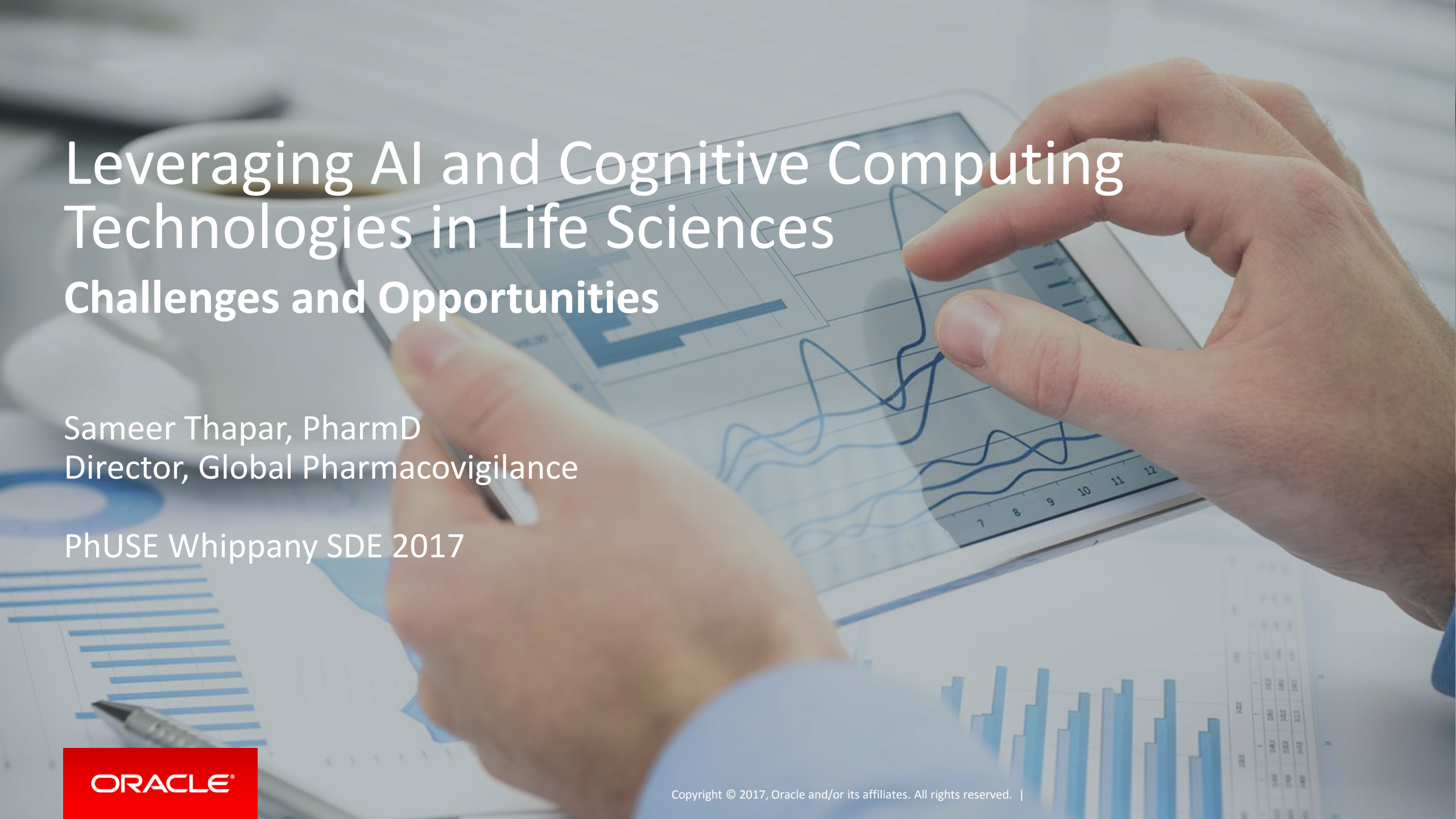


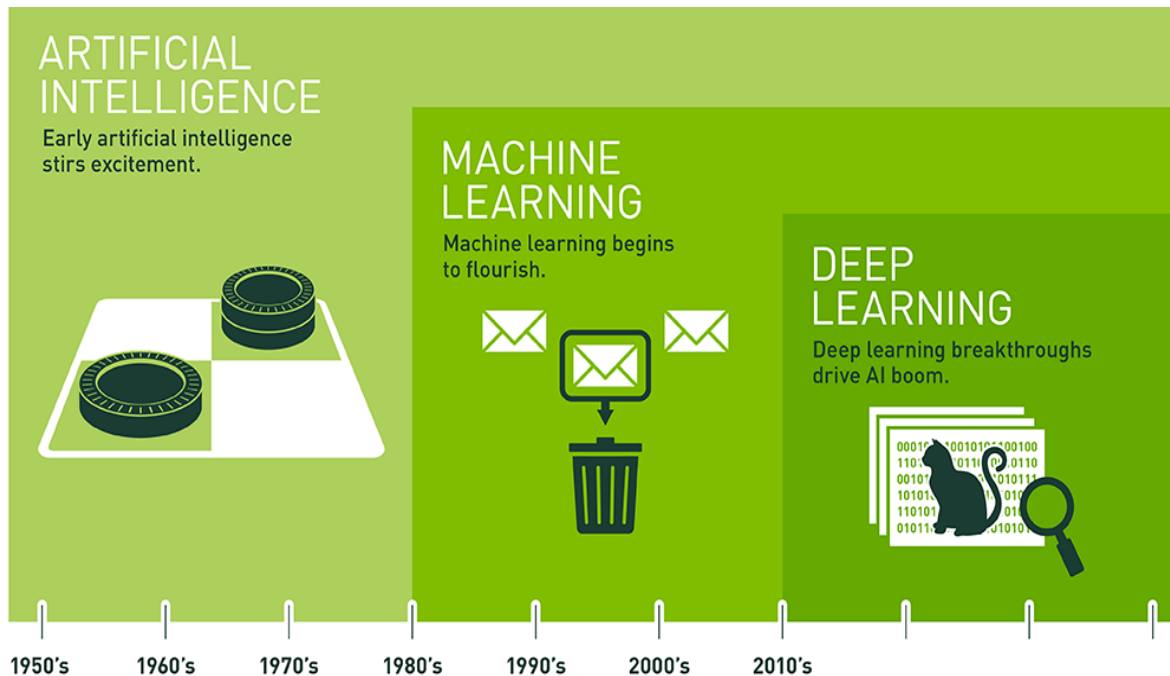


Leveraging AI and Cognitive Computing Technologies in Life Sciences Challenges and Opportunities

Sameer Thapar, PharmD
Director, Global Pharmacovigilance

PhUSE Whippany SDE 2017

Basics: AI versus ML versus DL



- **Artificial Intelligence:** computer systems able to perform tasks that normally require human intelligence, such as visual perception, speech recognition, decision-making, and translation between languages.
- **Machine learning:** subset of AI having the ability to automatically learn and improve from experience without being explicitly programmed.
- **Deep Learning:** subfield of machine learning concerned with algorithms called artificial neural networks.

Overview Only

This presentation will not go into detail on the extensive AI/CC/ML platforms and toolkits

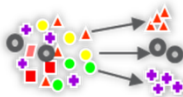
CLASSIFICATION

- Naïve Bayes
- Logistic Regression (GLM)
- Decision Tree
- Random Forest
- Neural Network
- Support Vector Machine
- Gaussian Mixture Models



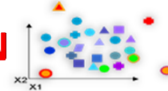
CLUSTERING

- Hierarchical K-Means
- Hierarchical O-Cluster
- Expectation Maximization (EM)



ANOMALY DETECTION

- One-Class SVM



TIME SERIES

- Single, Double Exp Smoothing

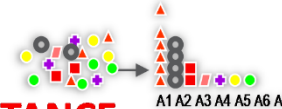
REGRESSION

- Linear Model
- Generalized Linear Model
- Support Vector Machine (SVM)
- Stepwise Linear regression
- Neural Network
- LASSO



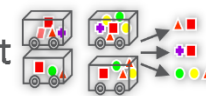
ATTRIBUTE IMPORTANCE

- Minimum Description Length
- Principal Comp Analysis (PCA)
- Unsupervised Pair-wise KL Div



ASSOCIATION RULES

- A priori/ market basket



PREDICTIVE QUERIES

- Predict, cluster, detect, features

SQL ANALYTICS

- SQL Windows, SQL Patterns, SQL Aggregates



FEATURE EXTRACTION

- Principal Comp Analysis (PCA)
- Non-negative Matrix Factorization
- Singular Value Decomposition (SVD)
- Explicit Semantic Analysis (ESA)

TEXT MINING SUPPORT

- Algorithms support text type
- Tokenization and theme extraction
- Explicit Semantic Analysis (ESA) for document similarity



STATISTICAL FUNCTIONS

- Basic statistics: min, max, median, stdev, t-test, F-test, Pearson's, Chi-Sq, ANOVA, etc.



R PACKAGES

- CRAN R Algorithm Packages through Embedded R Execution
- Spark MLlib algorithm integration



Case Study: Automotive Industry Social Text Mining

Using Social Listening Algorithm to Keep Customer Centric Focus

Automotive Industry

- A Large Automotive Company wanted to know when its customers talk about car problems
- Attempt to gauge context and sentiment
- Social Data Listening Platform helps companies monitor social media conversations around their brands
- A simple, supervised, document classification example

"Yooo the bakes on the Camaro stopped working, we ran like 4 stop signs & almost died lmaooo"

"Well the Cadillac wont start"

"I realized that money doesn't buy happiness but I'd rather do my crying in a Corvette."

Automotive Case Applied to Life Sciences

Challenges

- Regulations on company branded communications to public
- Potential of adverse event or inadvertent medical communication and handling
- Once you know of disinformation, what are next steps?
- How to overcome the nuances of specialized lexicons, medical dictionaries, and coding?

Opportunities

- Ability to gauge and prevent negative sentiments from proliferating
- Trend analysis insight
- Identification of brand advocates

Case Study: Facebook & Google Image Processing

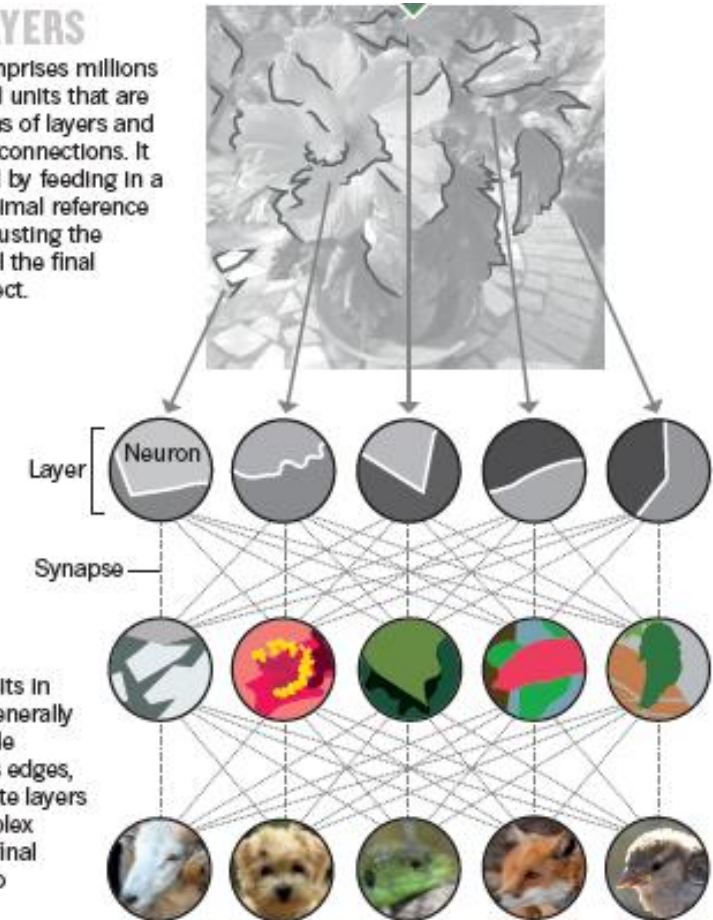
Using Deep Learning toolkits to decipher and classify images

Facebook & Google Image Classification

- Facebook Research accomplishes an internet-size image classification in an hour with 90% accuracy
- Six years ago, same feat involved days of effort
- Google's open source Deep Dream software is designed to detect faces and other patterns in images, with the aim of automatically classifying images

HIDDEN LAYERS

The network comprises millions of computational units that are stacked in dozens of layers and linked by digital connections. It has been trained by feeding in a vast library of animal reference images, then adjusting the connections until the final response is correct.



After training, units in the first layers generally respond to simple features, such as edges, while intermediate layers respond to complex shapes and the final layers respond to complete faces.

NATURE | VOL 538 | 6 OCTOBER 2016

Facebook & Google Case Applied to Life Sciences

Challenges

- Current proof of concept involves understanding the “black box” of how the AI classifies
- Transparency in current deep learning algorithms is improving but not at confidence levels for science
- Trust in machine derived analysis over human inferred ones

Opportunities

- When used as an adjunct to data discovery processes, it can narrow the choices
 - Breast cancer imaging
 - Stroke radiographs
- Can be reverse engineered by feeding it a set of known artifacts and asking it to “find” similar sets

Case Study: Candidate Search and Talent Monitoring

Using AI for predictive analysis and employee hiring/retention

Candidate Search & Talent Monitoring

- AI service lets companies analyze workers' email to tell if they're feeling unhappy about their job
- AI will collect and crunch data from a variety of sources to find people with the right talents, with experience to match
- Use AI to find employees who may be likely to jump ship based on variables

Will They Stay or Will They Go?

Some of the many types of information that go into an analysis by Workday software of how likely an employee is to leave a company

TIME since last promotion

BONUSES as compared to last year

TIME since last raise

EMPLOYEE performance and potential

MANAGER performance and potential

ATTRITION under employee's manager

TIME off taken

STOCK grants over time

TEAM size

LOCATION of the employee

LOCATION of the employee's team

LOCATION of the employee's manager

Source: Workday Inc.

THE WALL STREET JOURNAL.

HR Case Applied to Life Sciences

Challenges

- Privacy is an obvious concern when tracking employees just as HIPAA is when screening patients for trials
- Systems that sort people based on attributes also raise questions on the rigors of the parameters used
- Perceived bias in AI versus Human deduction

Opportunities

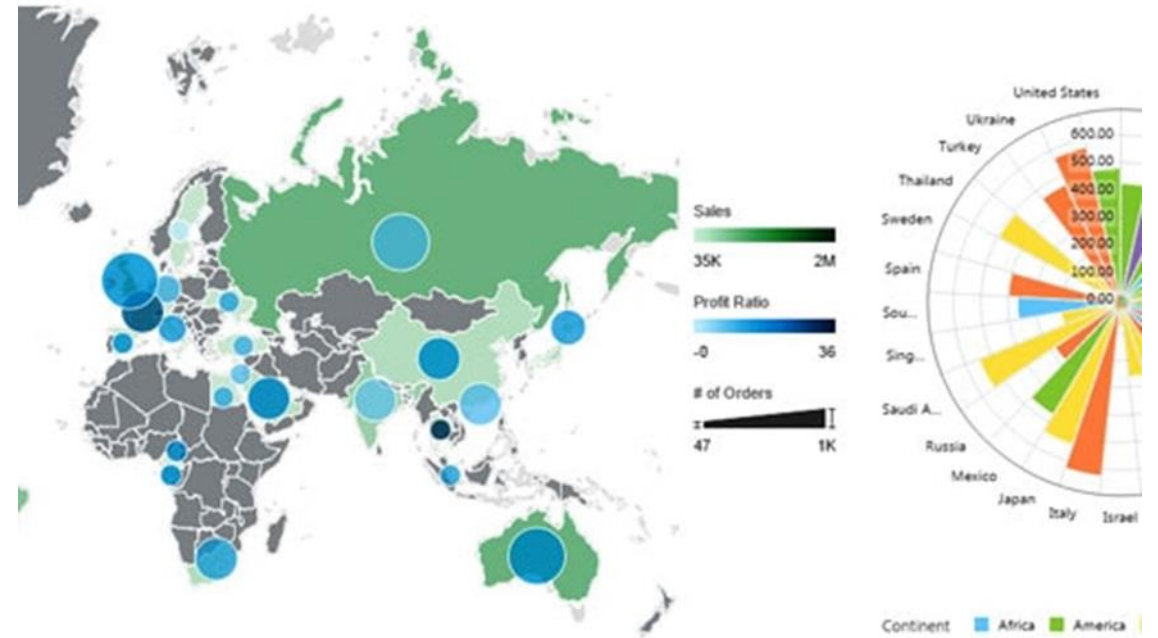
- AI systems as advisors by suggesting a set of options to human users, who ultimately make the final decisions (ie, PV)
- Can help organizations with decisions related to funding for clinical trials by offering new insights into the primary investigator, scientific topic and operational feasibility
- AI continually evolves based on new information, results and actions

Case Study: Financial Voice and Data Analysis

Using NLP to analyze structured and unstructured data

Financial Voice and Data Analysis

- Large global bank wanted its phone calls and voice records screened and analyzed into structured datasets
- Included multiple languages
- ML collapsed the “*raw data actionable insights/predictions*” process from 4-6+ weeks to mins/hours



Financial Case Applied to Life Sciences

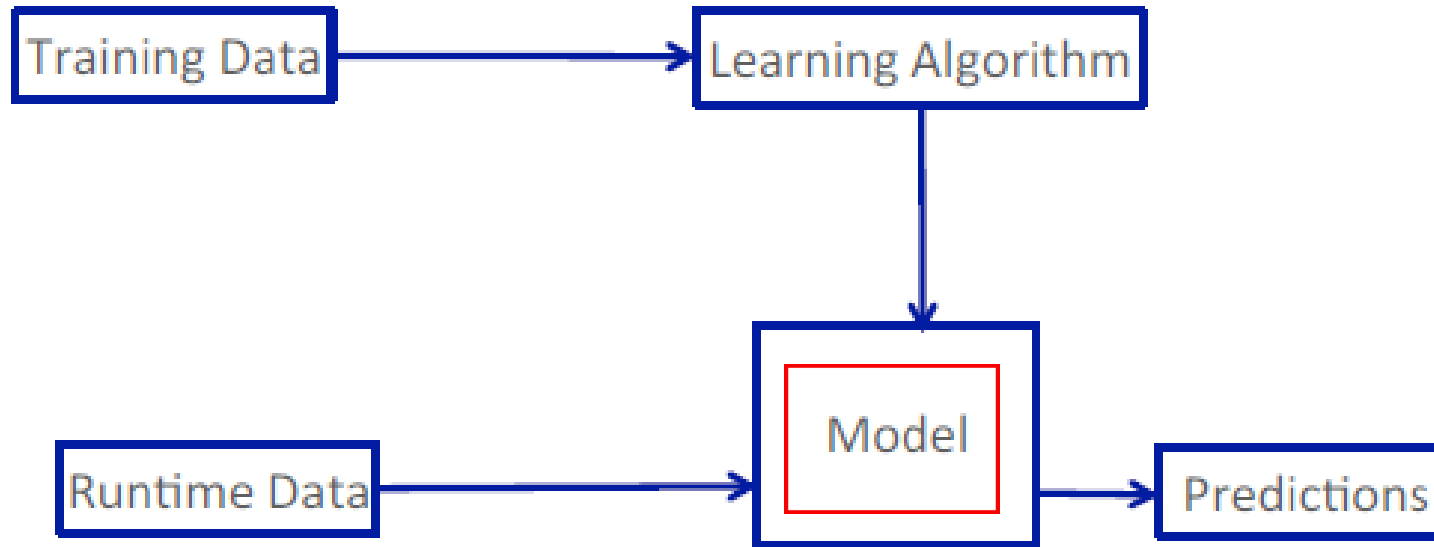
Challenges

- Multiple domains in Life Science industry, so data is in silos and bottlenecked
- Tapping into unstructured and structured datasets ideally requires a single source of truth to exist company-wide
- “Psychological embrace” – doubt that any AI/ML/DL can perform to the same accuracy and specificity as human who has learned the role via experience

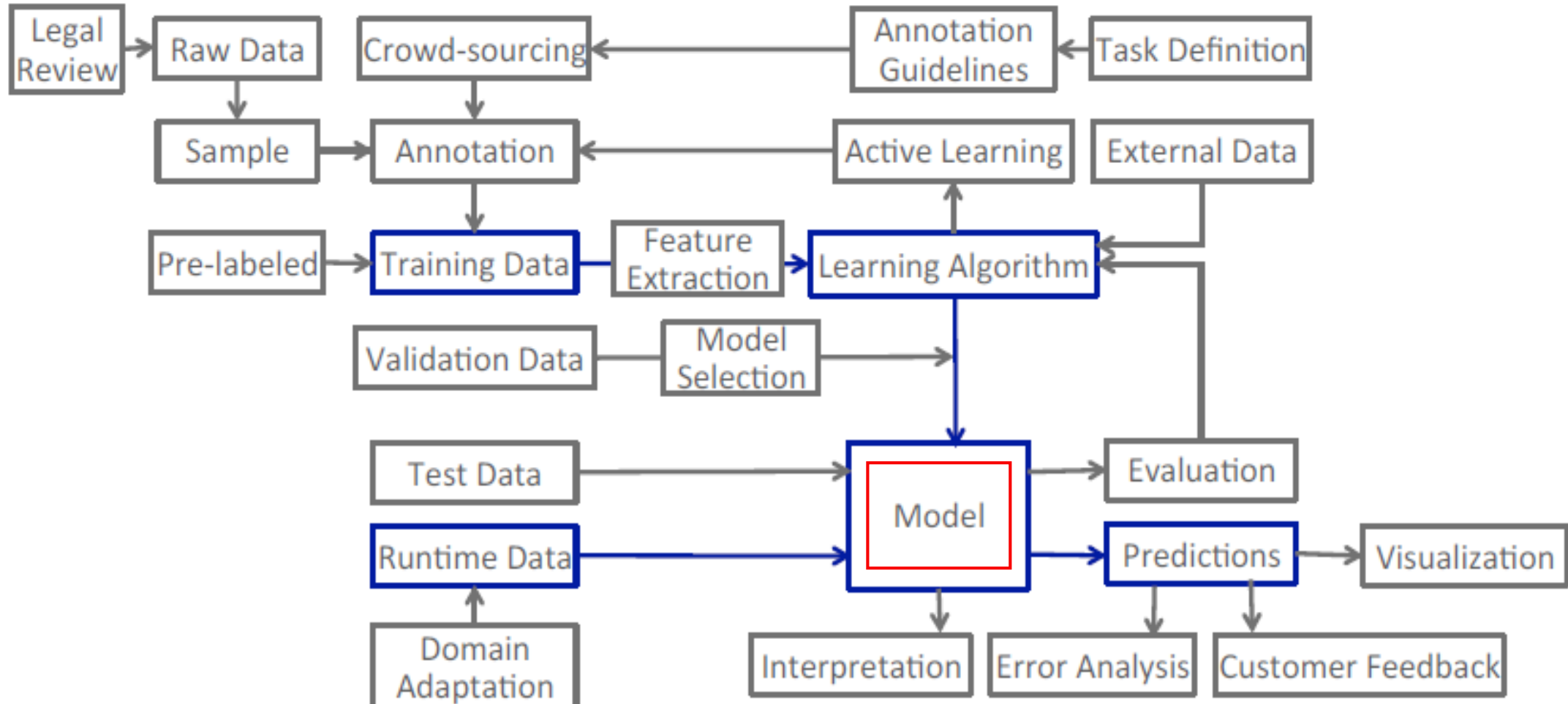
Opportunities

- Silo systems foster redundancies and breaking down these allows for AI to function optimally
- Analysis of voice data has scientific basis for disease discovery (i.e., Parkinson disease)
- Some human interaction will always remain

Machine Learning: Perception

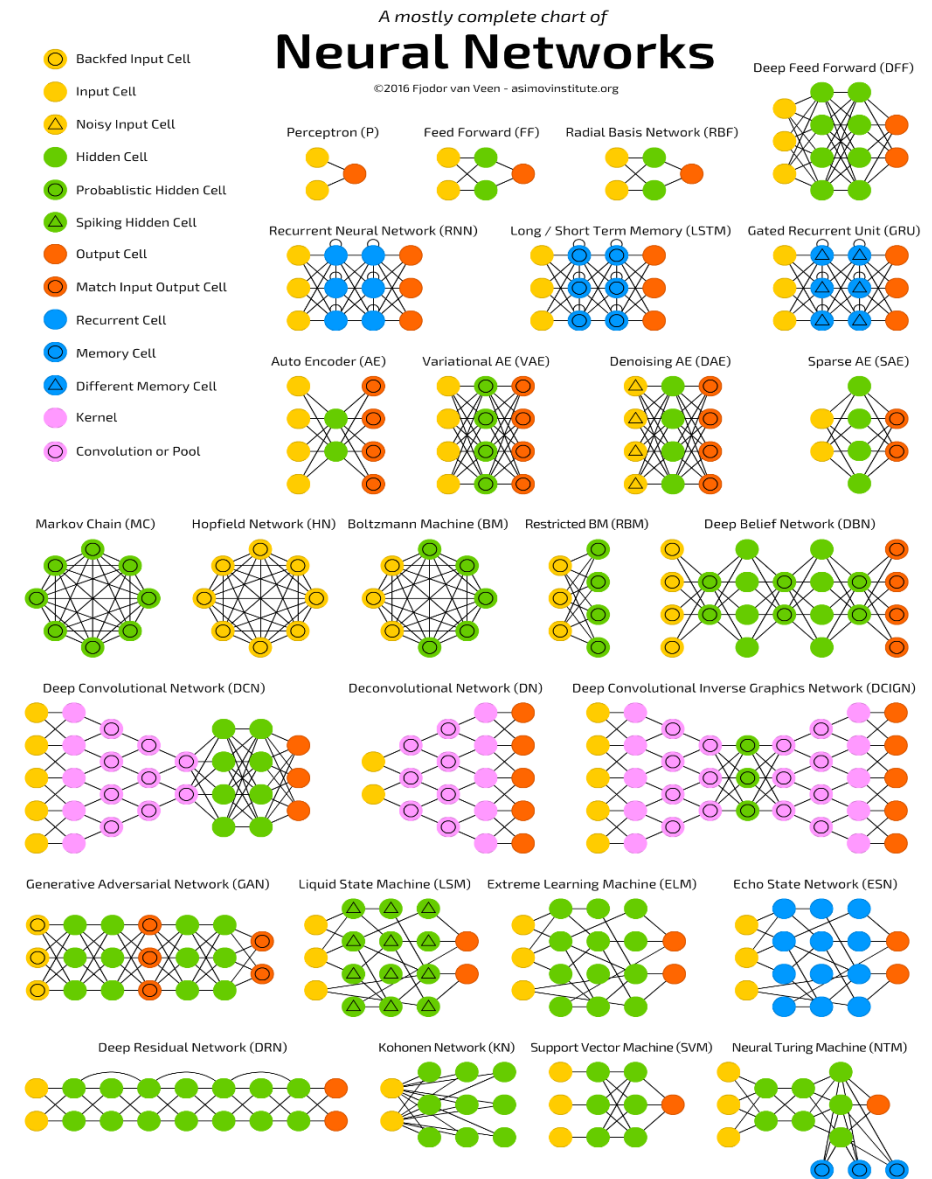


Machine Learning: Actuality



When Not To Use

- Low-budget or low-commitment problems
- Interpreting and communicating model parameters/feature importance to a general audience
- Establishing causal mechanisms
- Learning from “unstructured” features



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Q&A

Sameer Thapar, PharmD
Director, Global Pharmacovigilance
Oracle Health Sciences Consulting

[LinkedIn.com/in/Thapar](https://www.linkedin.com/in/Thapar)



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