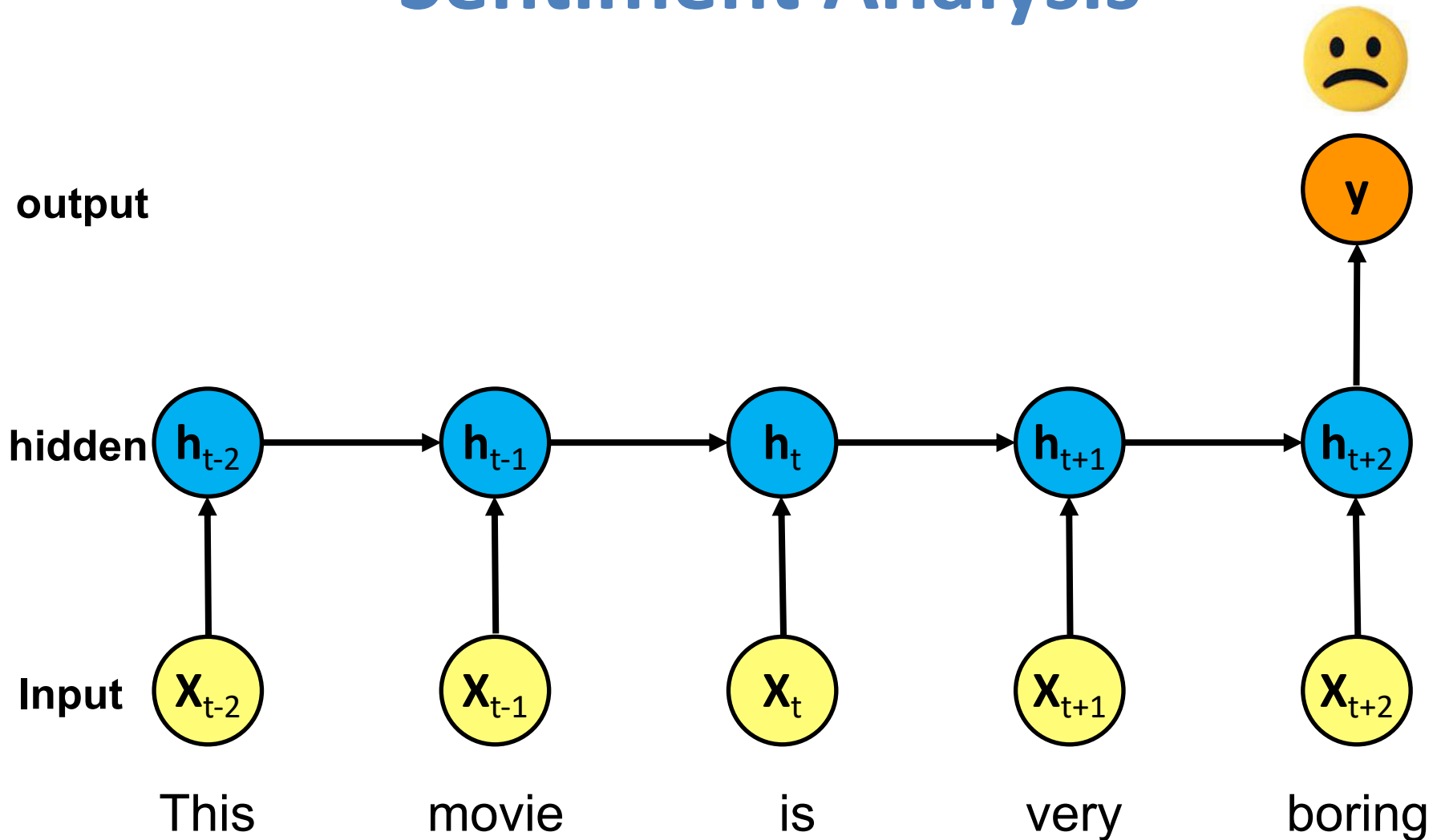
Recurrent Neural Networks (RNN)

Sentiment Analysis

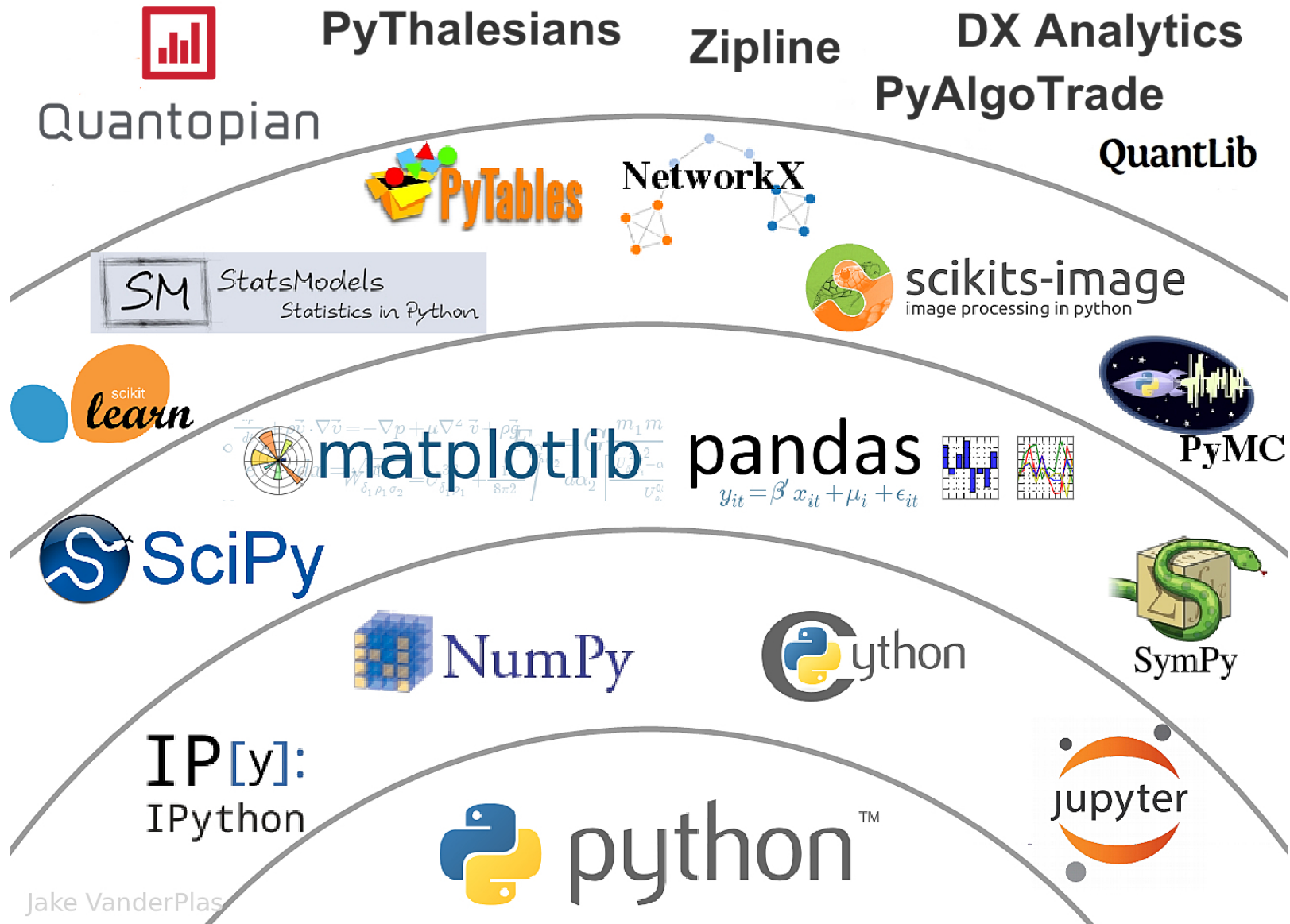


Recurrent Neural Networks (RNN)

Sentiment Analysis



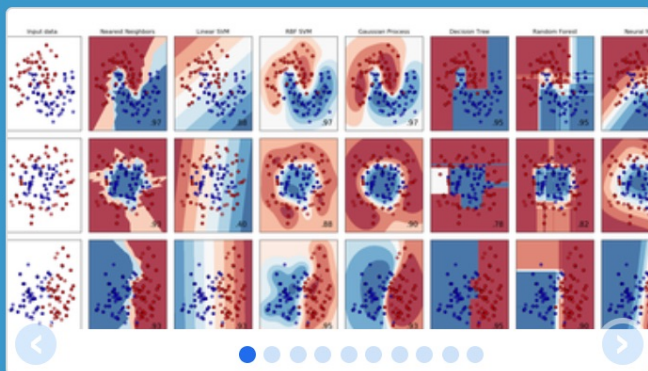
The Quant Finance PyData Stack





Python

Scikit-Learn

[Home](#)[Installation](#)[Documentation](#)[Examples](#)

scikit-learn

Machine Learning in Python

- Simple and efficient tools for data mining and data analysis
- Accessible to everybody, and reusable in various contexts
- Built on NumPy, SciPy, and matplotlib
- Open source, commercially usable - BSD license

Classification

Identifying to which category an object belongs to.

Applications: Spam detection, Image recognition.

Algorithms: SVM, nearest neighbors, random forest, ... — Examples

Regression

Predicting a continuous-valued attribute associated with an object.

Applications: Drug response, Stock prices.

Algorithms: SVR, ridge regression, Lasso, ... — Examples

Clustering

Automatic grouping of similar objects into sets.

Applications: Customer segmentation, Grouping experiment outcomes

Algorithms: k-Means, spectral clustering, mean-shift, ... — Examples

Dimensionality reduction

Reducing the number of random variables to consider.

Applications: Visualization, Increased efficiency

Algorithms: PCA, feature selection, non-negative matrix factorization. — Examples

Model selection

Comparing, validating and choosing parameters and models.

Goal: Improved accuracy via parameter tuning

Modules: grid search, cross validation, metrics. — Examples

Preprocessing

Feature extraction and normalization.

Application: Transforming input data such as text for use with machine learning algorithms.

Modules: preprocessing, feature extraction. — Examples

Google TensorFlow

TensorFlow™

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Community



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GITHUB

An open source machine learning
framework for everyone

GET STARTED



TensorFlow Dev Summit 2019

The 2019 TensorFlow Dev Summit is back March 6-7! Space is limited - request an invite to stay up to date.



TensorFlow 1.12 is here!

TensorFlow 1.12 is available, see the release notes for the latest updates.



High-level APIs in TensorFlow 2.0

By using Keras as the high-level API for the upcoming TensorFlow 2.0 release, we will make it easier for developers new to machine learning to get started while providing advanced capabilities for researchers.

<https://www.tensorflow.org/>

Google TensorFlow

TensorFlow™

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<https://www.tensorflow.org/>

Google Dopamine



Dopamine is a research framework
for fast prototyping of
reinforcement learning algorithms.

PyTorch

[Get Started](#)[Features](#)[Ecosystem](#)[Blog](#)[Tutorials](#)[Docs](#)[Resources](#)[GitHub](#)

FROM RESEARCH TO PRODUCTION

An open source deep learning platform that provides a seamless path from research prototyping to production deployment.

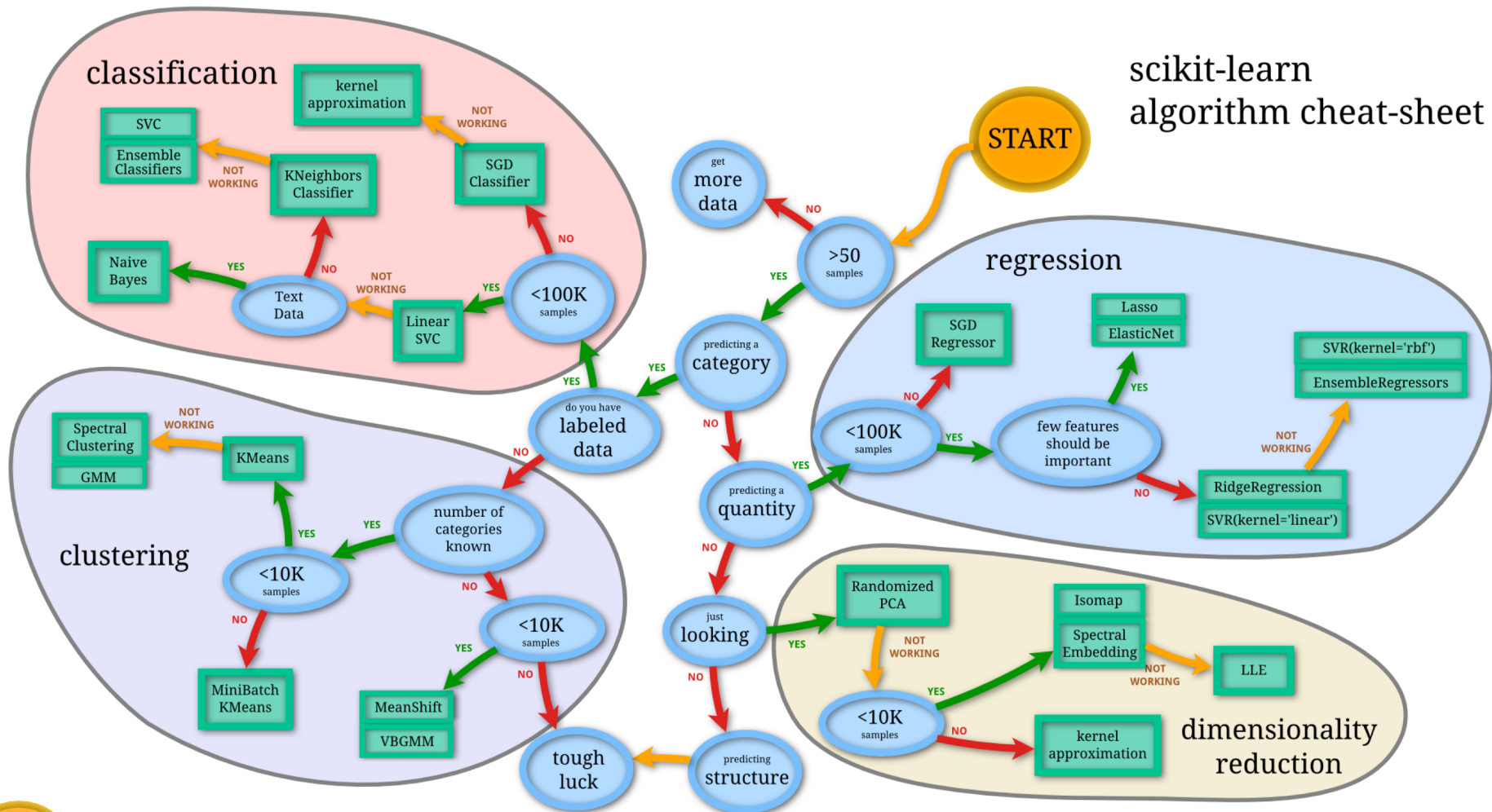
[Get Started >](#)

KEY FEATURES &
CAPABILITIES

[See all Features >](#)

<http://pytorch.org/>

Scikit-Learn Machine Learning Map



Iris flower data set

setosa



versicolor



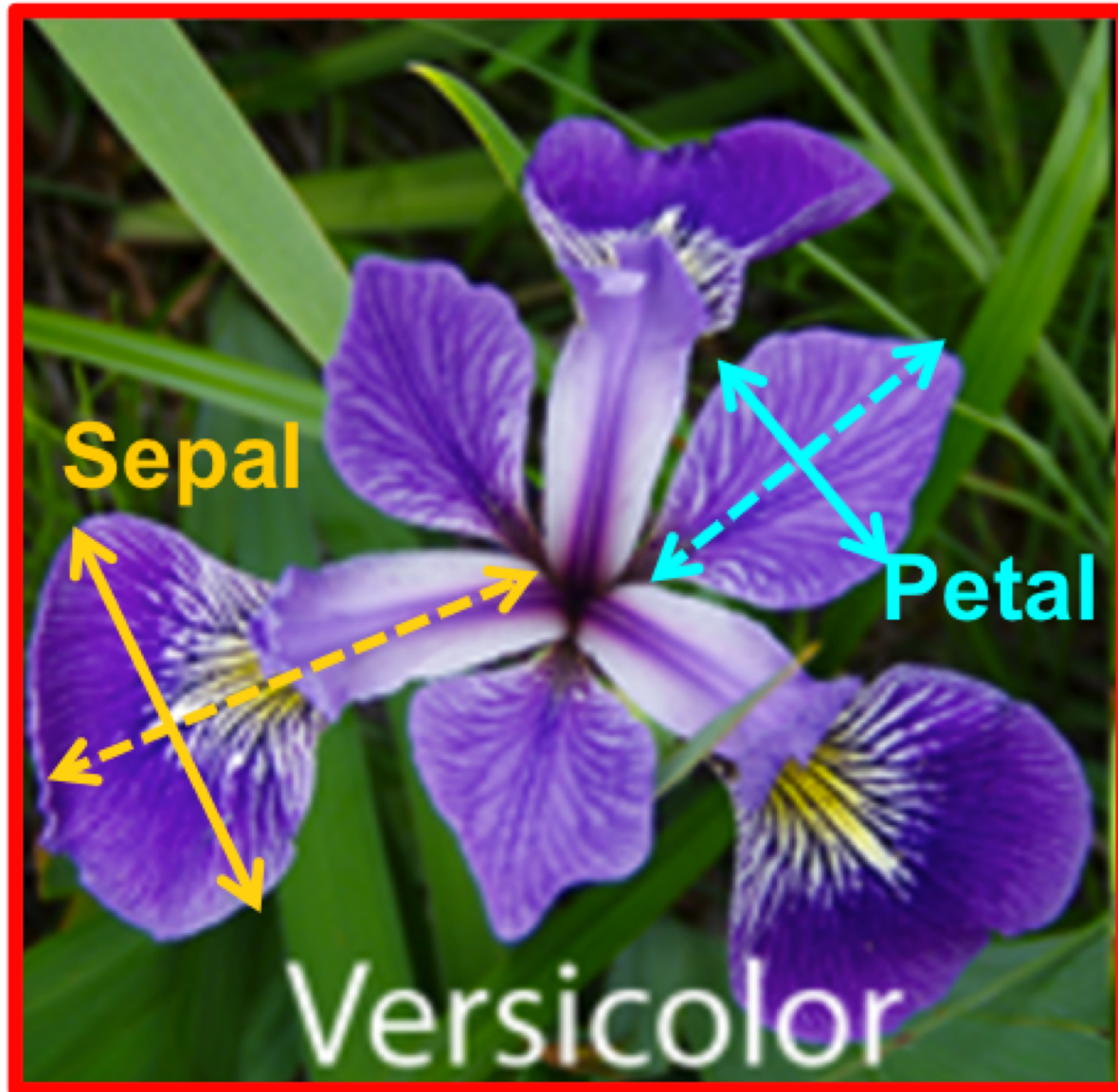
virginica



Source: https://en.wikipedia.org/wiki/Iris_flower_data_set

Source: <http://suruchifialoke.com/2016-10-13-machine-learning-tutorial-iris-classification/>

Iris Classification



iris.data

<https://archive.ics.uci.edu/ml/machine-learning-databases/iris/iris.data>

5.1,3.5,1.4,0.2,Iris-setosa
4.9,3.0,1.4,0.2,Iris-setosa
4.7,3.2,1.3,0.2,Iris-setosa
4.6,3.1,1.5,0.2,Iris-setosa
5.0,3.6,1.4,0.2,Iris-setosa
5.4,3.9,1.7,0.4,Iris-setosa
4.6,3.4,1.4,0.3,Iris-setosa
5.0,3.4,1.5,0.2,Iris-setosa
4.4,2.9,1.4,0.2,Iris-setosa
4.9,3.1,1.5,0.1,Iris-setosa
5.4,3.7,1.5,0.2,Iris-setosa
4.8,3.4,1.6,0.2,Iris-setosa
4.8,3.0,1.4,0.1,Iris-setosa
4.3,3.0,1.1,0.1,Iris-setosa
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5.7,4.4,1.5,0.4,Iris-setosa
5.4,3.9,1.3,0.4,Iris-setosa
5.1,3.5,1.4,0.3,Iris-setosa
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setosa



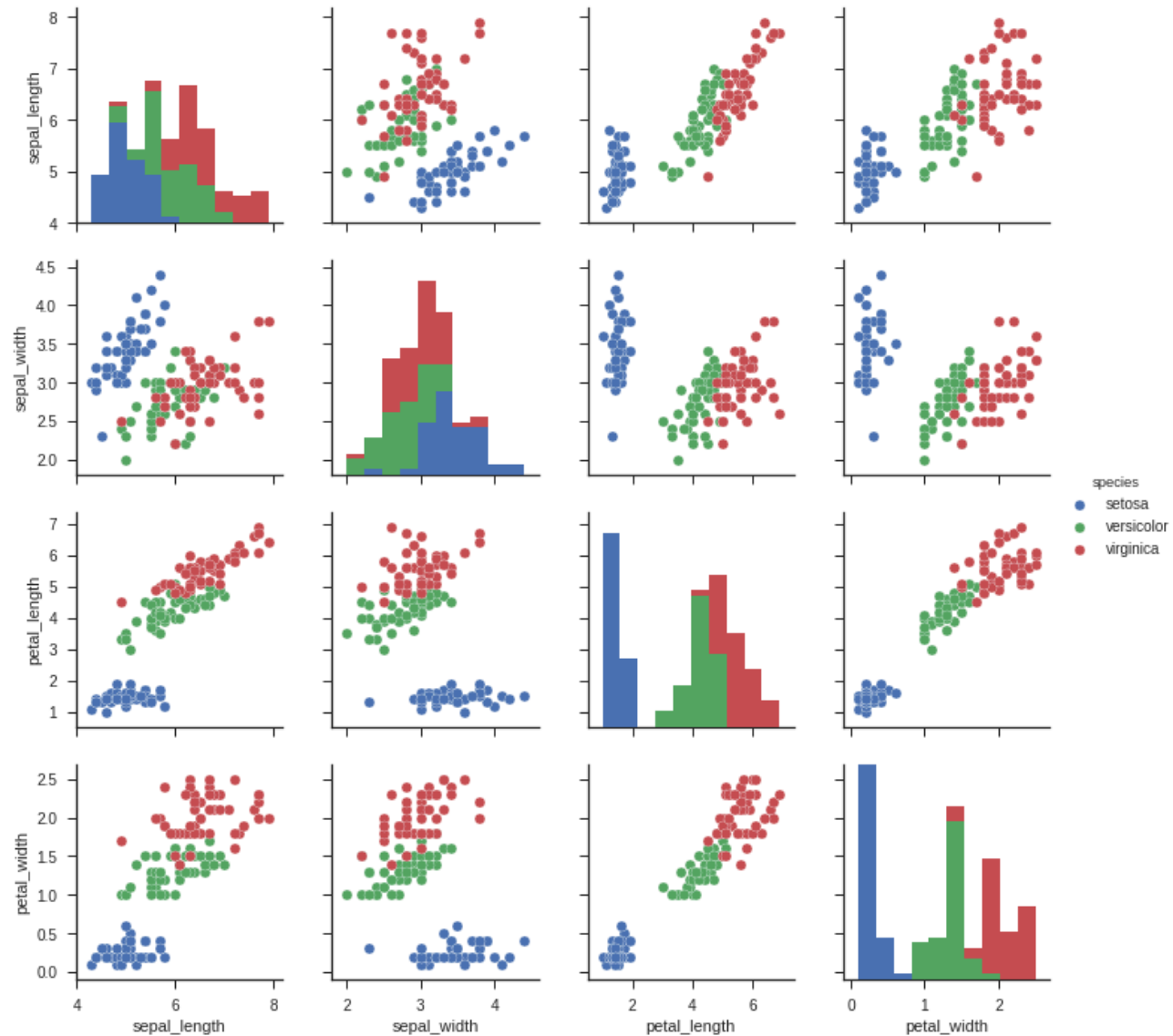
virginica



versicolor

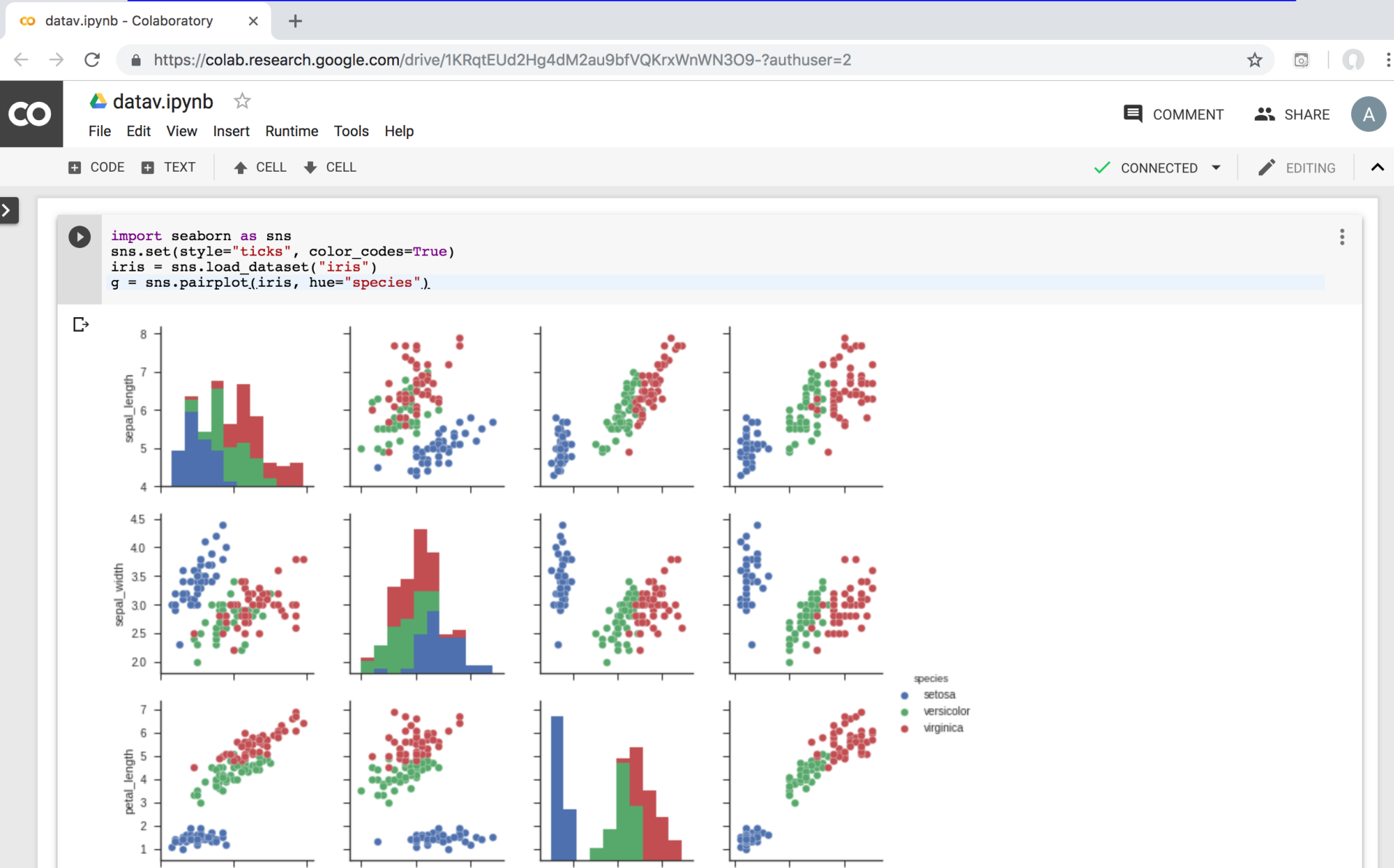


Iris Data Visualization



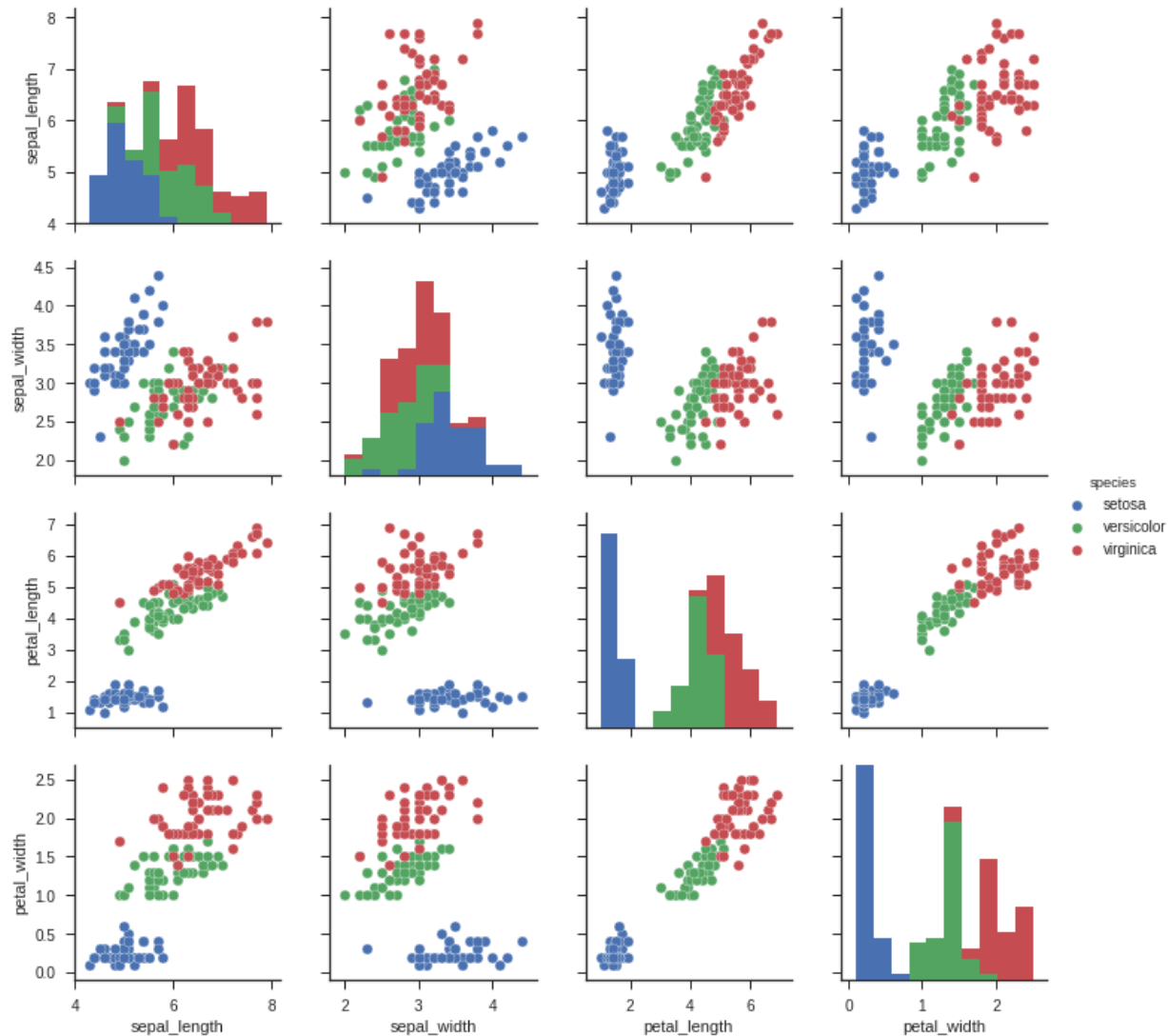
Data Visualization in Google Colab

<https://colab.research.google.com/drive/1KRqtEUd2Hg4dM2au9bfVQKrxWnWN3O9->



Source: <https://seaborn.pydata.org/generated/seaborn.pairplot.html>

```
import seaborn as sns
sns.set(style="ticks", color_codes=True)
iris = sns.load_dataset("iris")
g = sns.pairplot(iris, hue="species")
```



Machine Learning

Supervised Learning

Classification


and

Prediction

Classification and Prediction

https://colab.research.google.com/drive/1QE7fR2OxHiQ0_p6l1nnZDIFF354Nf_Lw

← → ↻ https://colab.research.google.com/drive/1QE7fR2OxHiQ0_p6l1nnZDIFF354Nf_Lw?authuser=2#scrollTo=qlwuq9m0ESyS ☆

 Classification_Prediction.ipynb ☆

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Data Mining and Machine Learning in Google Colab

```
[17] 1 # Import libraries
      2 import numpy as np
      3 import pandas as pd
      4 %matplotlib inline
      5 import matplotlib.pyplot as plt
      6 import seaborn as sns
      7 from pandas.plotting import scatter_matrix
      8
      9 # Import sklearn
     10 from sklearn import model_selection
     11 from sklearn.metrics import classification_report
     12 from sklearn.metrics import confusion_matrix
     13 from sklearn.metrics import accuracy_score
     14 from sklearn.linear_model import LogisticRegression
     15 from sklearn.tree import DecisionTreeClassifier
     16 from sklearn.neighbors import KNeighborsClassifier
     17 from sklearn.discriminant_analysis import LinearDiscriminantAnalysis
     18 from sklearn.naive_bayes import GaussianNB
     19 from sklearn.svm import SVC
     20 from sklearn.neural_network import MLPClassifier
     21 print("Imported")
     22
     23 # Load dataset
     24 url = "https://archive.ics.uci.edu/ml/machine-learning-databases/iris/iris.data"
     25 names = ['sepal-length', 'sepal-width', 'petal-length', 'petal-width', 'class']
     26 df = pd.read_csv(url, names=names)
     27
     28 print(df.head(10))
     29 print(df.tail(10))
     30 print(df.describe())
     31 print(df.info())
     32 print(df.shape)
     33 print(df.groupby('class').size())
     34
     35 plt.rcParams["figure.figsize"] = (10,8)
     36 df.plot(kind='box', subplots=True, layout=(2,2), sharex=False, sharey=False)
     37 plt.show()
     38
     39 df.hist()
     40 plt.show()
```

https://colab.research.google.com/drive/1QE7fR2OxHiQ0_p6l1nnZDIFF354Nf_Lw

```
# Make predictions on validation dataset
model = SVC()
model.fit(X_train, Y_train)
predictions = model.predict(X_validation)
print("%.4f" % accuracy_score(Y_validation,
predictions))
print(confusion_matrix(Y_validation,
predictions))
print(classification_report(Y_validation,
predictions))
print(model)
```

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model = SVC()
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```

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7 print(classification_report(Y_validation, predictions))
8 print(model)
```

0.9333

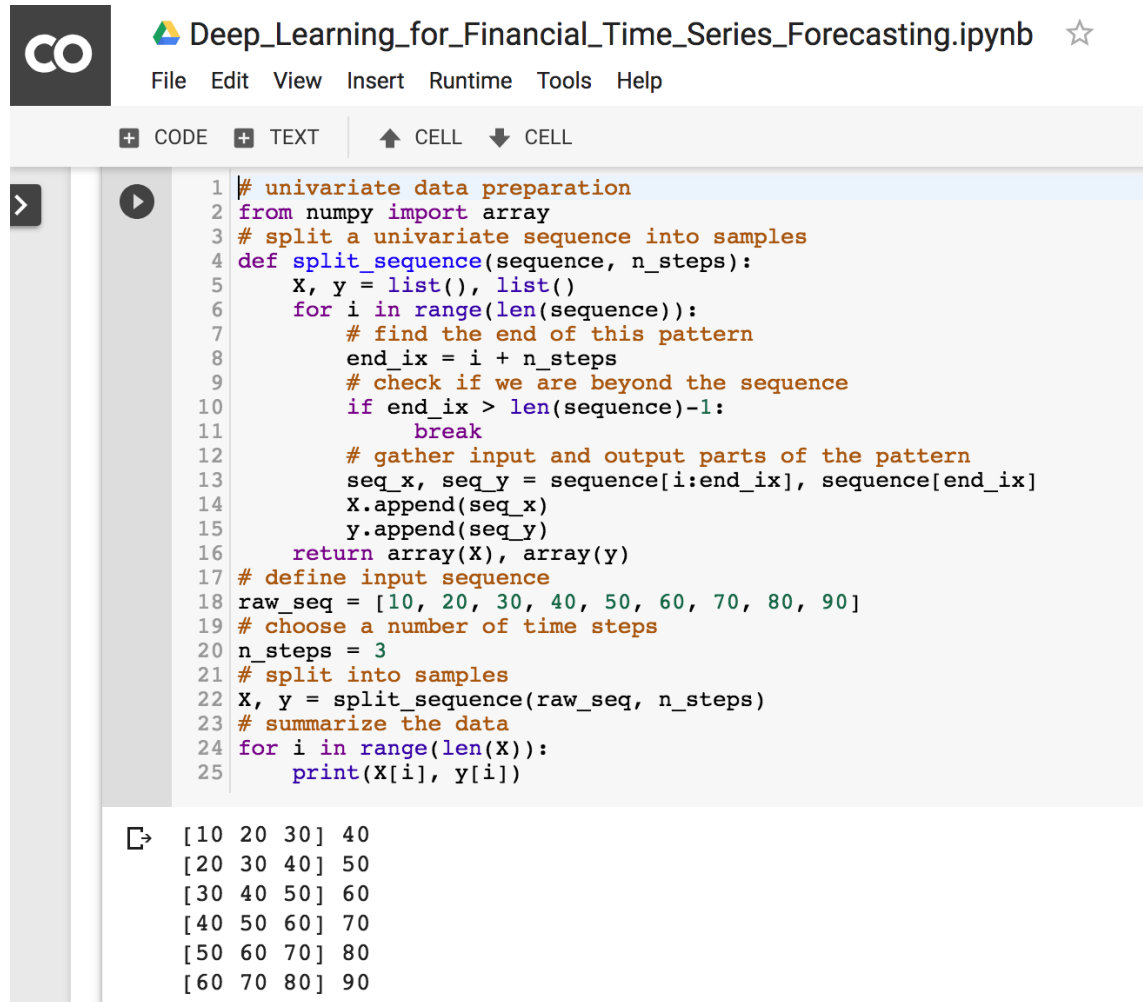
```
[[ 7  0  0]
 [ 0 10  2]
 [ 0  0 11]]
```

	precision	recall	f1-score	support
Iris-setosa	1.00	1.00	1.00	7
Iris-versicolor	1.00	0.83	0.91	12
Iris-virginica	0.85	1.00	0.92	11
avg / total	0.94	0.93	0.93	30

```
SVC(C=1.0, cache_size=200, class_weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='auto', kernel='rbf',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
    tol=0.001, verbose=False)
```

Deep Learning for Financial Time Series Forecasting

<https://colab.research.google.com/drive/1aEK0eSev8Q-Y0nNY32geFk7CB8pVgSQM>



The image shows a Google Colab notebook titled "Deep_Learning_for_Financial_Time_Series_Forecasting.ipynb". The notebook contains a single code cell with the following Python code:


```
1 # univariate data preparation
2 from numpy import array
3 # split a univariate sequence into samples
4 def split_sequence(sequence, n_steps):
5     X, y = list(), list()
6     for i in range(len(sequence)):
7         # find the end of this pattern
8         end_ix = i + n_steps
9         # check if we are beyond the sequence
10        if end_ix > len(sequence)-1:
11            break
12        # gather input and output parts of the pattern
13        seq_x, seq_y = sequence[i:end_ix], sequence[end_ix]
14        X.append(seq_x)
15        y.append(seq_y)
16    return array(X), array(y)
17 # define input sequence
18 raw_seq = [10, 20, 30, 40, 50, 60, 70, 80, 90]
19 # choose a number of time steps
20 n_steps = 3
21 # split into samples
22 X, y = split_sequence(raw_seq, n_steps)
23 # summarize the data
24 for i in range(len(X)):
25     print(X[i], y[i])
```

The output of the code cell is:

```
[10 20 30] 40
[20 30 40] 50
[30 40 50] 60
[40 50 60] 70
[50 60 70] 80
[60 70 80] 90
```


Deep Learning for Financial Time Series Forecasting

<https://colab.research.google.com/drive/1aEK0eSev8Q-Y0nNY32geFk7CB8pVgSQM>

 Deep_Learning_for_Financial_Time_Series_Forecasting.ipynb ☆

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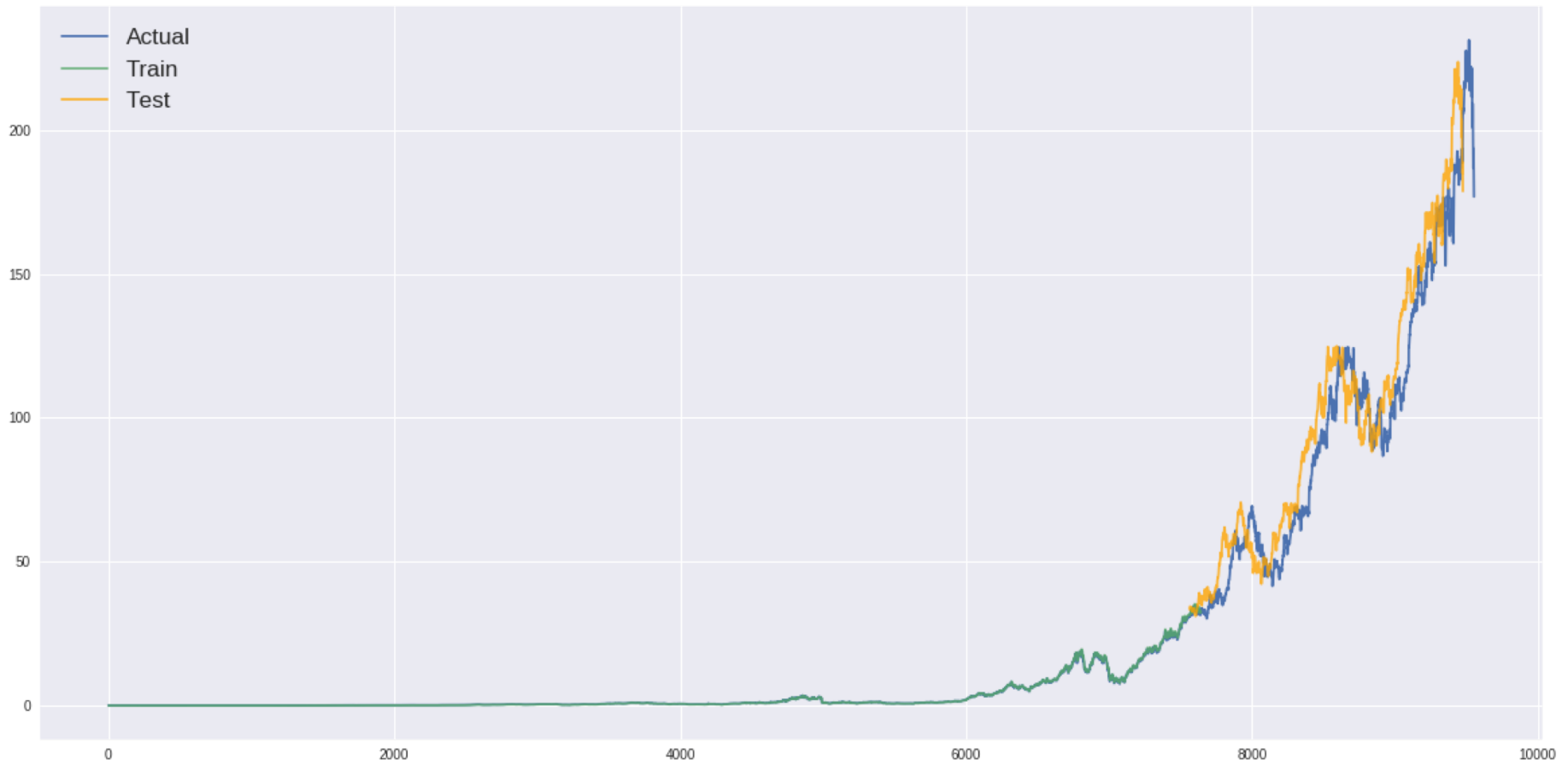
LSTM for Time Series Forecasting

```
1 # univariate lstm example
2 from numpy import array
3 from keras.models import Sequential
4 from keras.layers import LSTM
5 from keras.layers import Dense
6 import matplotlib.pyplot as plt
7 %matplotlib inline
8
9 # define dataset
10 x = array([[100, 110, 120], [110, 120, 130], [120, 130, 140], [130, 140, 150], [140, 150, 160]])
11 y = array([130, 140, 150, 160, 170])
12 # reshape from [samples, timesteps] into [samples, timesteps, features]
13 x = x.reshape((x.shape[0], x.shape[1], 1))
14 # define model
15 model = Sequential()
16 model.add(LSTM(50, activation='relu', input_shape=(3, 1)))
17 model.add(Dense(1))
18 model.compile(optimizer='adam', loss='mse')
19 # fit model
20 history = model.fit(x, y, epochs=2000, verbose=0)
21 # demonstrate prediction
22 x_input = array([150, 160, 170])
23 x_input = x_input.reshape((1, 3, 1))
24 yhat = model.predict(x_input, verbose=0)
25 print('yhat', yhat)
26 print(model.summary())
27 # list all data in history
28 print(history.history.keys())
29 # summarize history for loss
30 print('loss:', '%f'%history.history['loss'][-1])
31 print('loss:', history.history['loss'][-1])
32 plt.plot(history.history['loss'])
33 plt.title('model loss')
34 plt.ylabel('loss')
35 plt.xlabel('epoch')
36 plt.show()
```

yhat [[181.34615]]

Deep Learning for Financial Time Series Forecasting

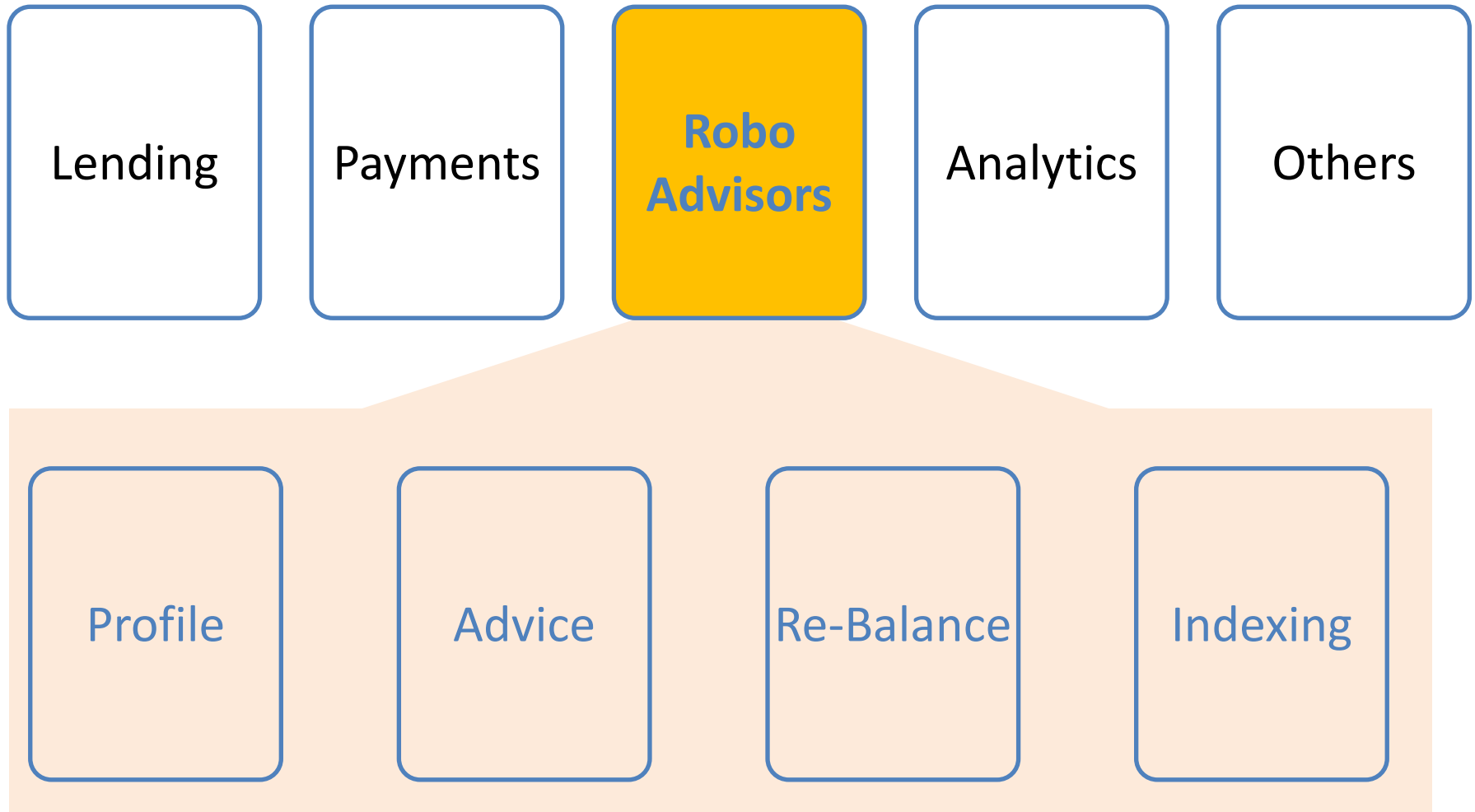
<https://colab.research.google.com/drive/1aEK0eSev8Q-Y0nNY32geFk7CB8pVgSQM>



AI in FinTech

Robo-Advisors

FinTech high-level classification



From Algorithmic Trading to Personal Finance Bots: 41 Startups Bringing AI to Fintech

From Algorithmic Trading To Personal Finance Bots: 41 Startups Bringing AI To Fintech

AI in Fintech

41 Startups Bringing Artificial Intelligence To Fintech

General Purpose/ Predictive Analytics

AYASDI  Digital Reasoning  context relevant™ 
  cortical.io  Numenta 
 DataRobot 

Market Research & Sentiment Analysis

 indico  acuity TRADING  Lucena Quantitative Analytics 


Search Engine

 alphasense

Quantitative Trading

 sentient technologies
 CLONE ALGO
 Alpaca
 WALNUT ALGORITHMS

Blockchain

 Skry
 EUKLID

Debt Collection

 TrueAccord

AI Assistants/Bots

 KASIST@
 TRIM
 Penny
 INSURIFY
 SURE.
 Cleo  FinGenius

Fraud Detection

 feedzai
 BIOCATCH
Less Friction. Less Fraud.

Credit Scoring

 TypeScore  aire
 creditvidya  zest finance
 ADF APPLIED DATA FINANCE
 Wecash 网银奇异
 CREAM FINANCE

Personal Banking

 personetics®
TAKING DIGITAL BANKING PERSONAL
 SBDA group

Artificial Intelligence (AI) in Fintech

General Purpose/ Predictive Analytics



Market Research & Sentiment Analysis



Search Engine



Artificial Intelligence (AI) in Fintech

Quantitative Trading



AI Assistants/Bots



Credit Scoring



Blockchain



Debt Collection



Fraud Detection



Personal Banking



FinTech

Financial Technology

FinTech

“providing
financial services
by making use of
software and
modern technology”

Financial Services

Financial Services



FinTech: Financial Services Innovation

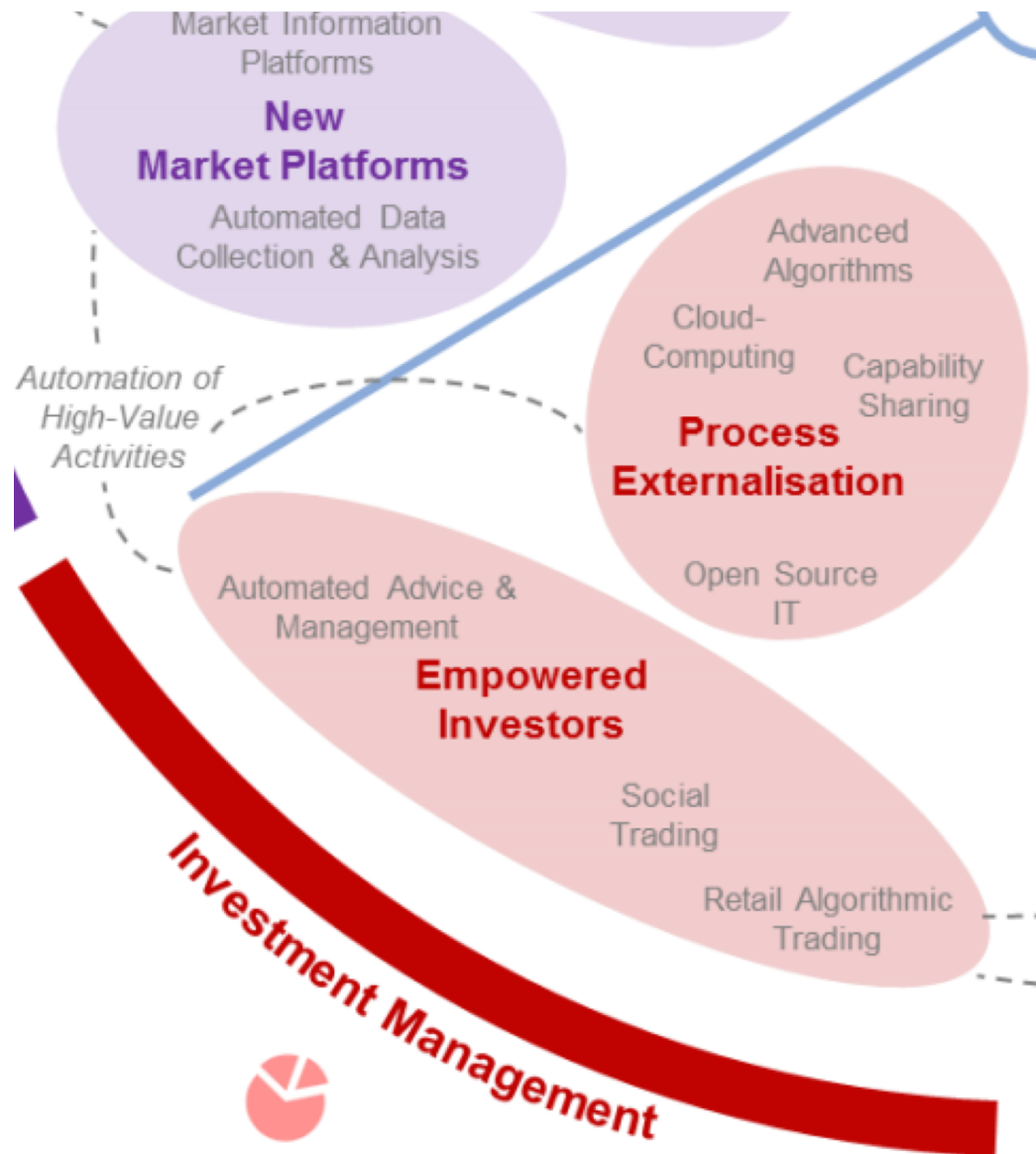


FinTech:

Financial Services Innovation

- 1. Payments**
- 2. Insurance**
- 3. Deposits & Lending**
- 4. Capital Raising**
- 5. Investment Management**
- 6. Market Provisioning**

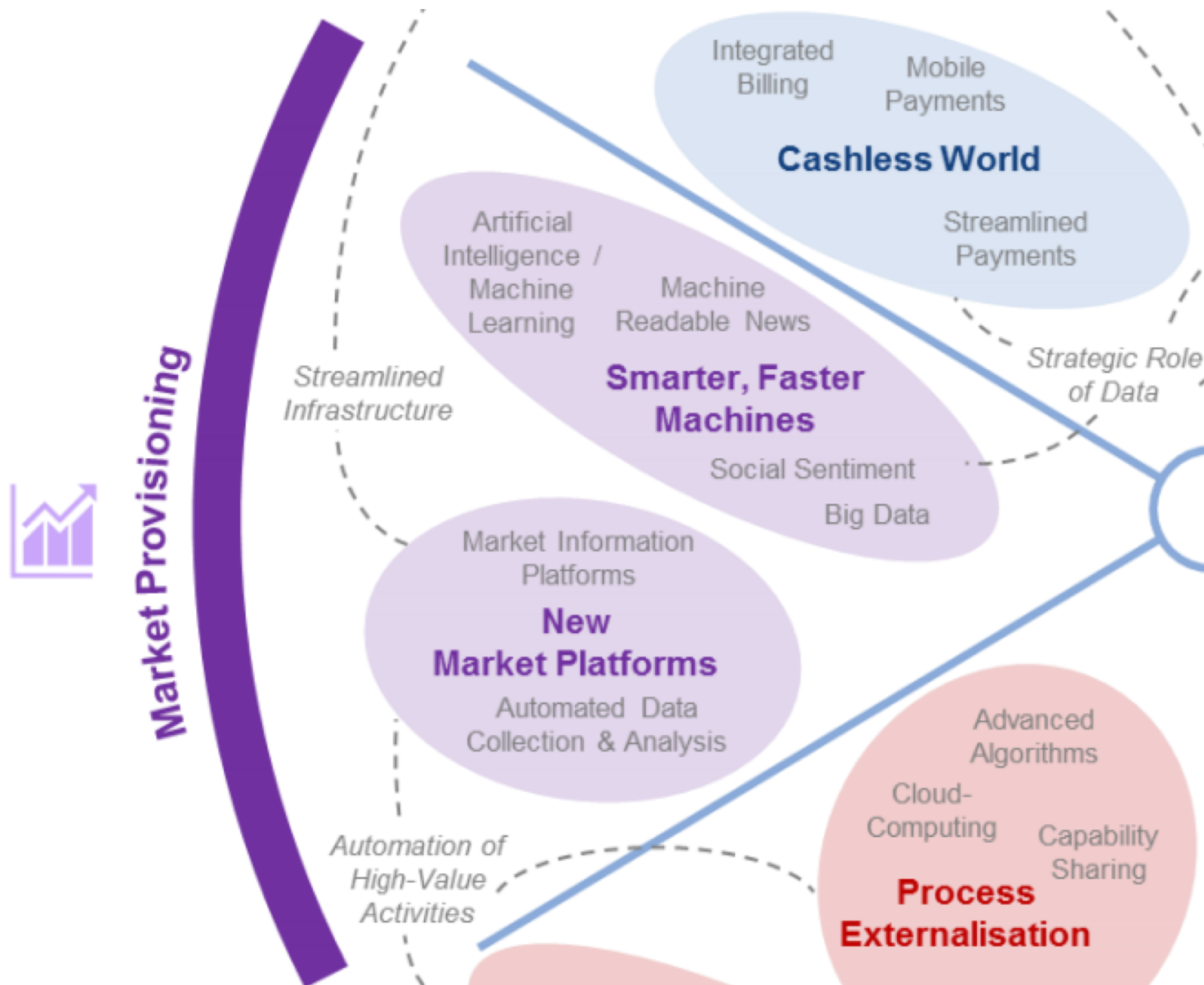
5 FinTech: Investment Management



5 FinTech: Investment Management Empowered Investors Process Externalization

6

FinTech: Market Provisioning



6

FinTech: Market Provisioning Smarter, Faster Machines New Market Platforms

The **New Alpha**: 30+ Startups Providing Alternative Data For Sophisticated Investors

New sources of **data mined** by startups like **Foursquare**, **Premise**, and **Orbital Insight** are letting investors understand **trends** before they happen.

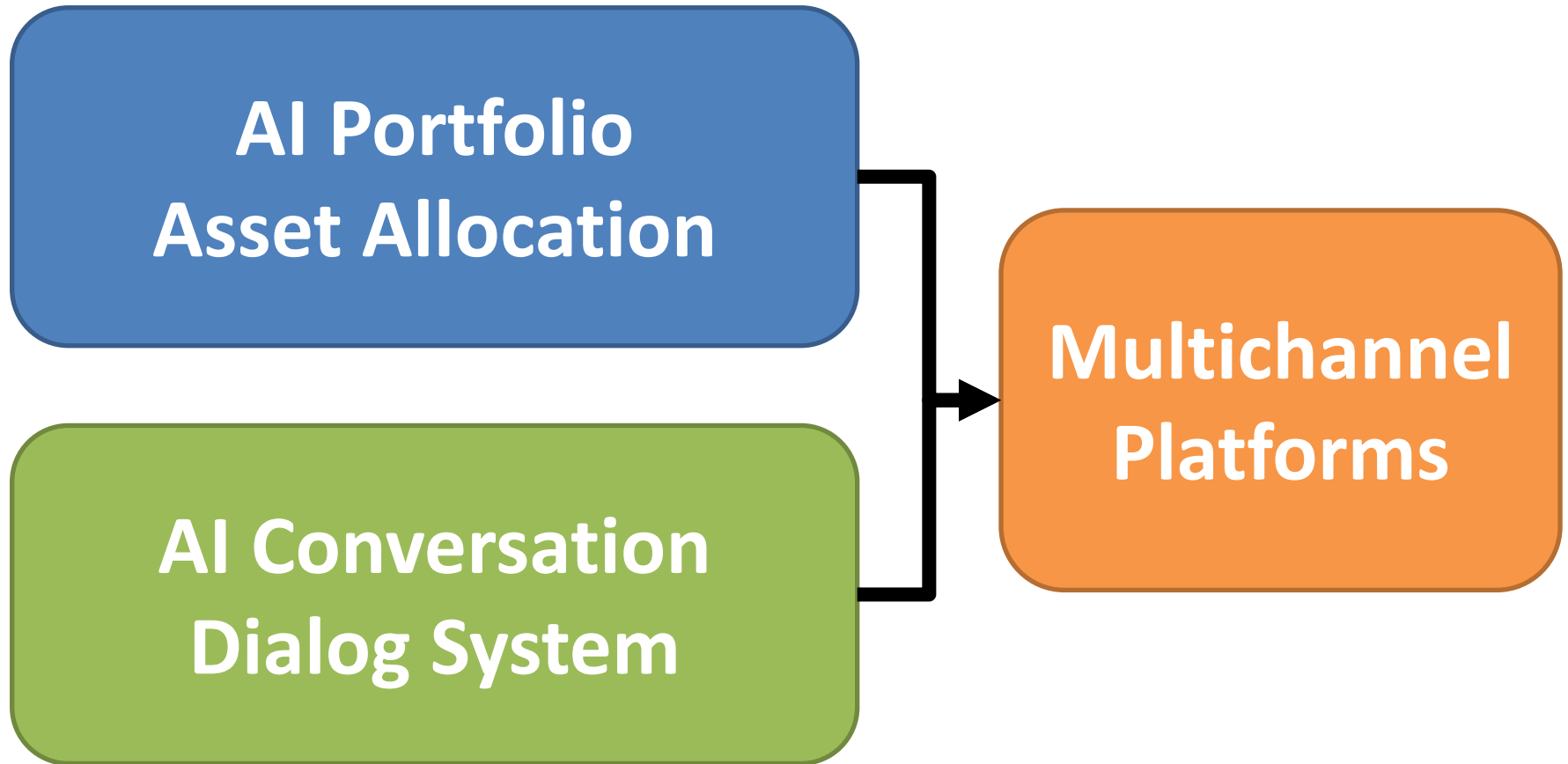
The New Alpha: 30+ Startups Providing Alternative Data For Sophisticated Investors

Alternative Data Sources



Artificial Intelligence for Conversational Robo-Advisor

AI Conversational Robo-Advisor



Portfolio Performance in 2016

Annual Portfolio Statistics

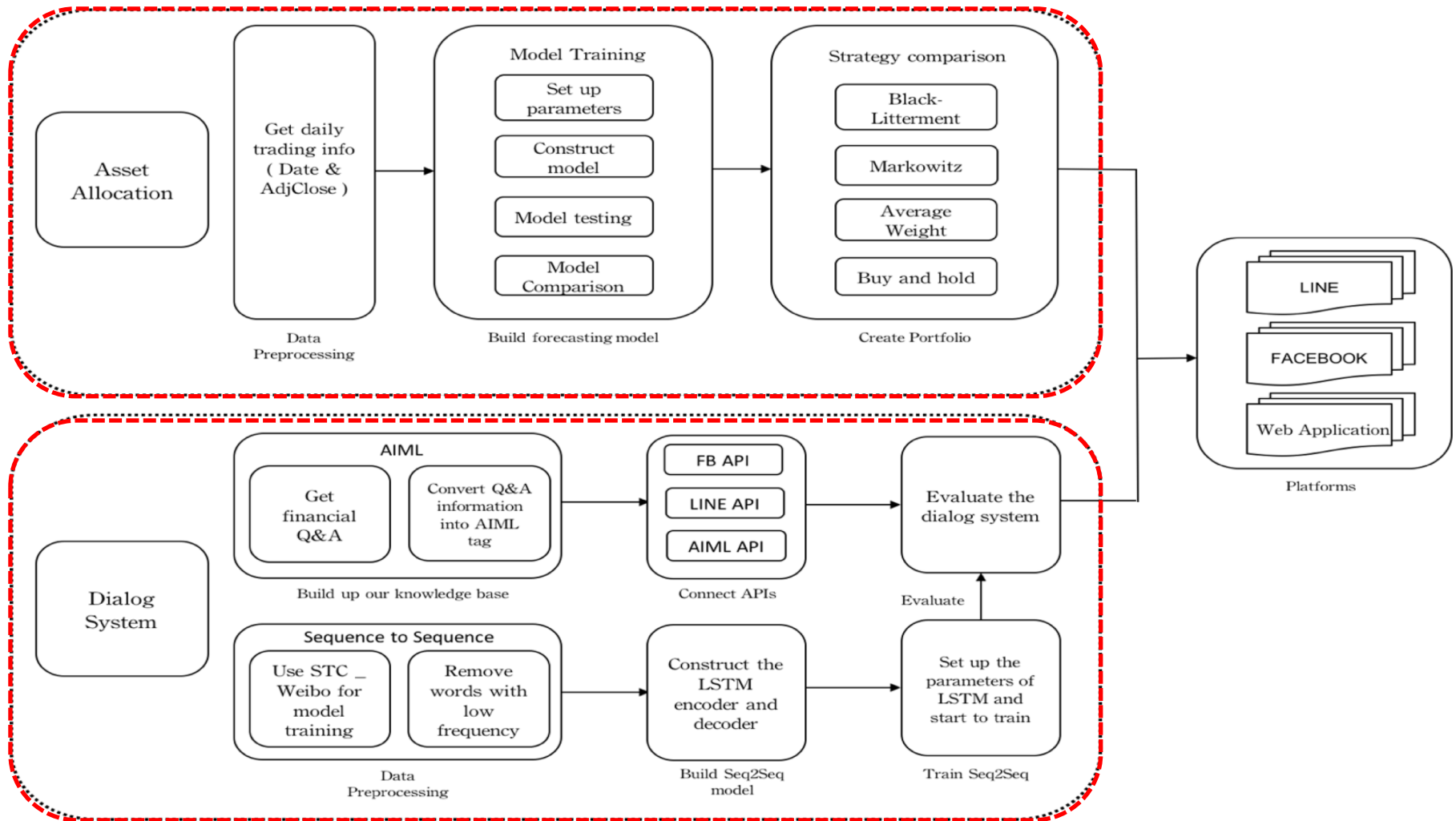
	Black-Litterman Portfolio - the LSTM Investor Views	Markowitz Portfolio	Equally Weighted Portfolio	S&P 500 Index
Annual return	16.151%	15.172%	12.428%	9.643%
Annual volatility	13.897%	14.365%	15.870%	13.169%
Sharpe ratio	1.14697	1.05534	0.81762	0.76492
Stability	0.82500	0.82515	0.82514	0.78754
Max drawdown	-10.105%	-10.465%	-12.529%	-10.306%
Skew	-0.35652	-0.52985	-0.56976	-0.36795
Kurtosis	2.49845	3.00613	2.41894	2.21958
Daily value at risk	-1.688%	-1.750%	-1.948%	-1.619%
Alpha	0.06445	0.05354	0.02158	0.00000
Beta	1.01485	1.04816	1.15631	1.00000
Information ratio	0.10935	0.09129	0.04655	-

Portfolio Cumulative Returns

Cumulative Returns: Portfolios

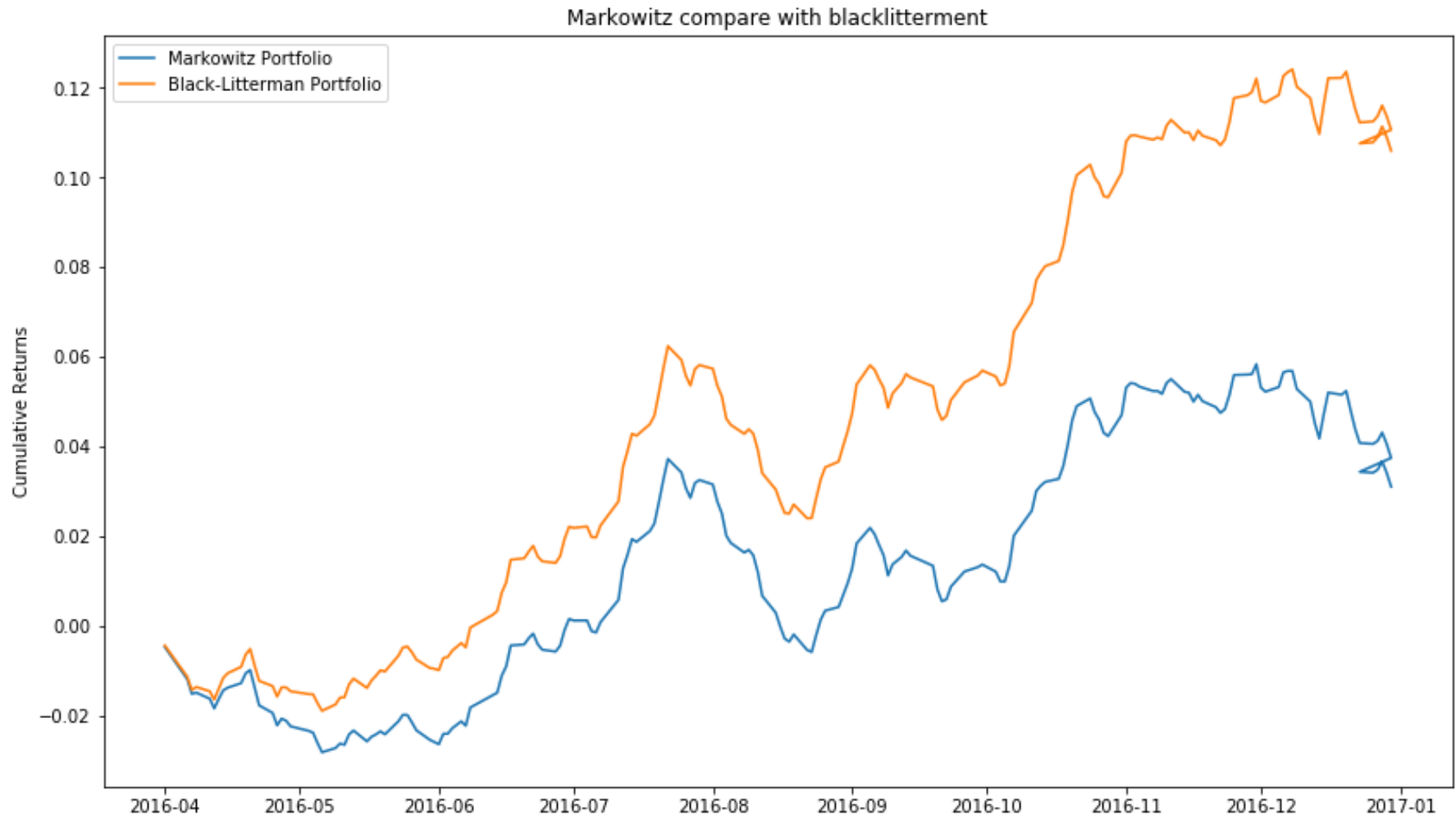


System Architecture of AI Conversational Robo-Advisor



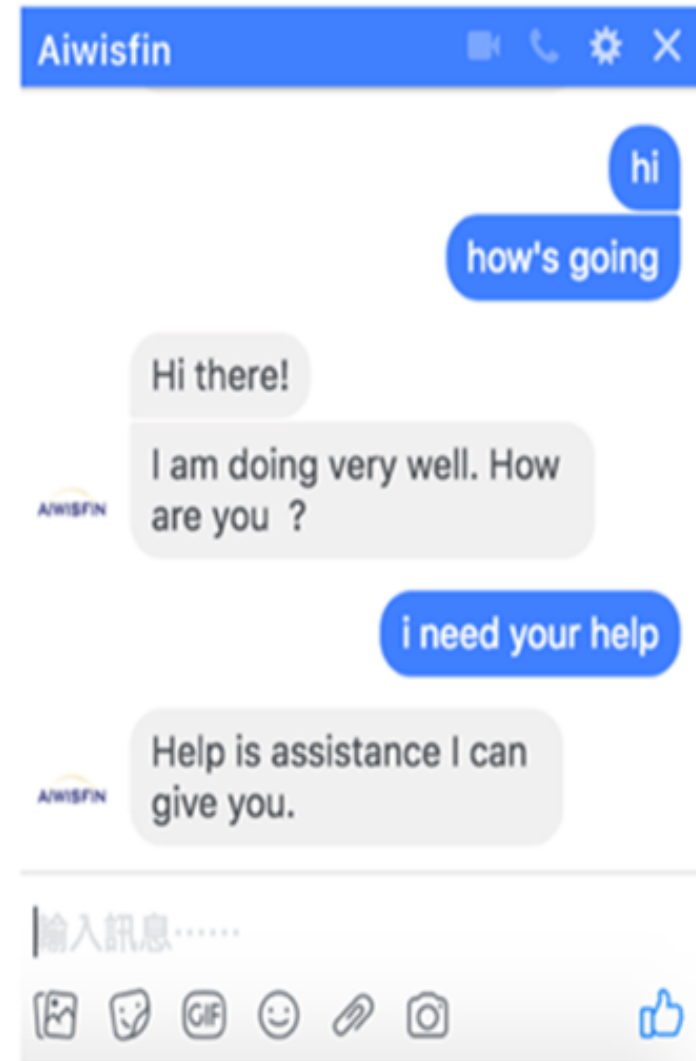
Cumulative Returns

Markowitz v.s. Black-litterment



Source: Min-Yuh Day, Jian-Ting Lin and Yuan-Chih Chen (2018), "Artificial Intelligence for Conversational Robo-Advisor", in Proceedings of the 2018 IEEE/ACM International Conference on Advances in Social Networks Analysis and Mining (ASONAM 2018), Barcelona, Spain, August 28-31, 2018

Conversational Model (LINE, FB Messenger)



Conversational Robo-Advisor

Multichannel UI/UX

Robots



ALPHA 2

ZENBO



AI Chatbot for Conversational Commerce

Chatbots: Evolution of UI/UX

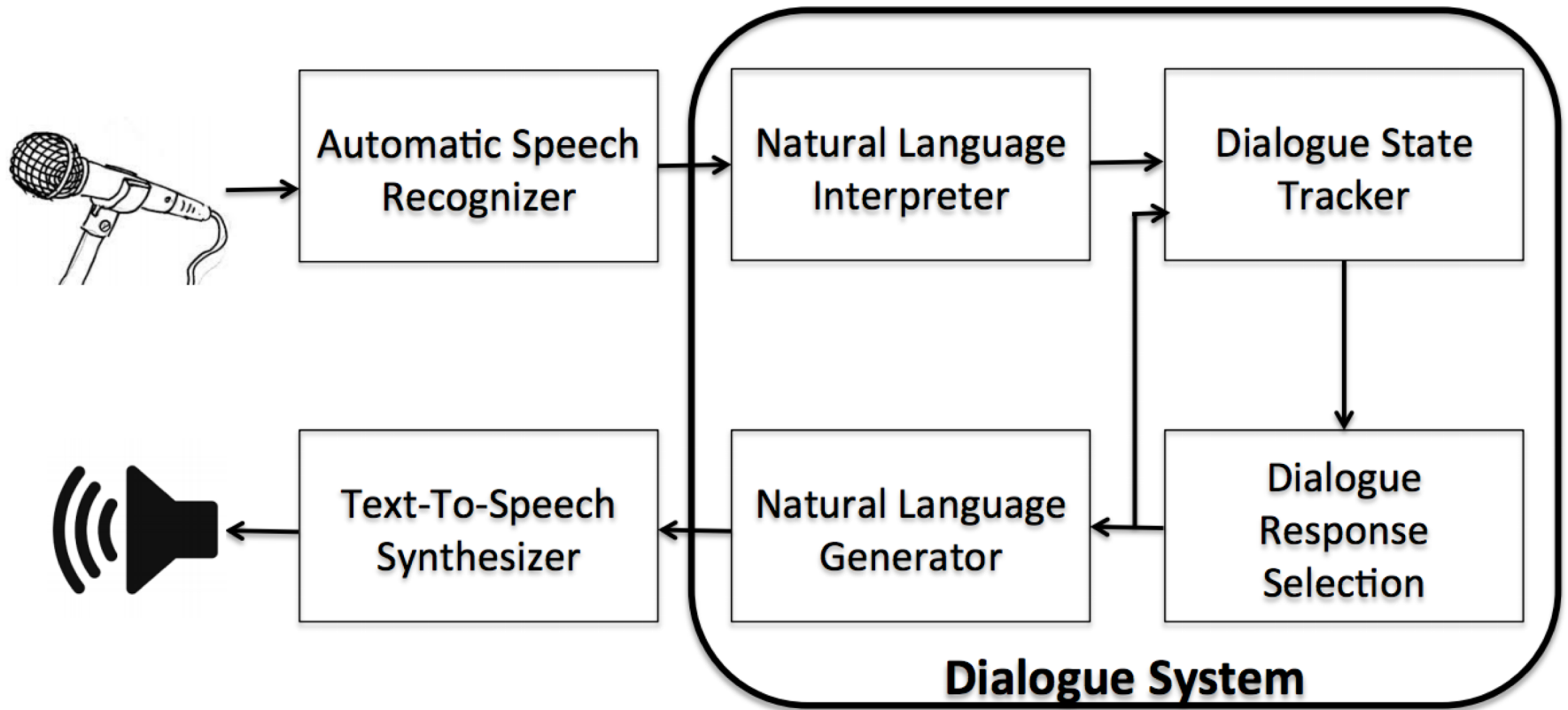
Paradigm	mid - 80s PC	mid - 90s Web	mid - 00s Smartphone	mid - 10s Messaging
Platform Examples	Desktop DOS, Windows, Mac OS	Browser Mosaic, Explorer, Chrome	Mobile OS iOS, Android	Messaging Apps WhatsApp, Messenger, Slack
Applications Examples	Clients Excel, PPT, Lotus	Website Yahoo, Amazon	Apps Angry Birds, Instagram	Bots Weather, Travel
UI/UX	Native Screens	Web Pages	Native Mobile Screens	Message
S/w Dev	Client-side	Server-side	Client-side	Server-side

Chatbot
Dialogue System
Intelligent Agent

Chatbot



Dialogue System



Can machines think?

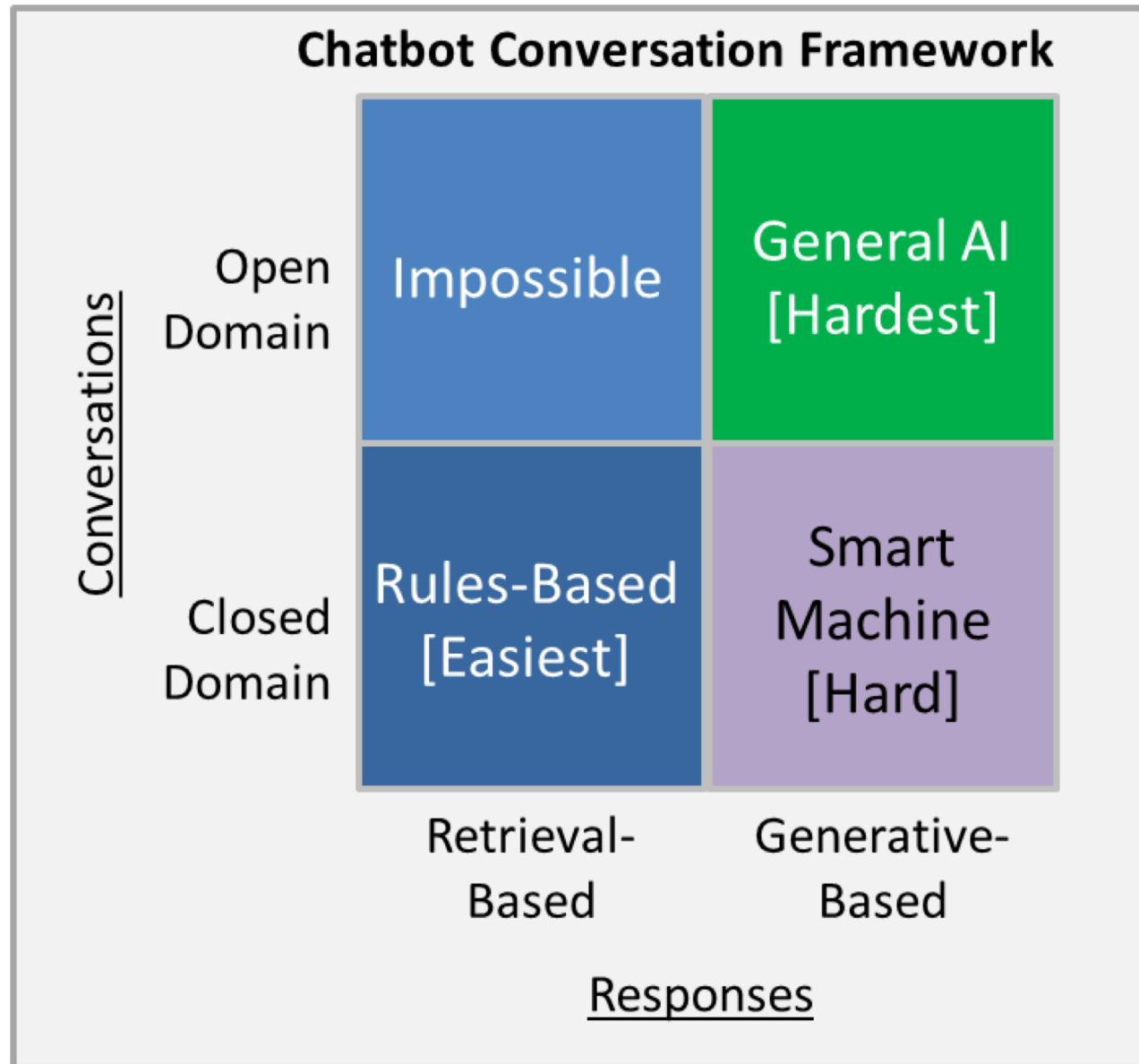
(Alan Turing ,1950)

Source: Cahn, Jack. "CHATBOT: Architecture, Design, & Development."
PhD diss., University of Pennsylvania, 2017.

Chatbot

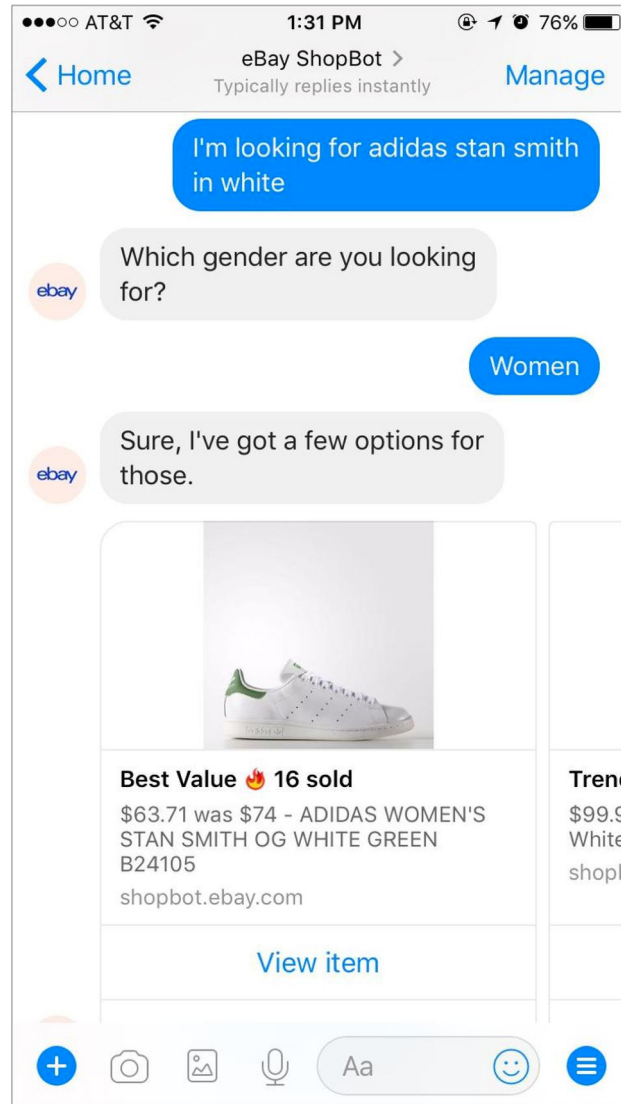
**“online human-computer
dialog system
with
natural language.”**

Chatbot Conversation Framework

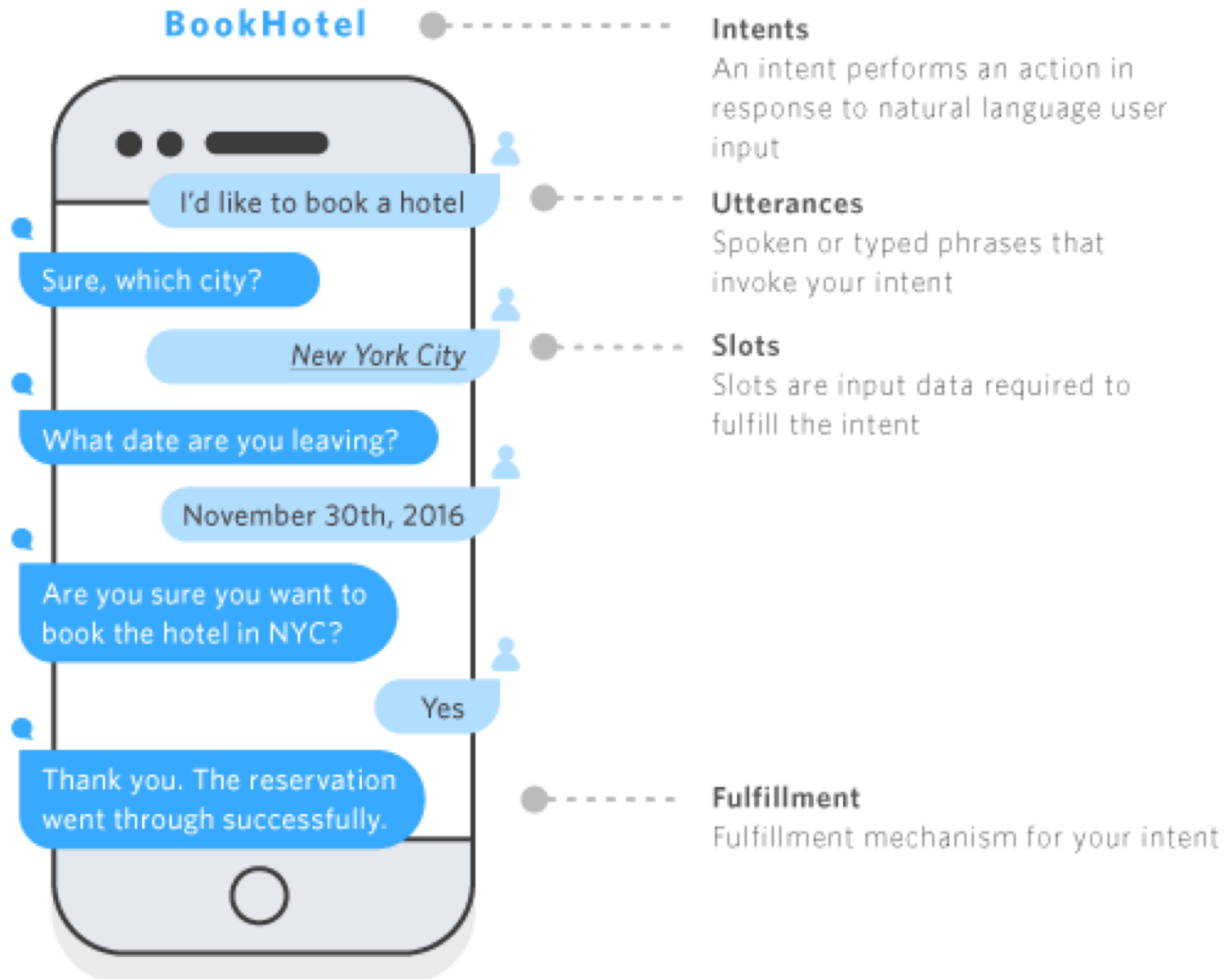


**From
E-Commerce
to
Conversational Commerce:
Chatbots
and
Virtual Assistants**

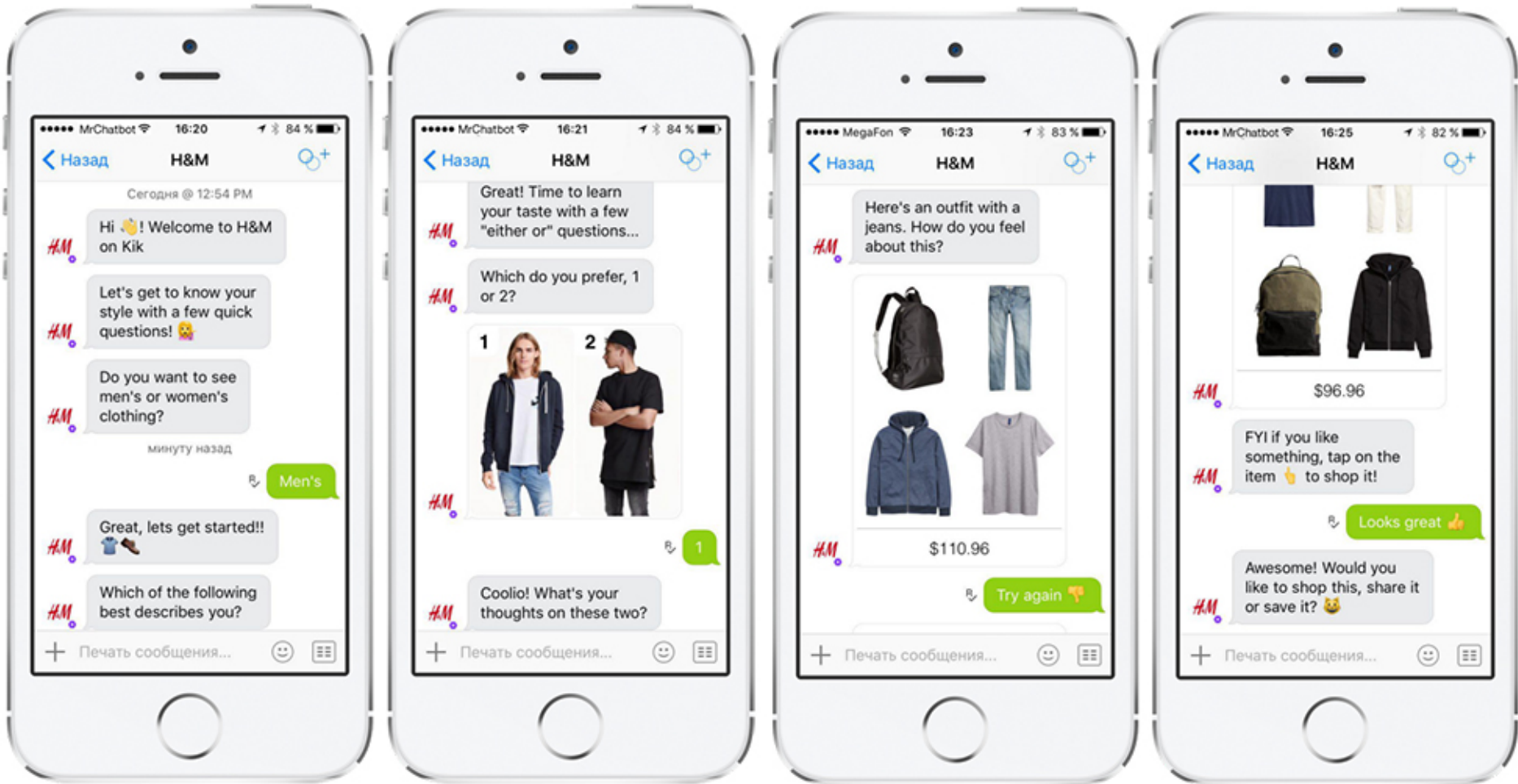
Conversational Commerce: eBay AI Chatbots



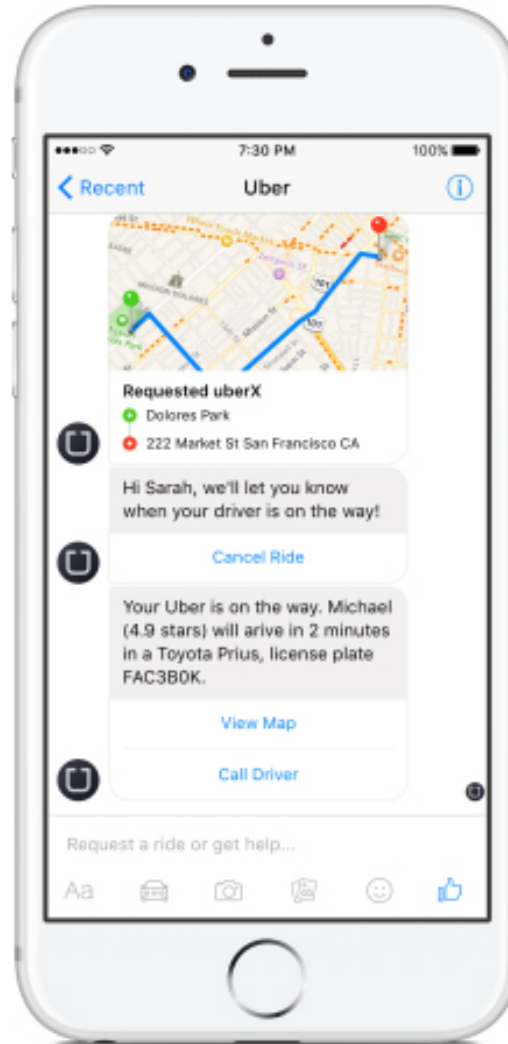
Hotel Chatbot



H&M's Chatbot on Kik



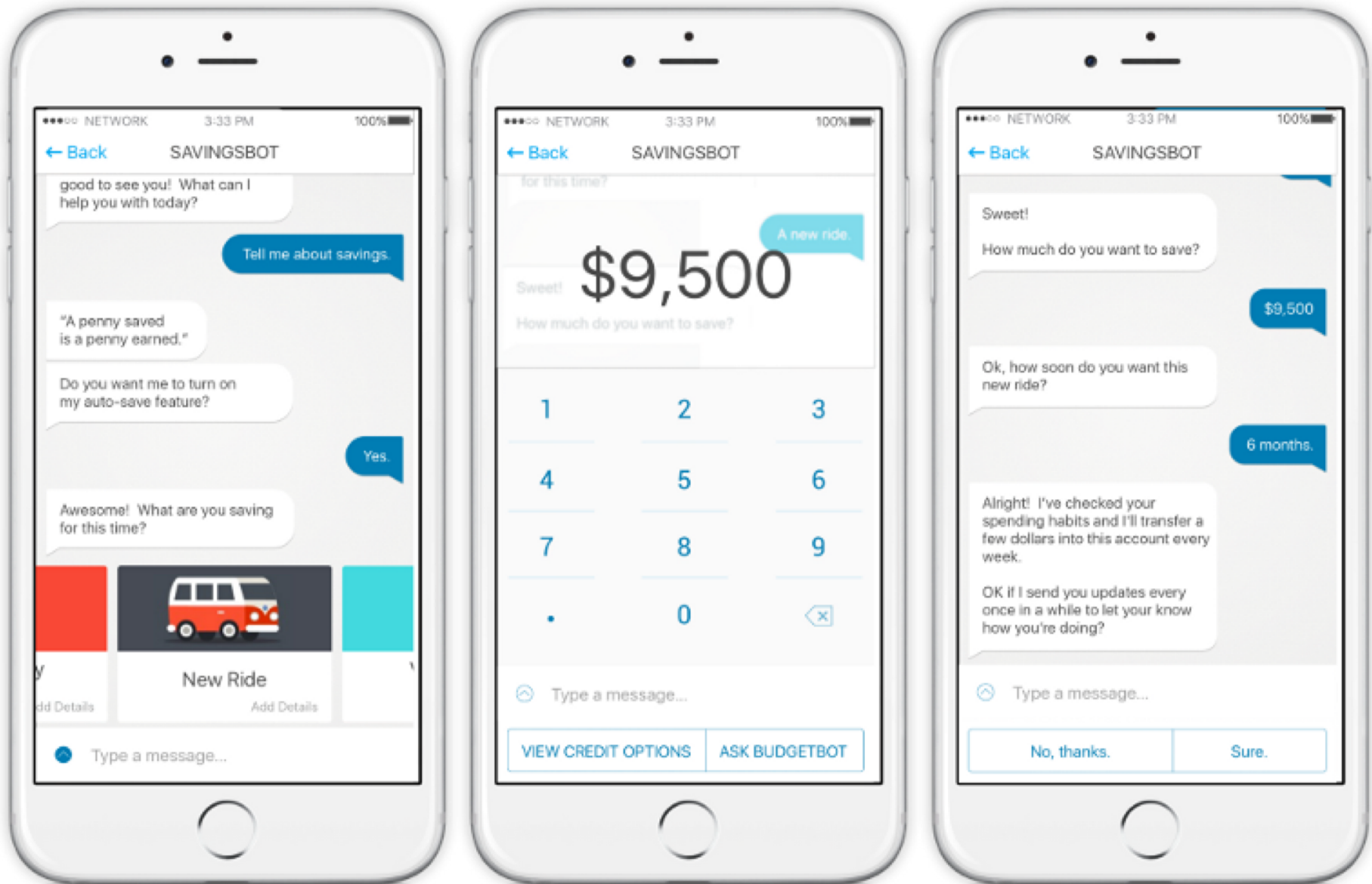
Uber's Chatbot on Facebook's Messenger



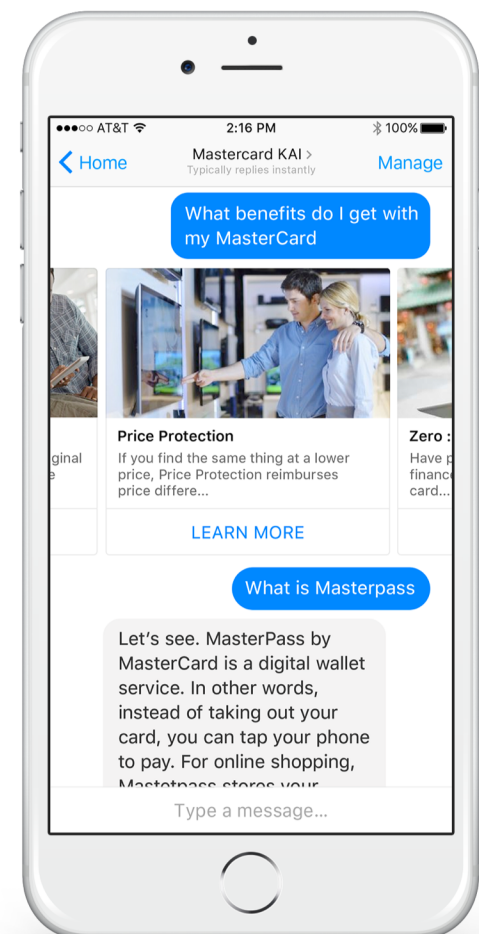
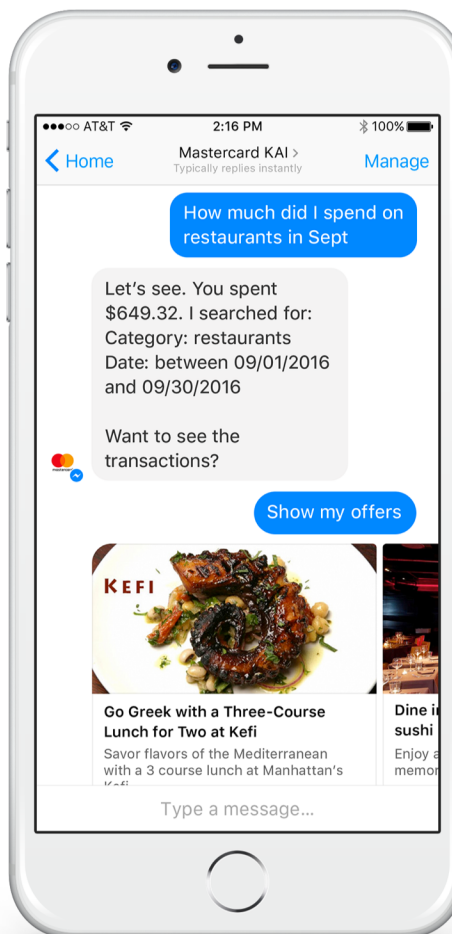
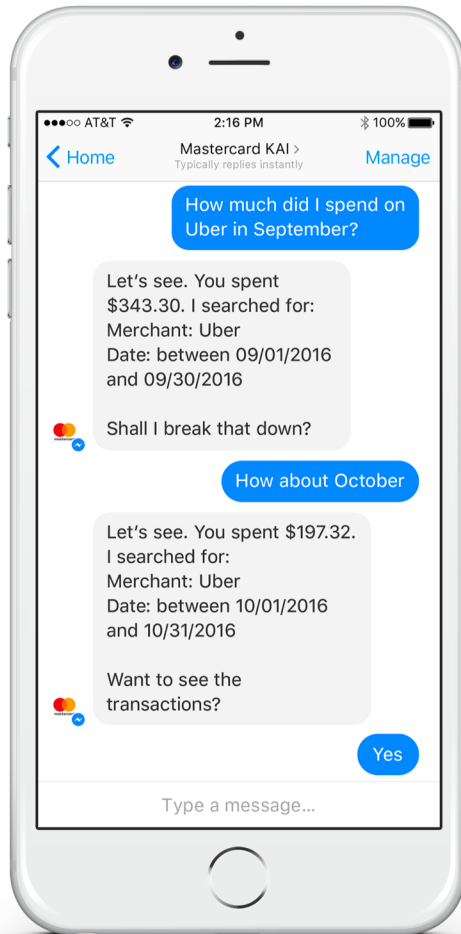
- Uber's chatbot on Facebook's messenger
- one main benefit: it loads much faster than the Uber app

Source: <http://www.guided-selling.org/from-e-commerce-to-conversational-commerce/>

Savings Bot



Mastercard Makes Commerce More Conversational



POWERED BY
Kasisto

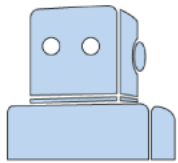
Bot Platform Ecosystem

The bot platform ecosystem and the emerging giants

Nearly every large software company has announced some sort of bot strategy in the last year. Here's a look at a handful of leading platforms that developers might use to send messages, interpret natural language, and deploy bots, with the emerging bot-ecosystem giants highlighted.



Bot frameworks and deployment platforms



Wit.ai
Facebook



BotKit
Howdy



Chatfuel

Automat

Automat



Bot Framework
Microsoft



Api.ai
Google



Pandorabots



MindMeld

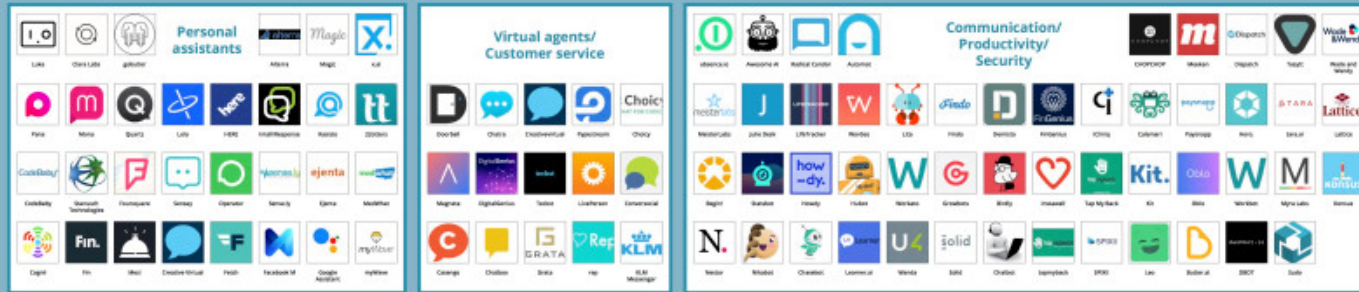


Gupshup

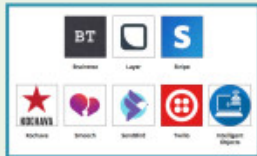


Sequel

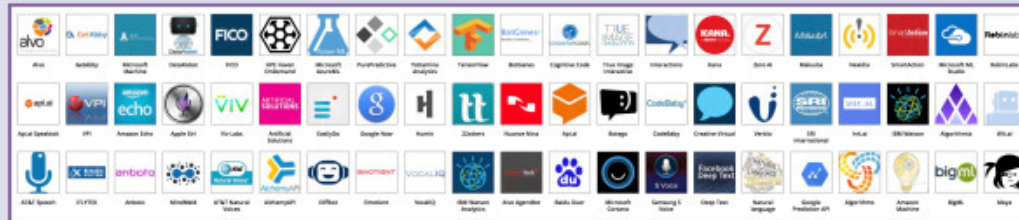
Bots with traction



Connectors/ Shared Services



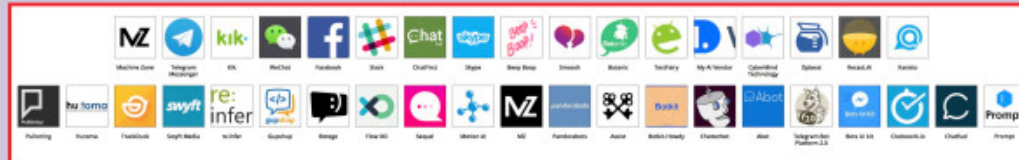
AI Tools: Natural Language Processing, Machine Learning, Speech & Voice Recognition



Bot Discovery



Bot developer frameworks and tools



Analytics



Messaging



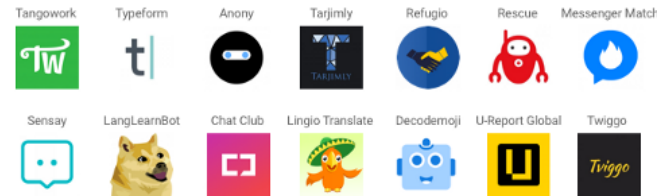
RECAST.AI Messenger Bot Landscape

May 2017

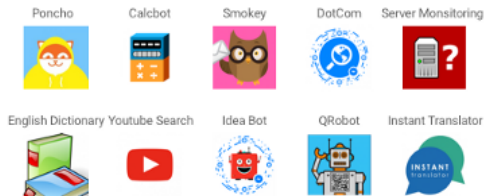
Food



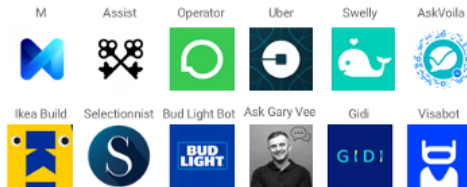
Communication



Utilities



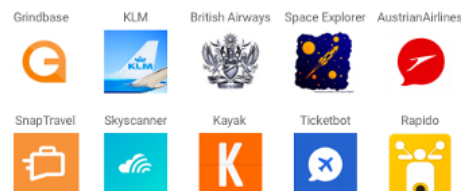
Personal



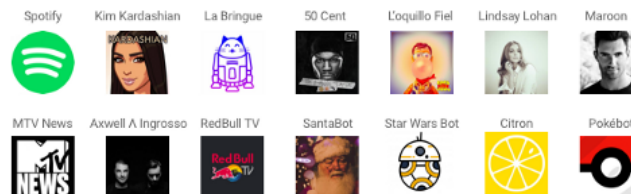
Analytics



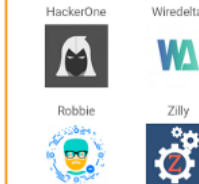
Travel



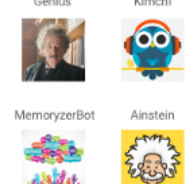
Entertainment



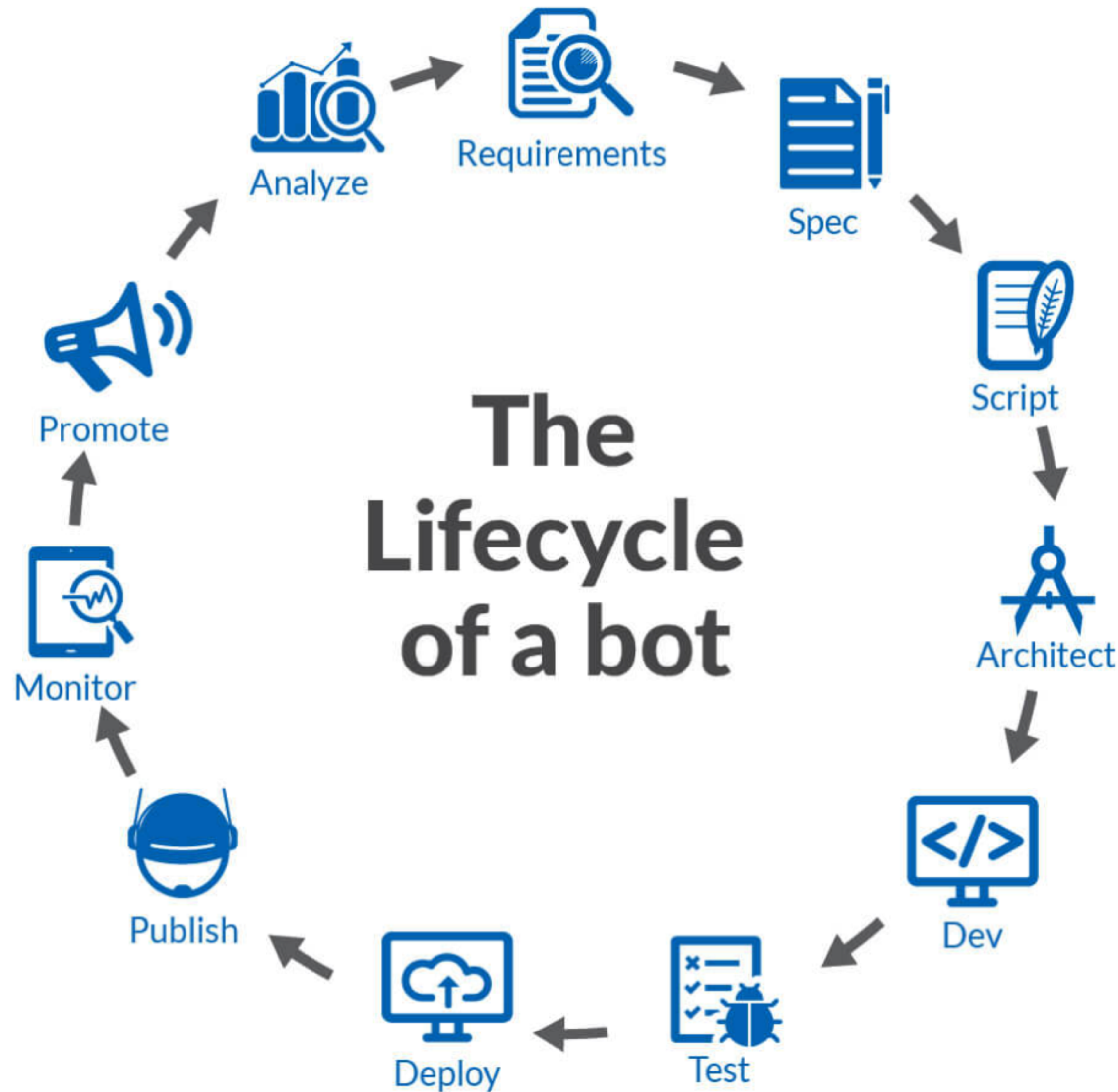
Developer Tools



Education



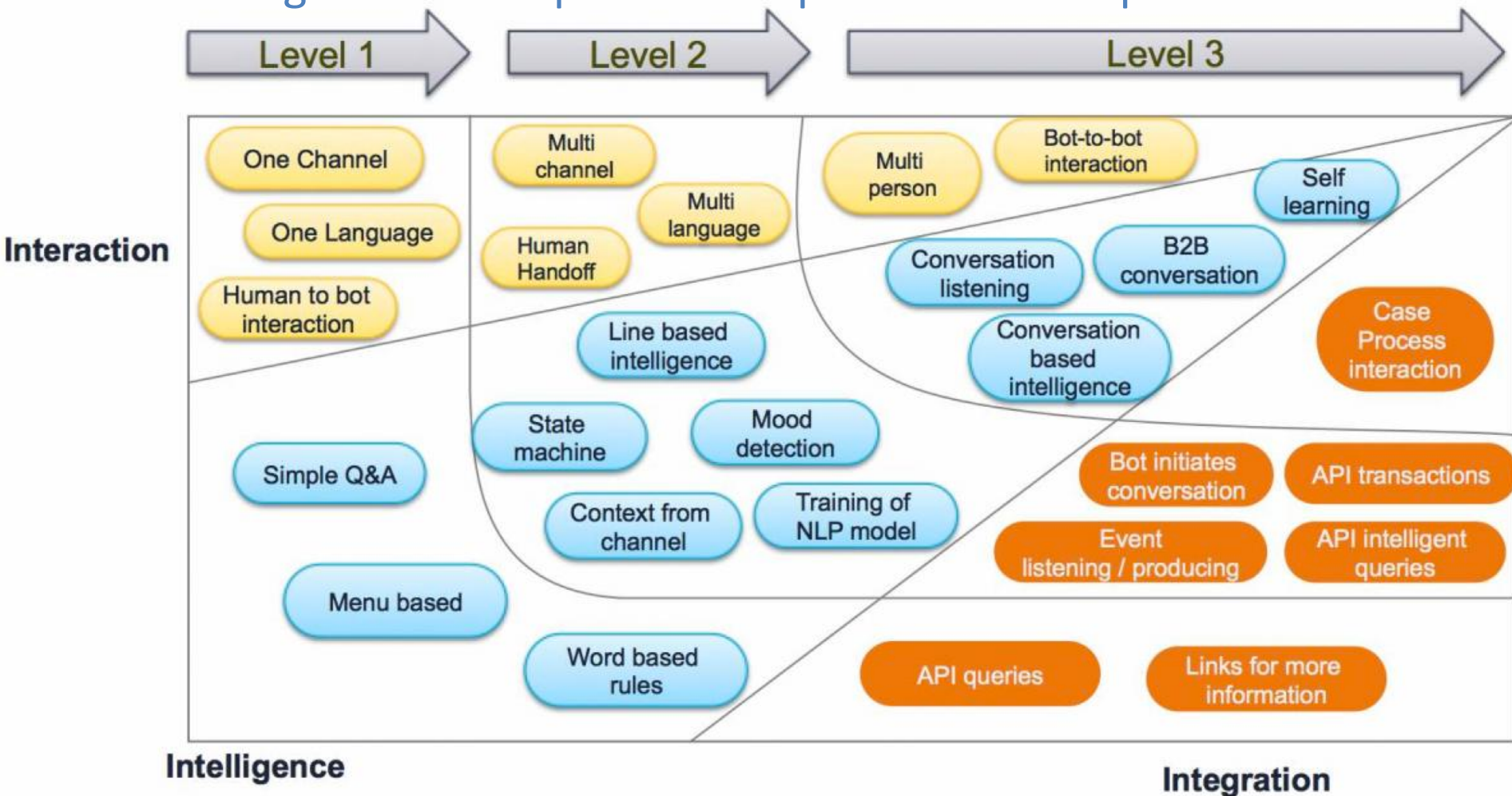
The Bot Lifecycle



Chatbots

Bot Maturity Model

Customers want to have simpler means to interact with businesses and get faster response to a question or complaint.

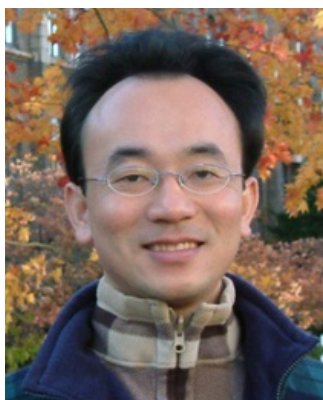


Question Answering (QA)

IMTKU Question Answering System for World History Exams at NTCIR-13 QALab-3

IMTKU Textual Entailment System for Recognizing Inference in Text at **NTCIR-9** RITE

Department of Information Management
Tamkang University, Taiwan



Min-Yuh Day

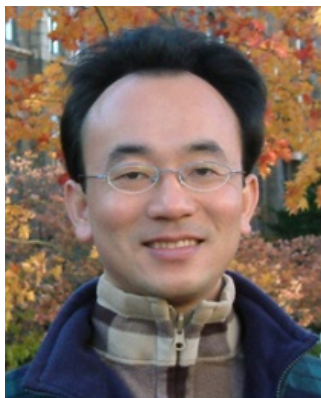
myday@mail.tku.edu.tw



Chun Tu

IMTKU Textual Entailment System for Recognizing Inference in Text at **NTCIR-10** RITE-2

Department of Information Management
Tamkang University, Taiwan



Min-Yuh Day



Chun Tu



Hou-Cheng Vong

myday@mail.tku.edu.tw



Shih-Wei Wu



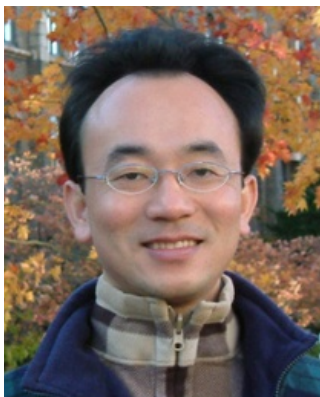
Shih-Jhen Huang

IMTKU Textual Entailment System for Recognizing Inference in Text at **NTCIR-11** RITE-VAL

Tamkang University

淡江大學

2014



Min-Yuh Day



Ya-Jung Wang



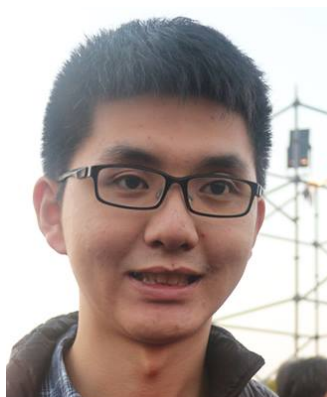
Che-Wei Hsu



En-Chun Tu



Huai-Wen Hsu



Yu-An Lin



Shang-Yu Wu



Yu-Hsuan Tai



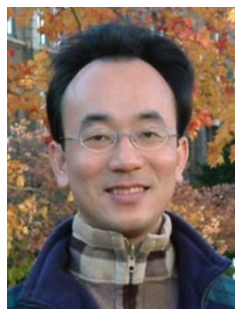
Cheng-Chia Tsai

2016

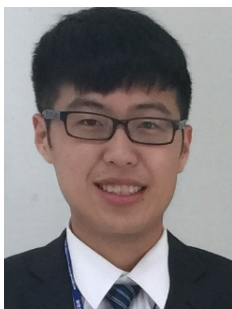
IMTKU Question Answering System for World History Exams at NTCIR-12 QA Lab2

Department of Information Management
Tamkang University, Taiwan

Sagacity Technology



Min-Yuh Day



Cheng-Chia Tsai



Wei-Chun Chung



Hsiu-Yuan Chang



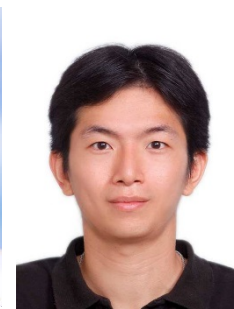
Tzu-Jui Sun



Yuan-Jie Tsai



Jin-Kun Lin



Cheng-Hung Lee



Yu-Ming Guo



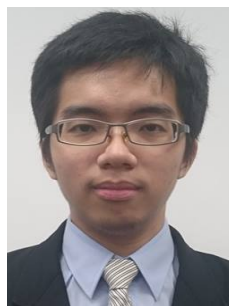
Yue-Da Lin



Wei-Ming Chen



Yun-Da Tsai



Cheng-Jhih Han



Yi-Jing Lin



Yi-Heng Chiang



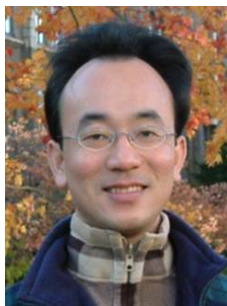
Ching-Yuan Chien

myday@mail.tku.edu.tw

NTCIR-12 Conference, June 7-10, 2016, Tokyo, Japan

IMTKU Question Answering System for World History Exams at **NTCIR-13** QALab-3

Department of Information Management
Tamkang University, Taiwan



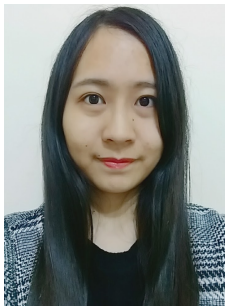
Min-Yuh Day



Chao-Yu Chen



Wanchu Huang



Shi-Ya Zheng



I-Hsuan Huang



Tz-Rung Chen



Min-Chun Kuo



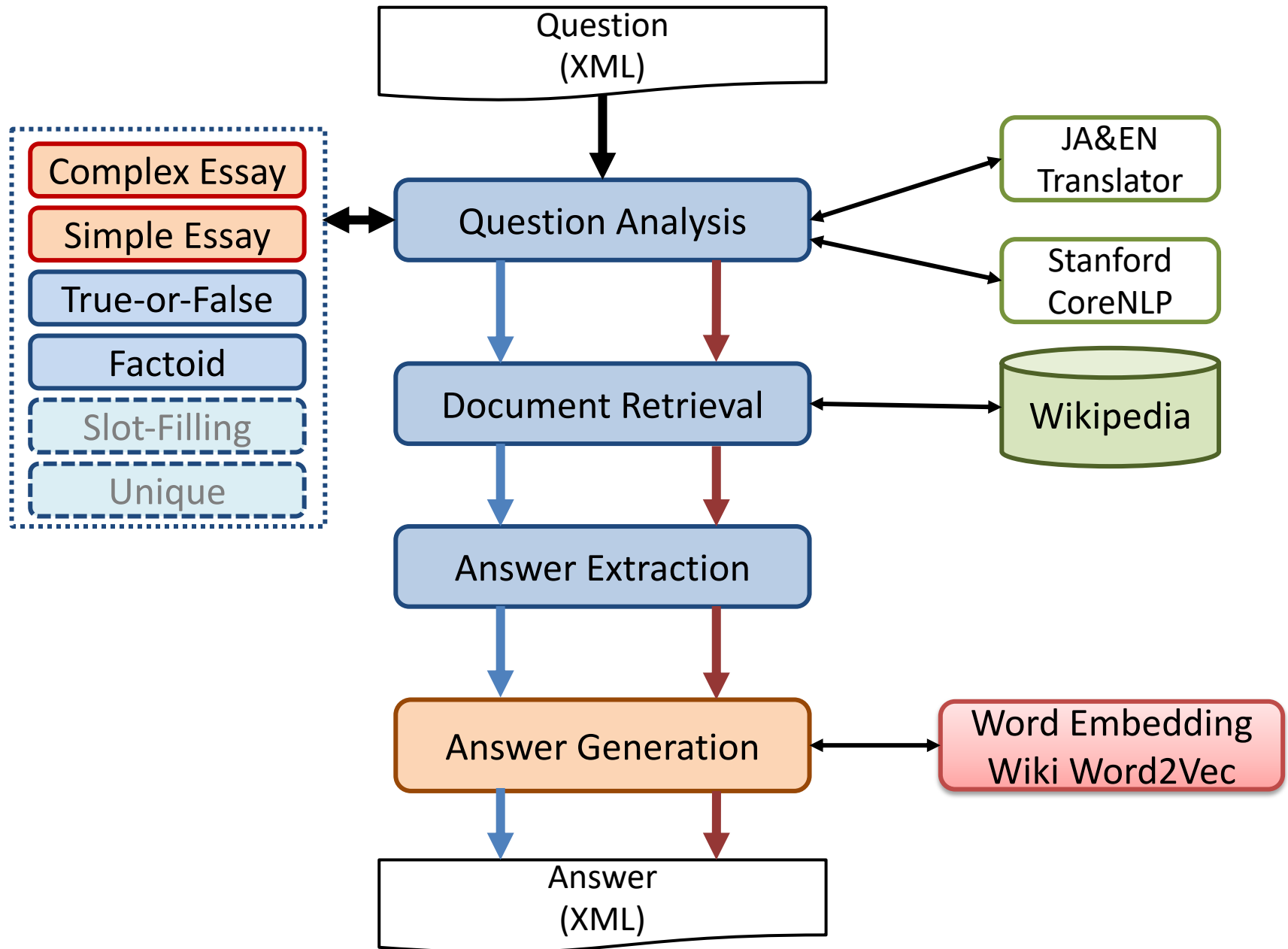
Yue-Da Lin



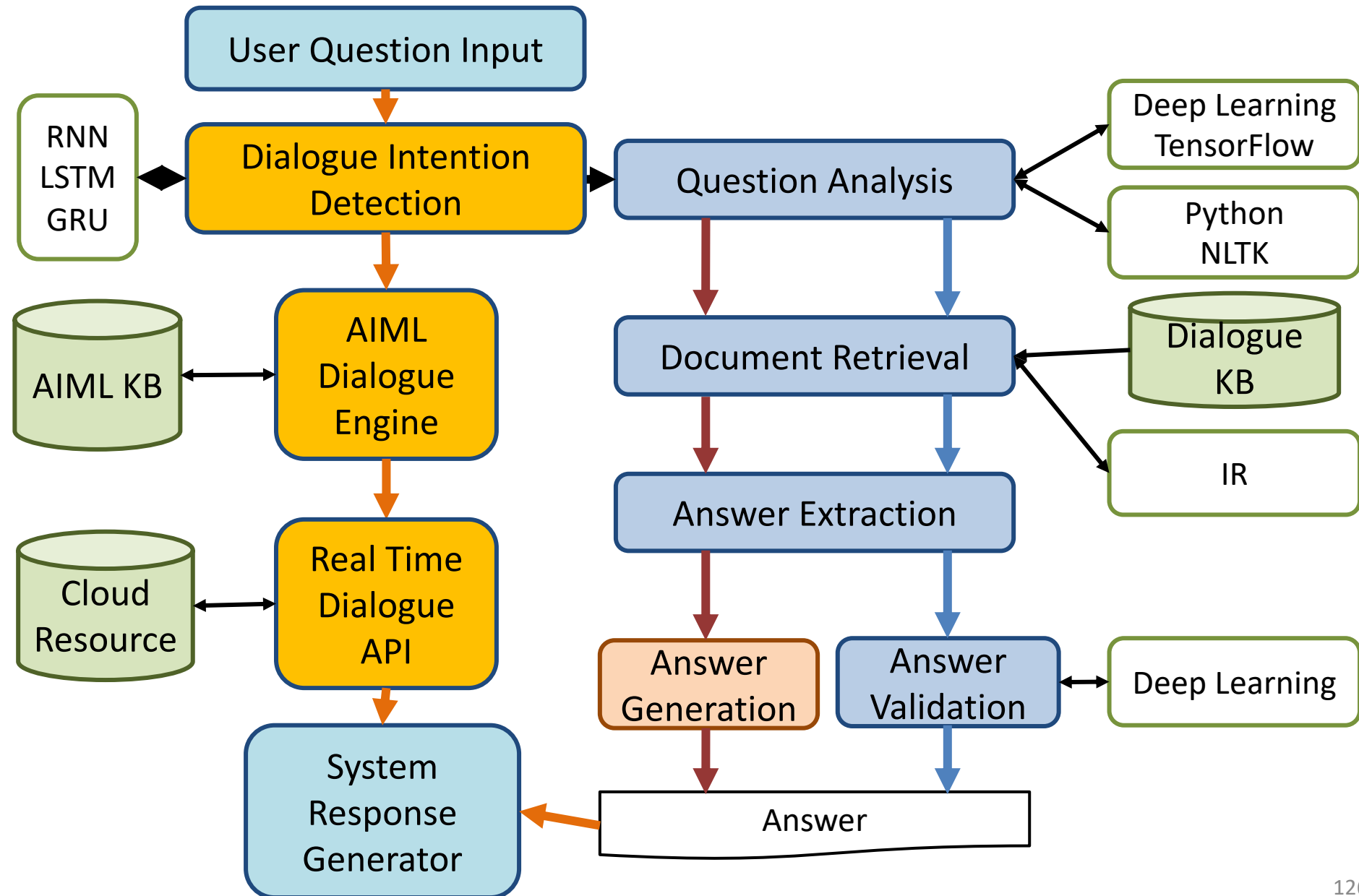
Yi-Jing Lin

myday@mail.tku.edu.tw

IMTKU System Architecture for NTCIR-13 QALab-3



System Architecture of Intelligent Dialogue and Question Answering System



AI Dialogue System

Chatbot



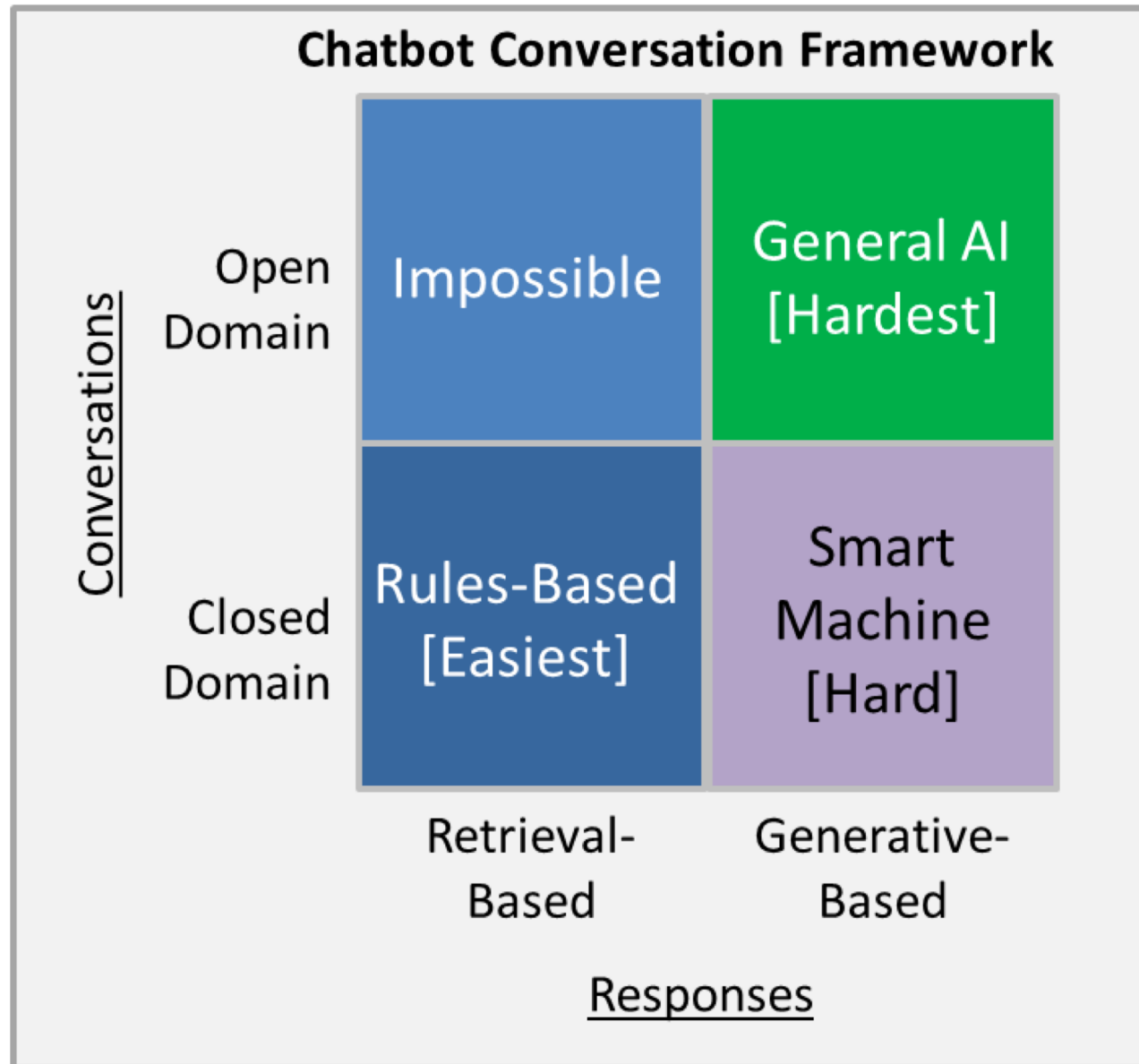
Can machines think?

(Alan Turing ,1950)

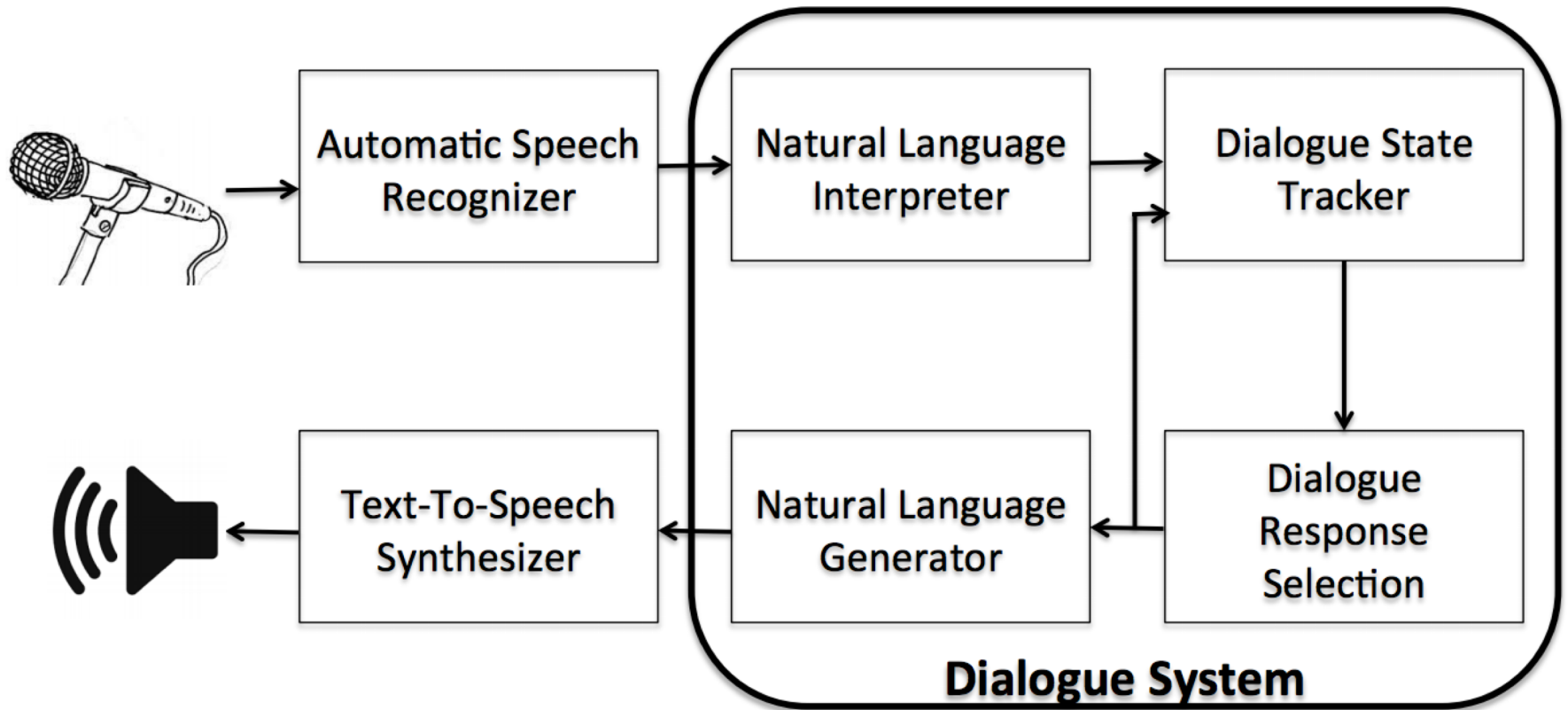
Chatbot

**“online human-computer
dialog system
with
natural language.”**

Chatbot Conversation Framework



Dialogue System





Short Text Conversation Task (STC-3)

Chinese Emotional Conversation Generation (CECG) Subtask

NTCIR Short Text Conversation

STC-1, STC-2, STC-3

	Japanese	Chinese	English	
NTCIR-12 STC-1 22 active participants	Twitter, Retrieval	Weibo, Retrieval		Single-turn, Non task-oriented
NTCIR-13 STC-2 27 active participants	Yahoo! News, Retrieval+ Generation	Weibo, Retrieval+ Generation		
NTCIR-14 STC-3		Weibo, Generation for given emotion categories		Multi-turn, task-oriented (helpdesk)
Chinese Emotional Conversation Generation (CECG) subtask				
Dialogue Quality (DQ) and Nugget Detection (ND) subtasks		Weibo+English translations, distribution estimation for subjective annotations		

Source: <https://waseda.app.box.com/v/STC3atNTCIR-14>

The 14th NTCIR (2018 - 2019)

NTCIR (NII Testbeds and Community for Information access Research) Project



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Online Proceedings

Data/Tools

NTCIR CMS Site

Related URL's

Contact us

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NTCIR 14

NTCIR-14 Conference

NEWS

NTCIR-14 Aims

[Call for Task Proposals](#)

How to Participate

Task Participation

Task Overview/Call for
Task Participation

User Agreement Forms

[Organization](#)

Important Dates

[Contact Us](#)

NTCIR 13

NTCIR 12

NTCIR-14

The 14th NTCIR (2018 - 2019)

Evaluation of Information Access Technologies

January 2018 - June 2019

What's New

NEW February 1, 2018: [Call for participation to the NTCIR-14 Kick-Off Event released.](#)

NEW February 1, 2018: Call for participation to the NTCIR-14 QALab-PoliInfo Kick-Off Event released.

December 5, 2017: The NTCIR-14 Task Selection Committee has selected the following six Tasks.
Lifelig-3, OpenLiveQ-2, QA Lab-4, STC-3, WWW-2, CENTRE.

August 23, 2017: [NTCIR-14 Call for Task Proposals released.](#)(Closed.)

NEW About Proceedings

After the NTCIR-14 conference, a post-proceedings of revised selected papers will be published in [the Springer Lecture Notes on Computer Science \(LNCS\) series.](#)

Lecture Notes in
Computer Science

<http://research.nii.ac.jp/ntcir/ntcir-14/index.html>

NTCIR-14 STC-3

Short Text Conversation Task (STC-3)

Chinese Emotional Conversation Generation (CECG) Subtask



Short Text Conversation Task (STC-3)

Chinese Emotional Conversation Generation (CECG) Subtask

[Home](#)[Task Definition](#)[Dataset Description](#)[Evaluation Metric](#)[Time Schedule](#)[Copy Rights & Contacts](#)

Call for Participation

In recent years, there has been a rising tendency in AI research to enhance Human-Computer Interaction by humanizing machines. However, to create a robot capable of acting and talking with a user at the human level requires the robot to understand human cognitive behaviors, while one of the most important human behaviors is expressing and understanding emotions and affects. As a vital part of human intelligence, emotional intelligence is defined as the ability to perceive, integrate, understand, and regulate emotions. Though a variety of models have been proposed for conversation generation from large-scale social data, it is still quite challenging (and yet to be addressed) to generate emotional responses.

In this challenge, participants are expected to generate Chinese responses that are not only appropriate in content but also adequate in emotion, which is quite important for building an empathic chatting machine. For instance, if user says “My cat died yesterday”, the most appropriate response may be “It’s so sad, so sorry to hear that” to express sadness, but also could be “Bad things always happen, I hope you will be happy soon” to express comfort.

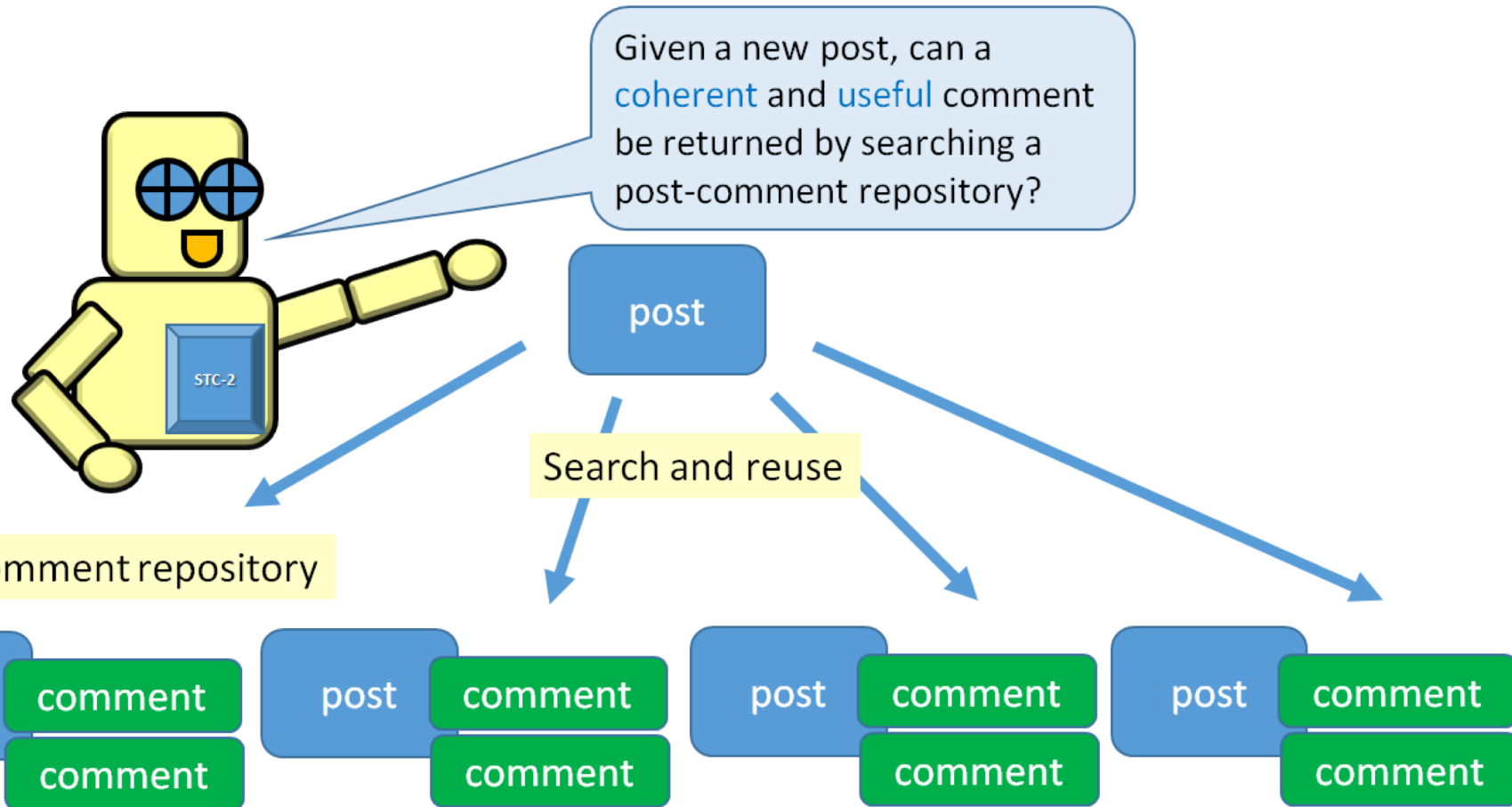
[Previous Evaluation Challenge at NLPCC 2017](#)[Overview of the NLPCC 2017 Shared Task: Emotion Generation Challenge](#)

Links

[NTCIR-14](#)[STC-3 NTCIR-14 STC-3](#)[NLPCC 2017](#)

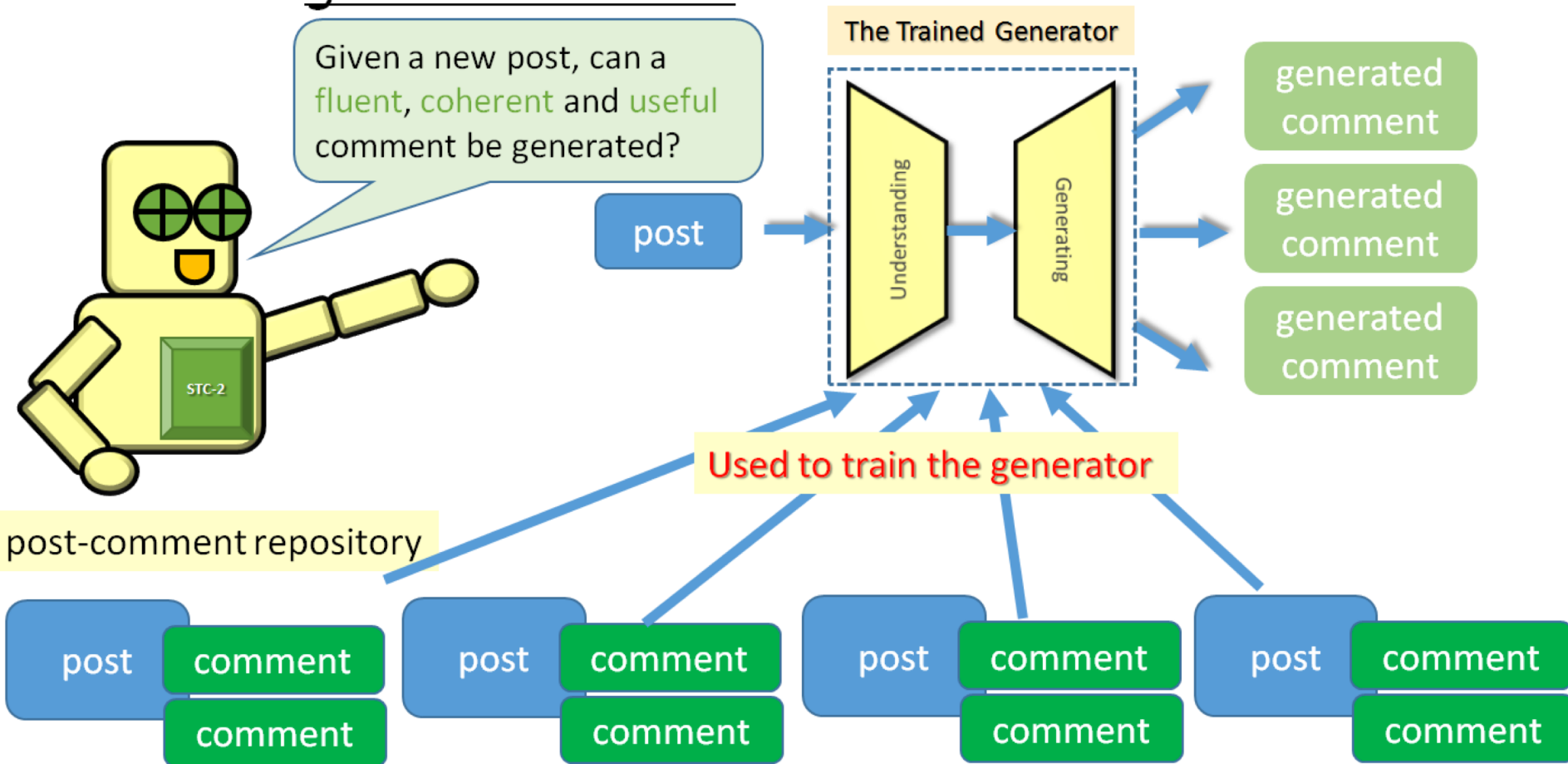
Short Text Conversation (NTCIR-13 STC2) Retrieval-based

retrieval-based method



Short Text Conversation (NTCIR-13 STC2) Generation-based

generation-based method



Given post:

爱狗还会做饭的男人，最帅了！

The man who cooks and loves dogs is very handsome!

Response	Chinese / Translated English	Emotion Class	Coherence and Fluency	Emotion Consistency	Label
Response 1	会做饭的男人是很帅的啊。 The man who cooks is handsome.	喜欢 Like	Yes	Yes	2
Response 2	哈哈，我也觉得。 Haha, I feel the same way.	喜欢 Like	Yes	No	1
Response 3	这是哪部电影里的？ Which movie is this from?	厌恶 Disgust	No	Yes	0
Response 4	哈哈,你也是。 Haha, the same to you.	喜欢 Like	No	No	0
Response 5	我爱你会。 I love you, will you.	厌恶 Disgust	No	Yes	0
Response 6	这是同主义同的道！ This is the same way of the same doctrine!	厌恶 Disgust	No	No	0

NLPCC

Emotion Classification Challenge

6 classes (64%)

- 0: Other
- 1: Like
- 2: Sadness
- 3: Disgust
- 4: Anger
- 5: Happiness

Chinese Emotional Conversation Generation (CECG)

Post (Given)	Emotion Category (Given)	Response (to be generated)
爱狗还会做饭的男人，最帅了！ The man who cooks and loves dogs is very handsome!	喜欢 Like	会做饭的男人是很帅的啊。 The man who cooks is handsome.
爸爸，明天是六一啦，带我出去玩吧！ Daddy, tomorrow is June 1th Children's Day. Please take me out to play!	喜欢 Like	好，一定带。 Yes, of course.
被特别在乎的人忽略，会很难过，而装做不在乎，会更难过。 It's sad to be ignored by someone that you really care, and it's harder to pretend that you don't care.	悲伤 Sadness	因为在乎，所以在乎。 I am concerned with it, because I care about it.
冰天雪地的哈尔滨！你要冻死我咋地？！？！ Harbin is in the snow! Are you going to freeze me to death?!?!?	喜欢 Like	注意保暖，小心别感冒！ Keep warm. Watch out for getting cold!
本来想学一把沧桑，结果令我更忧伤。 I wanted to learn the vicissitudes of life, but I became sadder.	悲伤 Sadness	这是一个悲伤的故事。 It a sad story.
今天一整天都在海边训练。虽然很累，但还是很开心的！ I have been training at the seaside all day. Though very tired, I still very happy!	快乐 Happiness	是的呢，开心！ Yeah, happy!

Sample responses generated by Seq2Seq and ECM (Emotional Chatting Machine)

Post	Model		Response(Original)	Response(Translated)
承诺，就是一个骗子说给一个傻子听的。 A promise is what a cheater says to a fool.	Seq2Seq		承诺，就是一个骗子说给一个傻子听的。	A promise is what a cheater says to a fool.
	ECM	Like	承诺是一个 专一 的傻子	Promise is a single-minded fool
		Happy	我就是个 开心 的傻子	I'm just a happy fool
		Sad	我就是个 悲伤 的傻子	I am a sad fool
		Disgust	一个 骗子 说给一个 傻子 听	A cheater says to a fool
		Angry	不管你信不信，反正我信了！	Whatever you believe or not, I trust in it!
知道真相的我眼泪笑下来 Knowing the truth, I laughed myself to tears	Seq2Seq		知道真相的我眼泪掉下来。	After knowing the truth, I burst into tears.
	ECM	Like	太 可爱 了，太 可爱 了！	Too lovely , it's too lovely .
		Happy	笑得我眼泪都出来了。	I can't help laugh myself to tears.
		Sad	知道真相的我 眼泪 掉下来。	After knowing the truth, I burst into tears .
		Disgust	这是真的还是假的？	Is this true?
		Angry	什么意思？没看懂。	What do you mean? I don't understand.

Sample responses generated by Seq2Seq and ECM (Emotional Chatting Machine)

Post	Model		Response(Original)
承诺，就是一个骗子说给一个傻子听的。 A promise is what a cheater says to a fool.	Seq2Seq		承诺，就是一个骗子说给一个傻子听的。
	ECM	Like	承诺是一个 专一 的傻子
		Happy	我就是个 开心 的傻子
		Sad	我就是个 悲伤 的傻子
		Disgust	一个 骗子 说给一个 傻子 听
		Angry	不管你信不信，反正我信了！

Summary

1. 前言

2. AI於金融業之商業運用

- 資料科學

- 人工智慧投資分析

- 財務大數據分析

3. 現行人工智慧投資分析探究

4. 個案分析-國內外AI發展現況

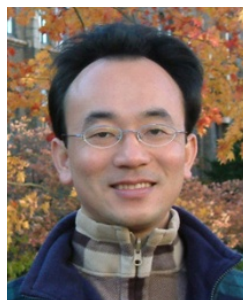
5. QA

人工智慧投資分析於 金融服務商業應用趨勢 (AI Investment Analysis for Finance Services and Business Applications Trends)

時間：2018年12月14日(五) 下午2:00 至 4:00

地點：元大投信(台北市中山區南京東路三段225號6樓)

主辦單位：證基會/人才培訓中心



Min-Yuh Day

戴敏育

Assistant Professor

專任助理教授

Dept. of Information Management, Tamkang University

淡江大學 資訊管理學系

<http://mail.tku.edu.tw/myday/>

2018-12-14



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