

Introduction to Machine Learning

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June 11, 2016

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Goal

*Breakdown the perception that **machine learning** is **complex**.*

***Address assumptions** of what machine learning **IS**, and **what it's NOT**.*

***Demonstrate the basics** of machine learning.*



Machine Learning can be
perceived as **Science Fiction**

Machine Learning can be
perceived as **Complex and Vast**
in terms of *research, terminology,*
and technology

FIELD ENHANCEMENT: ON MAGNIFY: VAR WAVELENGTH: 2

01. T-1000 STATUS
02. RATIO OF ATTACK
03. 83%
04. A1 = S4889
05. A2 = 00933
06. A3 = F8367
07. A4 = G0894
08. DISTANCE 4FT
09. A5 = H9837
10. A6 = J0948
11. A7 = K8364
12. A8 = L3748
13. A9 = Z3864

POSSIBILITY OF
T-1000 TERMINATION:
52%

14. 9846592834
15. 2094875204
16. 8764523456
17. VISUAL IN SITE
18. 8735358724
19. 2345987087
20. 5983745809
21. 0987435234
22. 0876863456
23. 87560983475689347
24. 87676345665421234
25. 74657483230856723
26. 74652983745692367
27. 76747603784785747
28. 46584395344857984
29. 45786384760847508
30. 57630984763098959
31. 47560238745089763
32. 47577458868388568
33. 45437594756289775
34. 74295884858943534

TARGET AQUIRED

21 BULLETS LEFT IN CLIP

**VISUAL:
TERMINATOR
MODEL 1000**

**CAUTION: T-1000
CAPABLE OF KNIVES
AND STABBING WEAPONS
EQUIPED WITH HANDGUN**

**VULNERABLE TO
MOLTEN STEEL
AND LIQUID NITROGEN**

**PRIMARY MISSION: ENSURE THE
SURVIVAL OF JOHN CONNOR**



Latitude: 51.5001402°
Longitude: -0.126193°

Visual Analysis

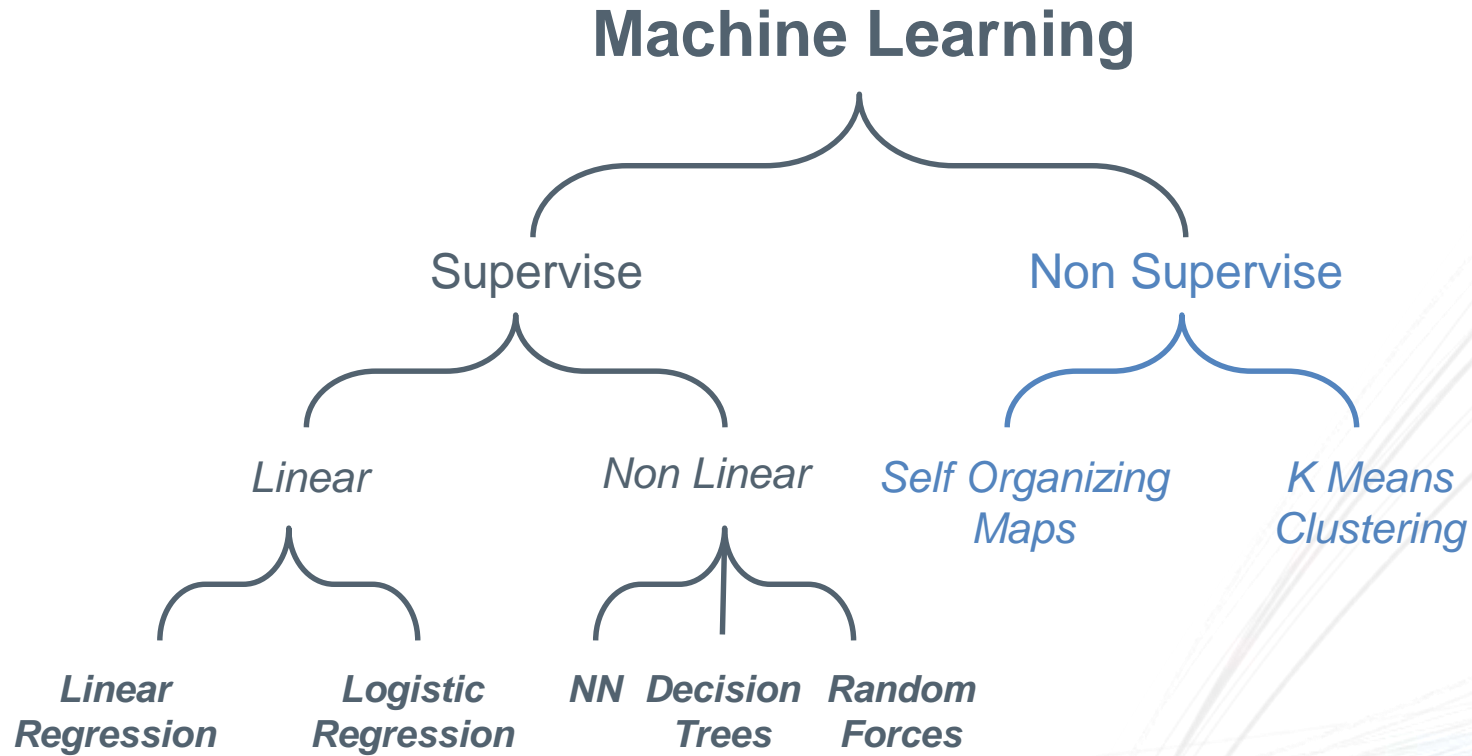
Carbon based
Humanoid...
Geek

Threat Level

TARGET ACQUIRED

What we'll accomplish today

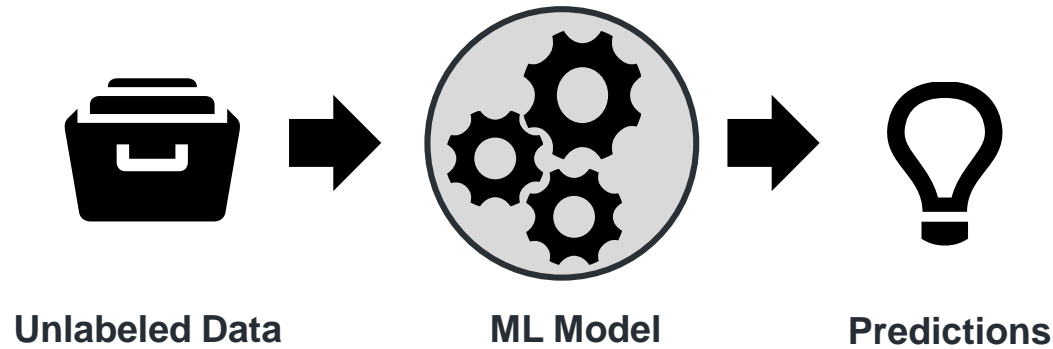
- Machine Learning Basics in 6 Slides
- How to get started
- Let's teach machines (Demo)
- Machine Learning in Healthcare
- Q & A
- References



Machine Learning in 8 Slides | Key Terms

Term	Description
Machine Learning	Algorithms that automatically improve with data
Model	Statistical representation of the algorithm's experience
Supervised Learning	Algorithms that improves using labeled examples
Linear Algorithm	Predictions are proportional to the feature/input values
Feature	Data point that affect the <i>target</i> you are trying to predict.
Target Variable	Data point that you are trying to predict
Classification	Predicting a binary or categorical target variable
Regression	Predicting a continuous target variable

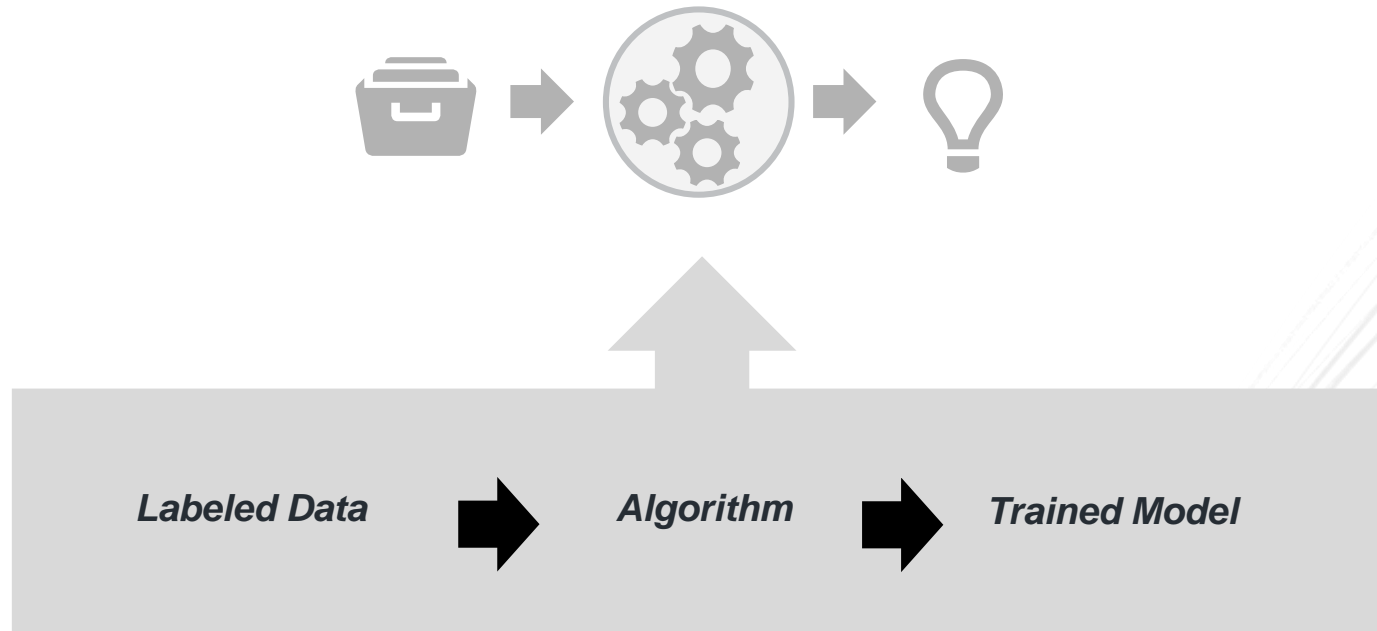
Machine Learning in 8 Slides | Supervised Models



- ML algorithms process “unseen” data to give predictions
- A model is a statistical representation of the algorithms experiences/training

How do you get a model?

Machine Learning in 8 Slides | Supervised Models

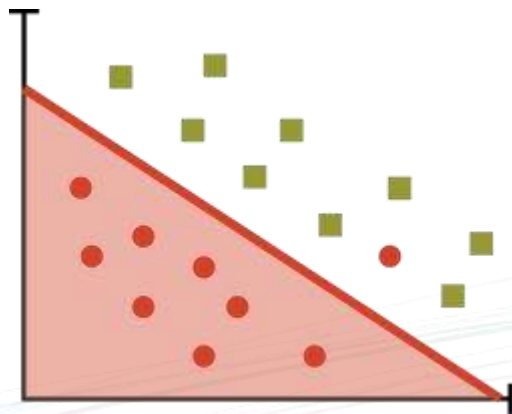


- Supervised ML algorithms get their name because they learn with help
- You have to provide them experience in labeled examples
- The algorithm translates that data into a representation called a model
- It can be used later to make predictions
- The more data the better

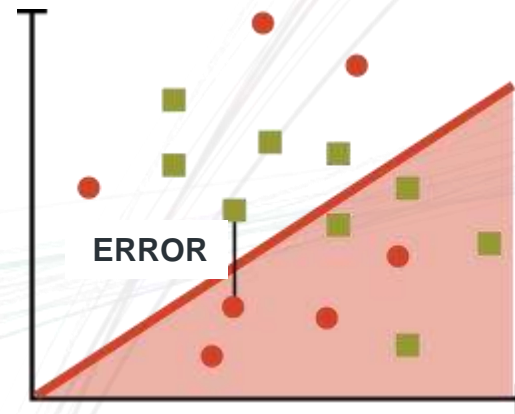
Machine Learning in 8 Slides | Linear Models



Labeled Data → *Algorithm* → *Trained Model*

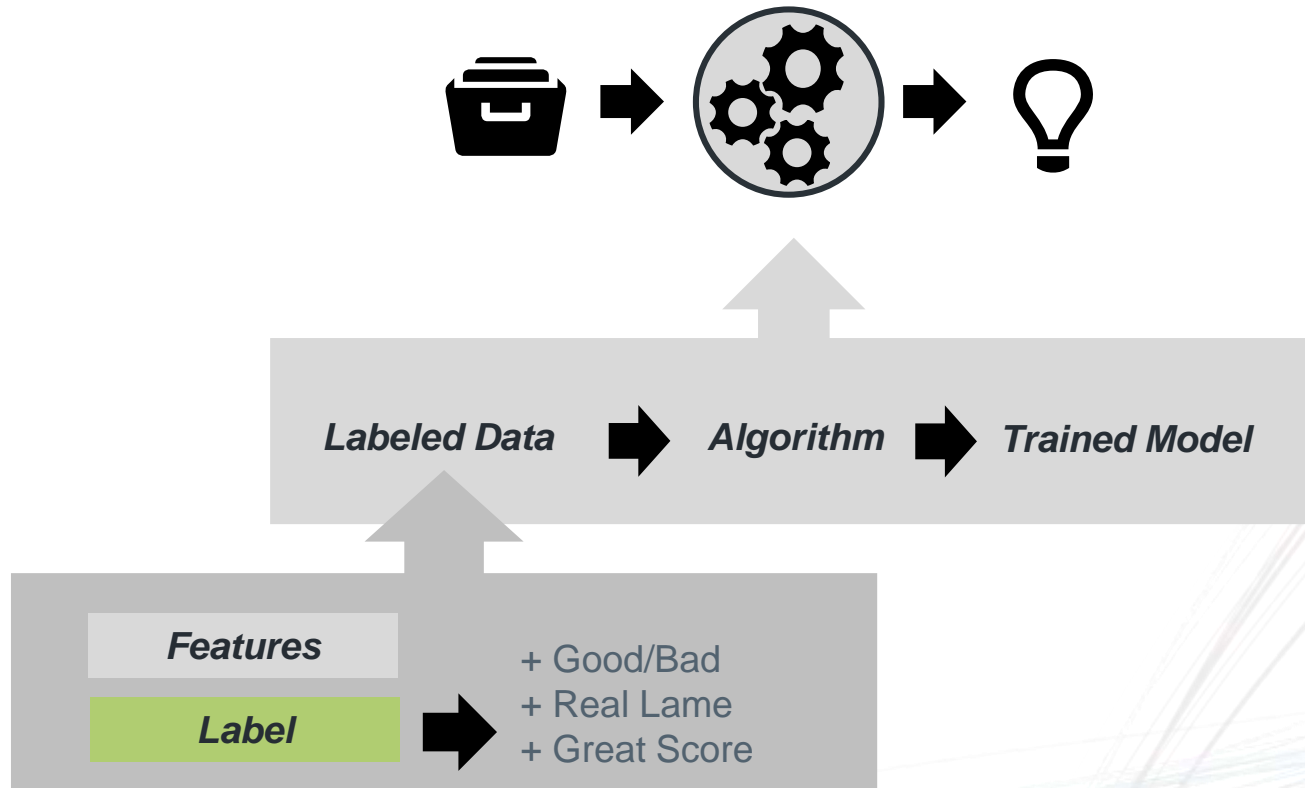


Linear Classification



Linear Regression

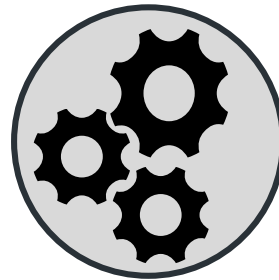
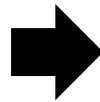
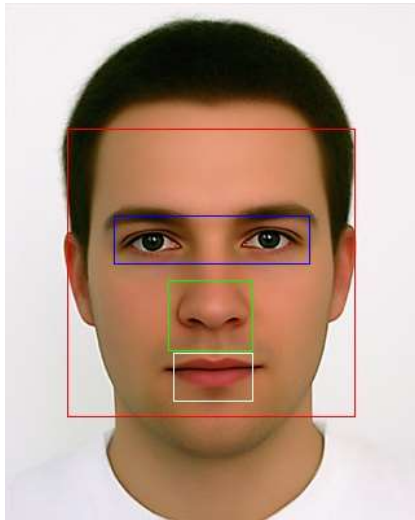
Machine Learning in 8 Slides | Labels & Features



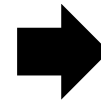
- During training features with labels are given to the algorithm to generate the model
- Both features and labels can be categorical or continuous
- The type of your target affects the flavor of algorithm you chose

Machine Learning in 8 Slides | Labels & Features

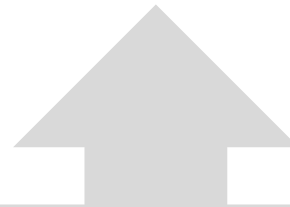
Facial Recognition



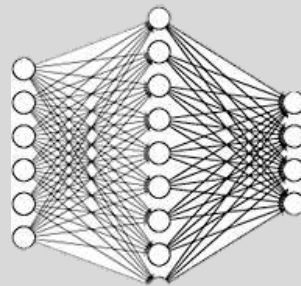
ML Model



Predictions



Geometric features
called "Haars"



Non-linear model



Is this a face?

Machine Learning in 8 Slides | Recap

Term	Description
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How to get started | Training, Data, Tools & Approaches



Training

Books
MOOCs
Instructor Led



Data

Government Agencies
Vendors



Tools

Data Pipeline
Spark
Python



Apache Spark

Cluster aware execution
MLLib for machine
learning



Python

Loosely typed language
Libraries – scikit, pandas
and numpy



R

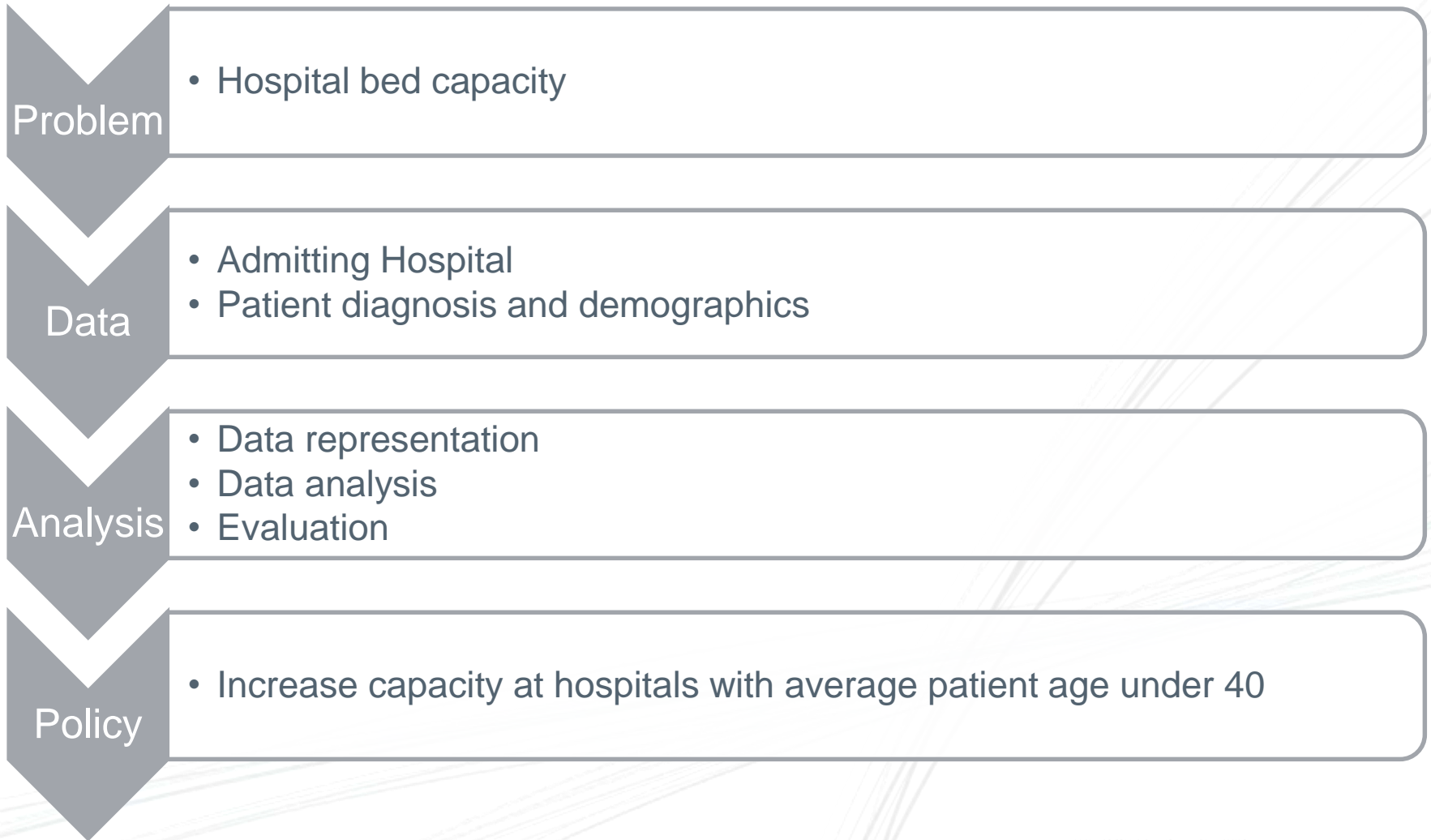
Statistical language,
IDE ecosystem

How to get started | Tool Details

```
/**
 * train the model
 */
val model = KMeans.train(train_data,CLUSTERS,ITERATIONS)

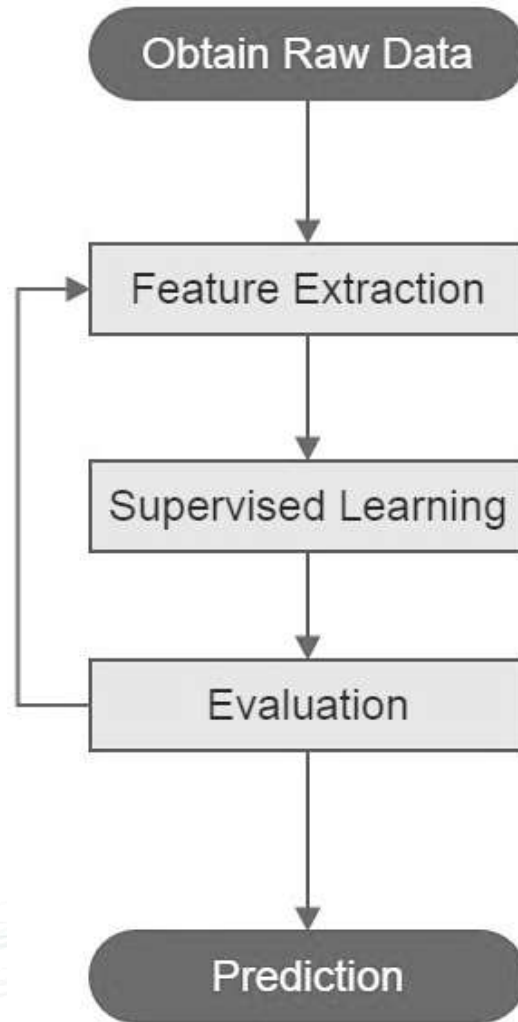
/**
 * score the data
 */
val results = grouped_model_data.map(
  v => (
    v._1
    ,model.predict(
      Array.tabulate[Double](field_cnt)(
        i => zscore(v._2(i),sample_mean(i),sample_stddev(i))
      )
    )
  )
)
results.saveAsTextFile(output_path)
```

How to get started | Data Driven Approach



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How to get started | Data Driven Approach



Let's teach machines

- Demo

ETDAS® Intelligence

Project A

Uses ML and patient Records to predict negative patient outcomes

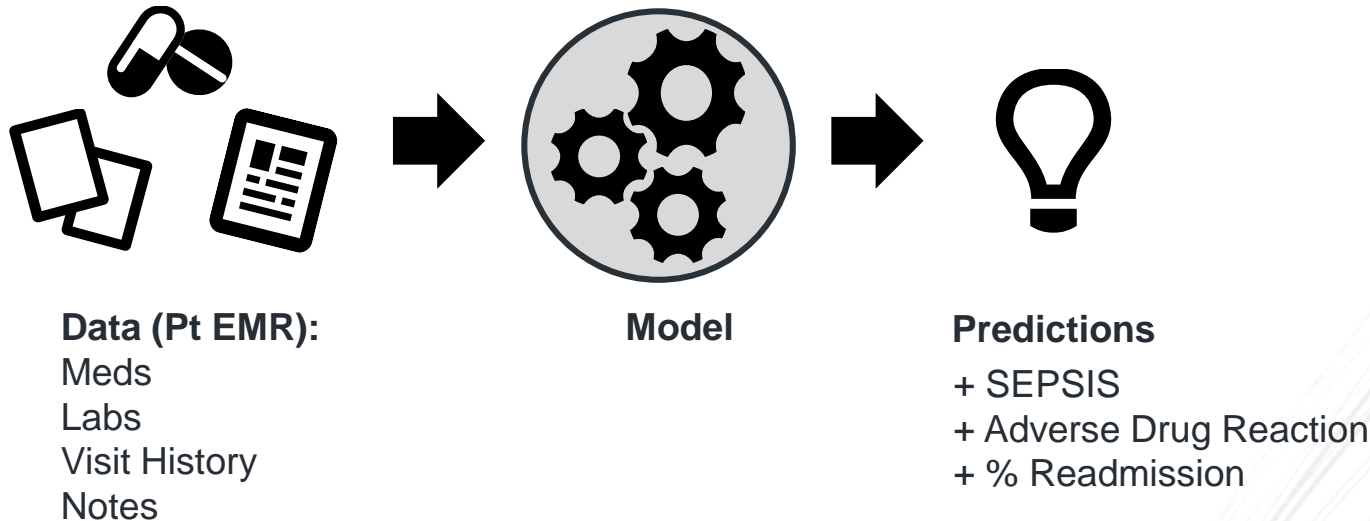
Project B

- Uses user-feedback to innovatively improve the ML Model

Project C

- Grass roots ML project that yielded insight into factors that affect the length of a patients stay.

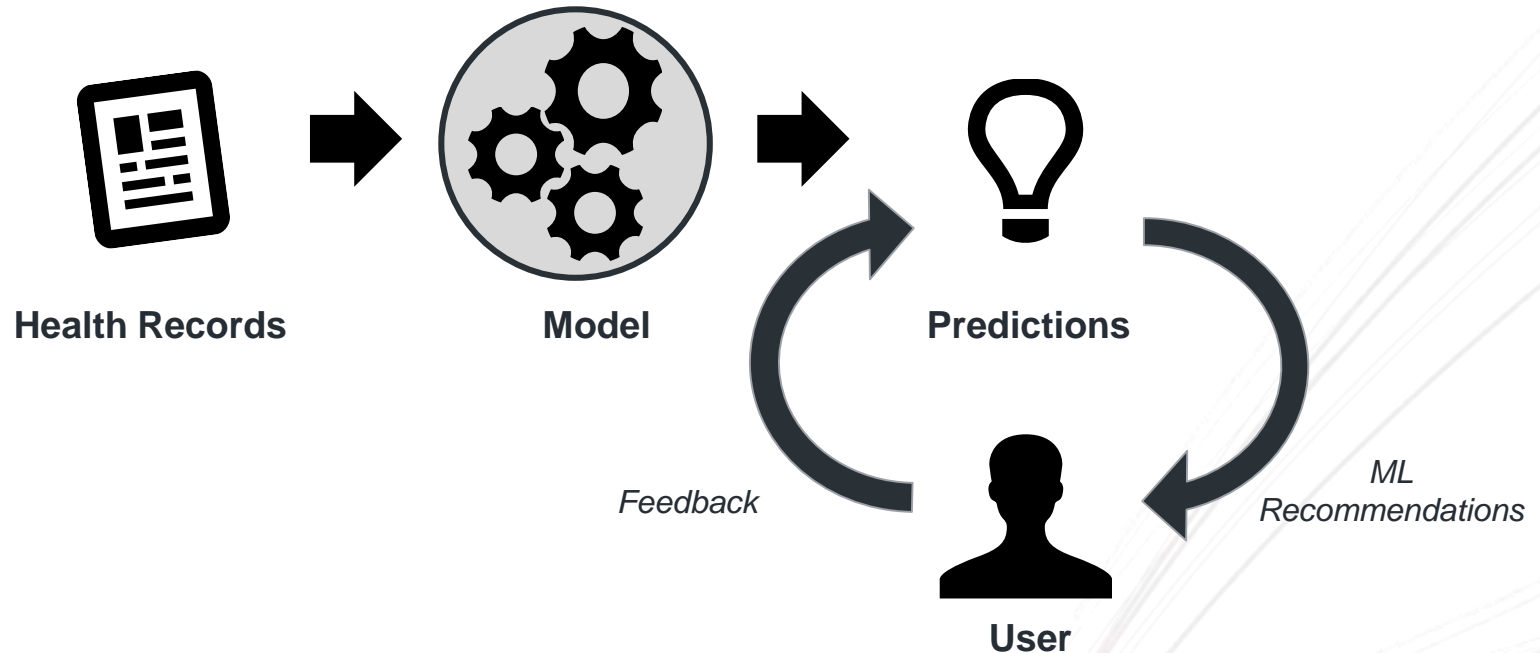
Machine Learning in Healthcare | Project A



- Hospitals already have digital records, because of the ACA
- Can use public algorithm and software to make predictions
- Predictions save lives and/or money

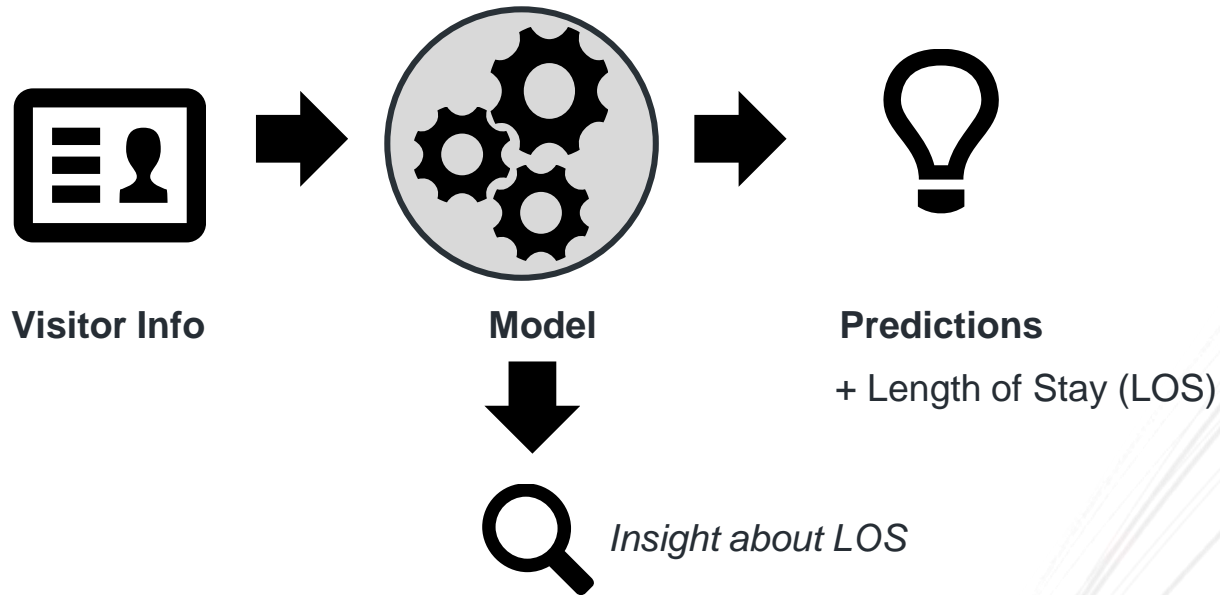
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Machine Learning in Healthcare | Project B



- System uses proprietary model to make recommendations to clinicians
- Hospital decided to improve system by developing a way to incorporate feedback
- Hospital improves the model overtime without having to buy a new system

Machine Learning in Healthcare | Project C



- Interested individuals built a model using visit data
- Predictions were just **OKAY**
- **BUT** inspecting the model allowed the team to learn new factors that were unknown about the different factors affected the clinical outcomes

Questions

The background of the slide features abstract, flowing lines in shades of purple, blue, and green, creating a sense of movement and depth. The lines are layered and overlap, with some appearing more prominent than others. The overall color palette is soft and pastel-like, with the purple being the most vibrant and the blue and green providing a calming contrast.

References

Resource	Location
Presentation link	
EDX Course list (see data)	https://www.edx.org/course
Coursera Course list	https://www.coursera.org/courses/?domains=data-science
Spark Download Page	http://spark.apache.org/downloads.html
Anaconda Python Install Page	http://docs.continuum.io/anaconda/install
R install page	https://cran.r-project.org/
R Studio install page	https://www.rstudio.com/products/rstudio/download/
Ebay Tech Blog on Spark	http://www.ebaytechblog.com/2014/05/28/using-spark-to-ignite-data-analytics/