

# Machine Learning Specialization Welcome

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Machine Learning Specialization  
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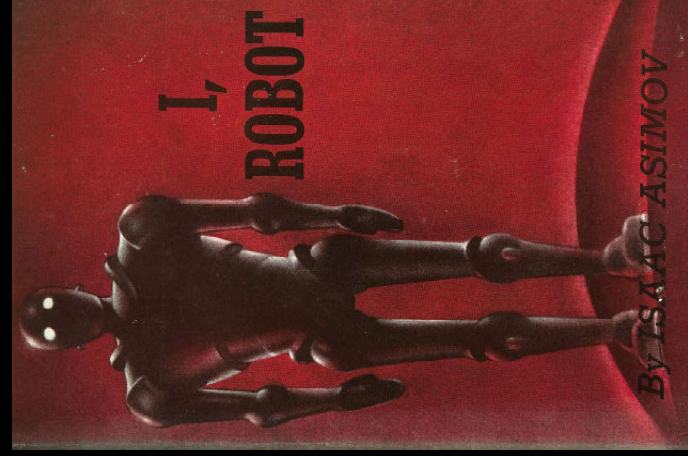
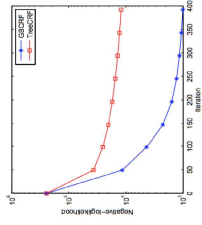
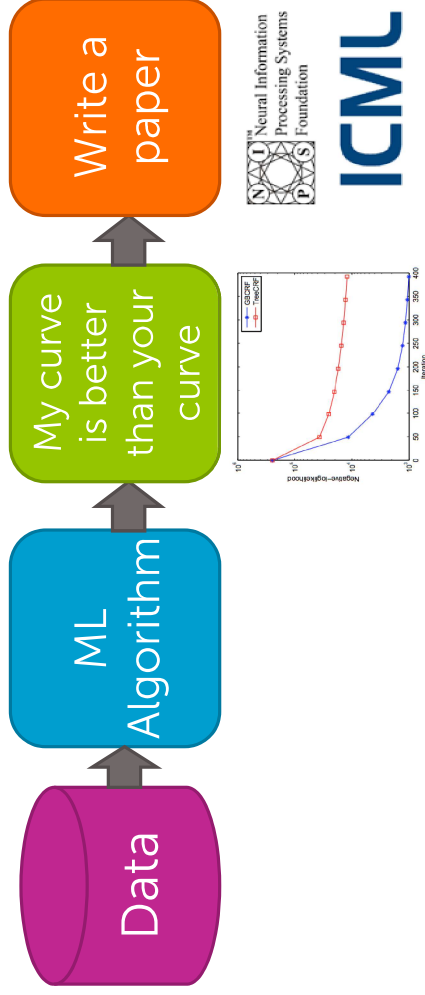
Machine learning is  
changing the world

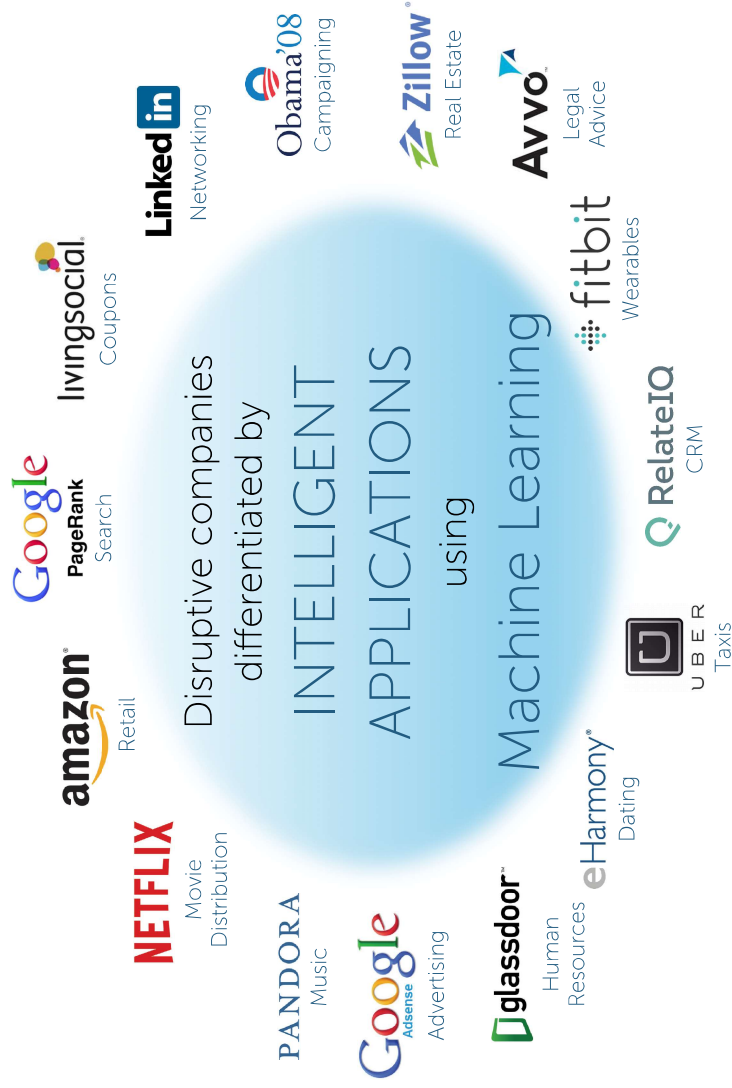
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# Old view of ML



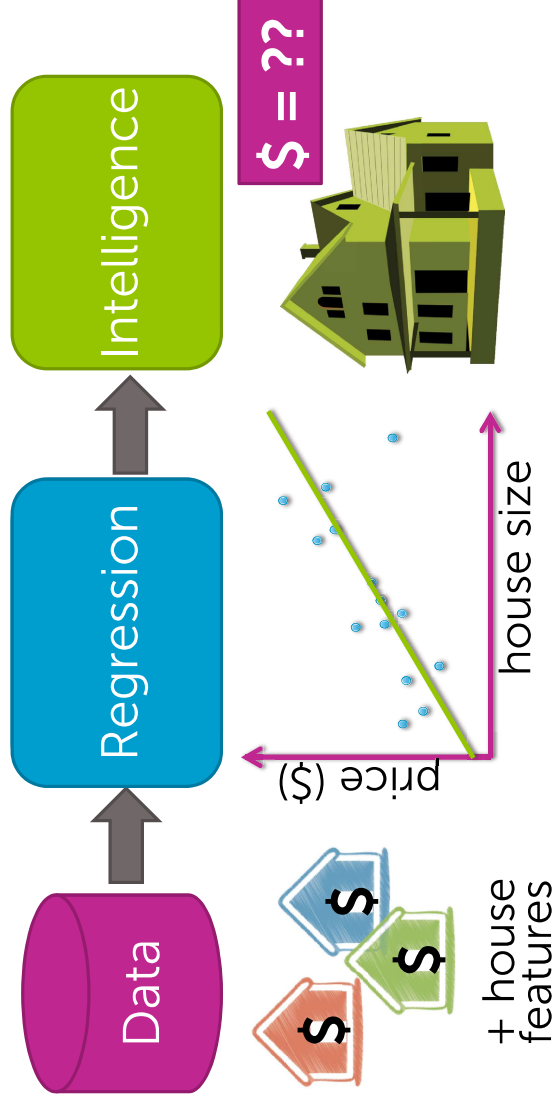


## The machine learning pipeline

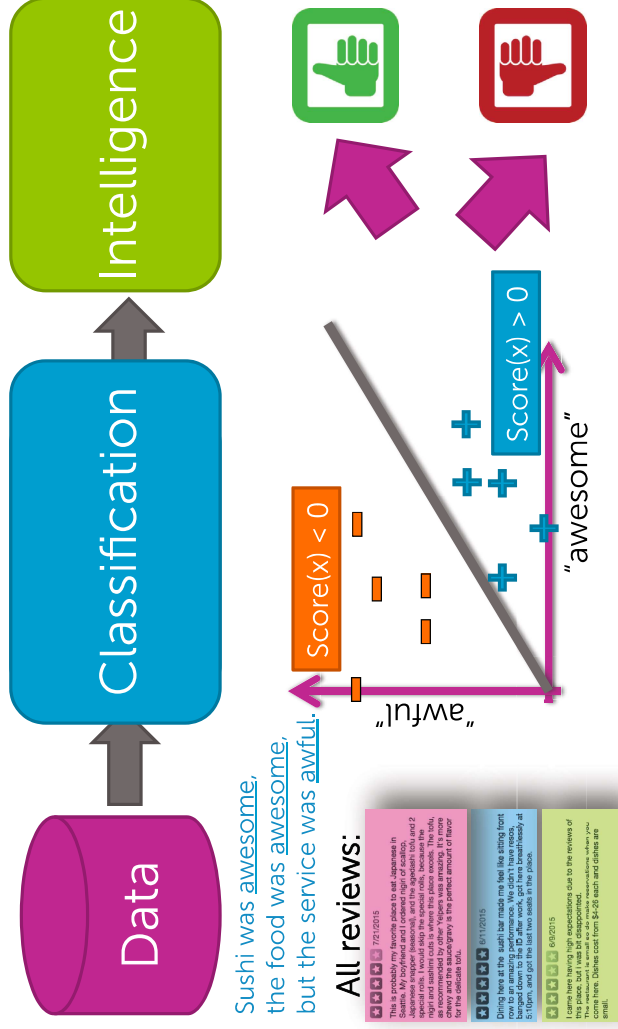


# ML case studies

## Case Study 1: Predicting house prices



## Case Study 2: Sentiment analysis

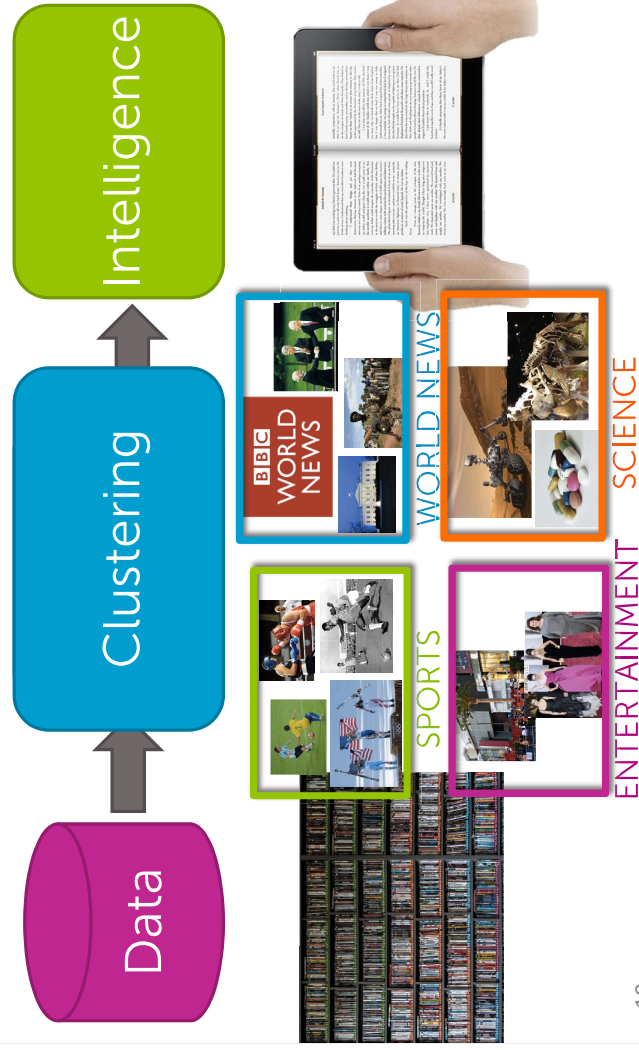


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## Case Study 3: Document retrieval



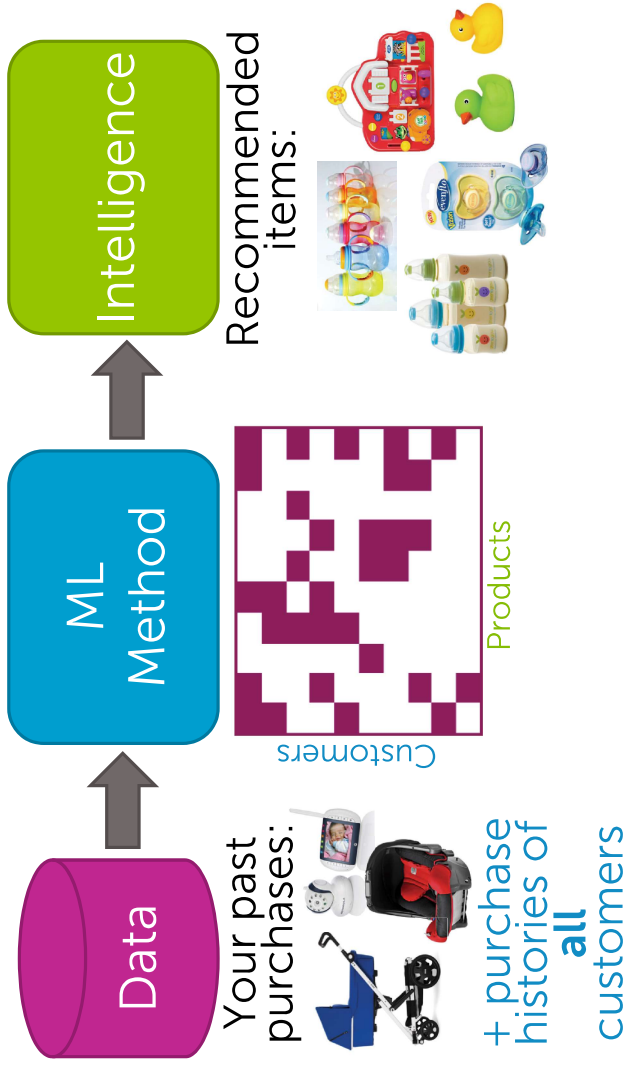
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## Case Study 4:

### Product recommendation

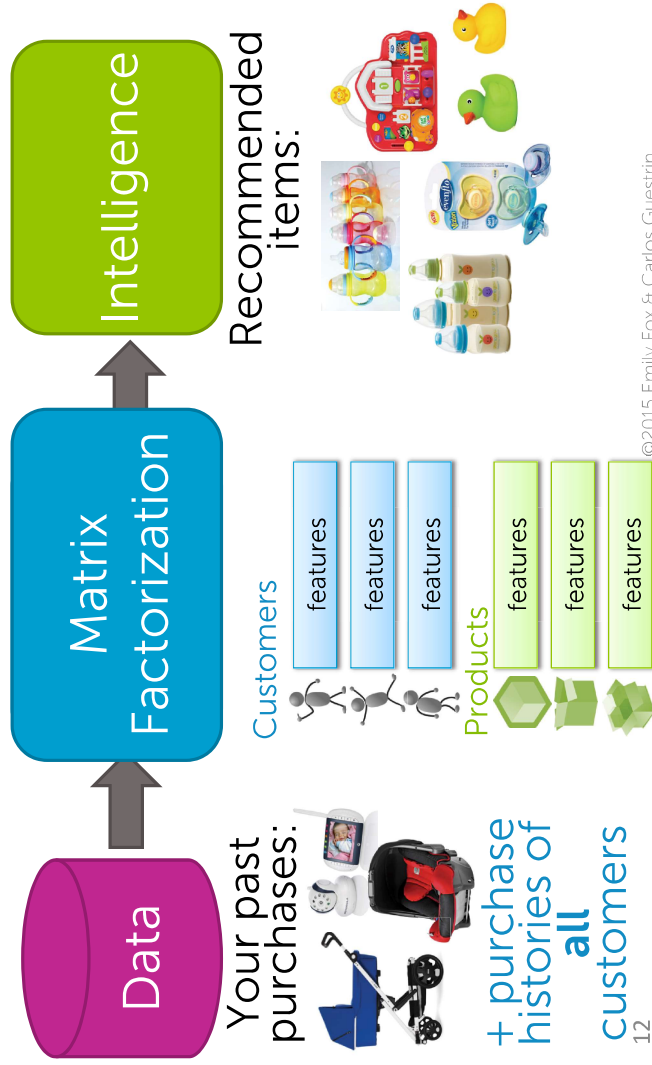


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## Case Study 4:

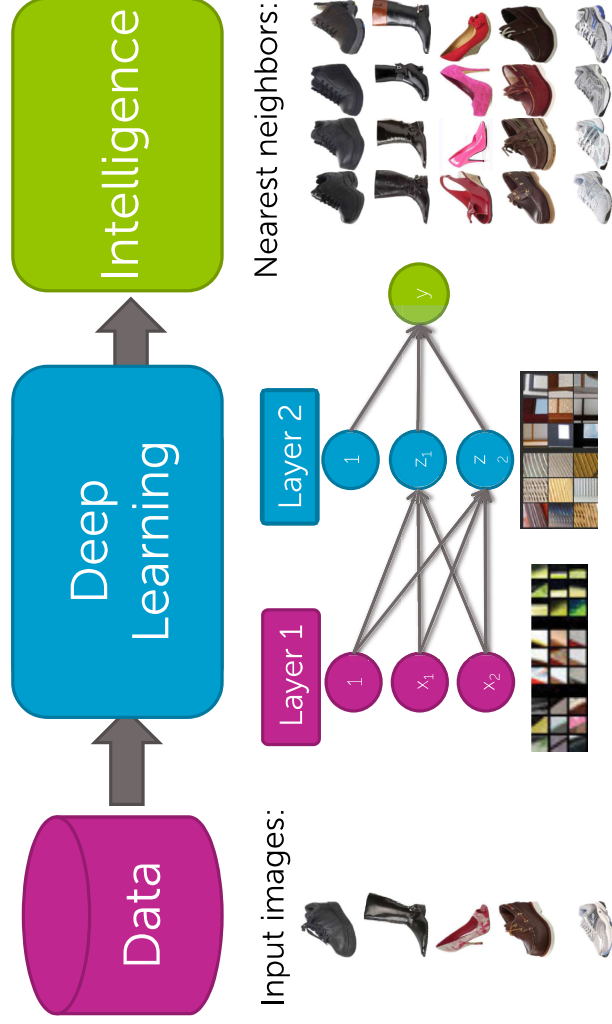
### Product recommendation



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## Case Study 5: Visual product recommender



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# A unique ML specialization

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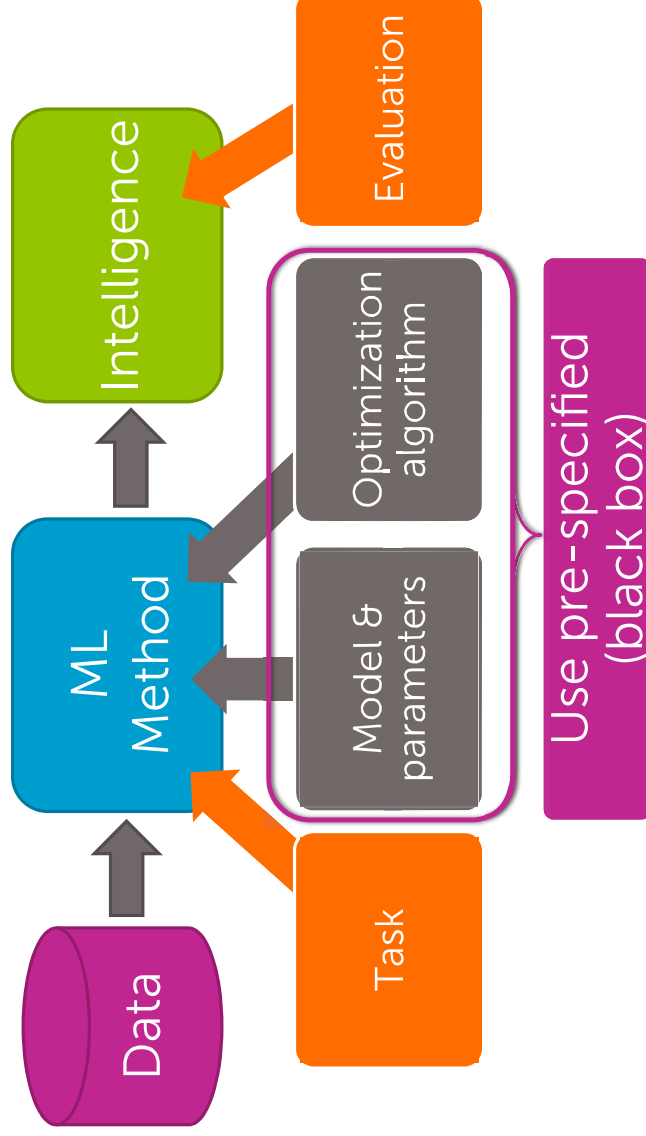
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Not like other ML courses out there...

From use cases to models & algorithms

First course is about building, evaluating and deploying *intelligence* in each case study...





## Subsequent courses provide depth in models & algorithms, but still use case studies

2. Regression
3. Classification
4. Clustering & Retrieval
5. Matrix Factorization & Dimensionality Reduction
6. *Capstone*: Build an Intelligent Application with Deep Learning

## 2. Regression

### *Case study: Predicting house prices*

#### Models

- Linear regression
- Regularization: Ridge (L2), Lasso (L1)

#### Algorithms

- Gradient descent
- Coordinate descent

#### Concepts

- Loss functions, bias-variance tradeoff, cross-validation, sparsity, overfitting, model selection

## 3. Classification

### *Case study: Analyzing sentiment*

#### Models

- Linear classifiers (logistic regression, SVMs, perceptron)
- Kernels
- Decision trees

#### Algorithms

- Stochastic gradient descent
- Boosting

#### Concepts

- Decision boundaries, MLE, ensemble methods, random forests, CART, online learning

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## 4. Clustering & Retrieval

### *Case study: Finding documents*

#### Models

- Nearest neighbors
- Clustering, mixtures of Gaussians
- Latent Dirichlet allocation (LDA)

#### Algorithms

- KD-trees, locality-sensitive hashing (LSH)
- K-means
- Expectation-maximization (EM)

#### Concepts

- Distance metrics, approximation algorithms, hashing, sampling algorithms, scaling up with map-reduce

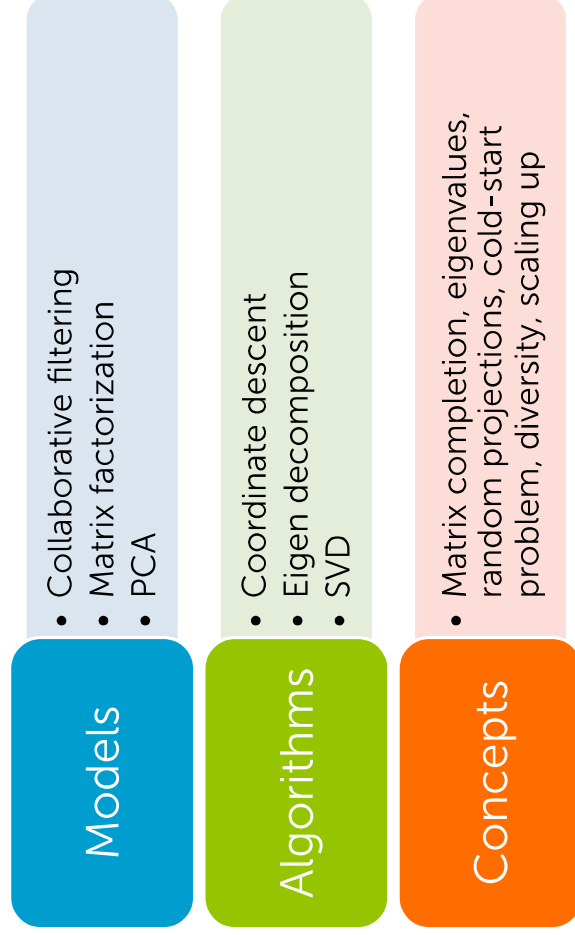
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## 5. Matrix Factorization & Dimensionality Reduction

*Case study: Recommending Products*



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## 6. Capstone:

*An intelligent application using deep learning*

Build & deploy  
a recommender using  
product images and  
text sentiment

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This specialization is for you if...

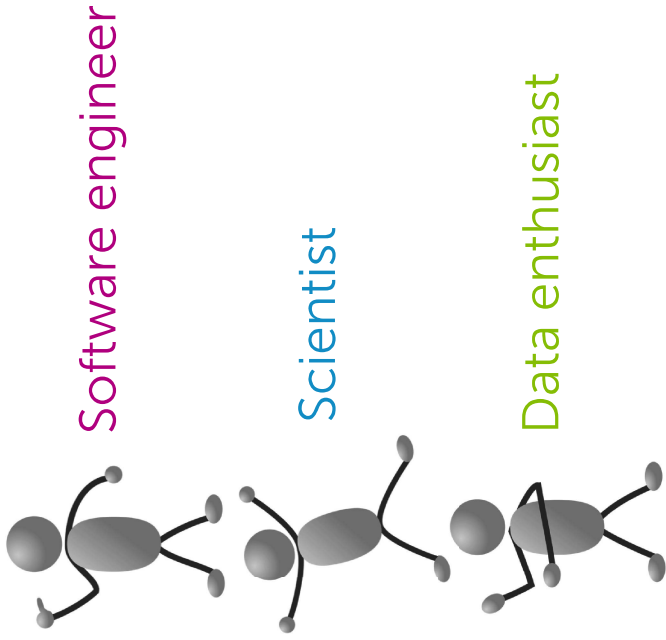
## Level of the specialization

### Motto:

*tough concepts made intuitive  
and applicable*

minimize prereq knowledge  
maximize ability to develop and deploy  
learn concepts through case studies

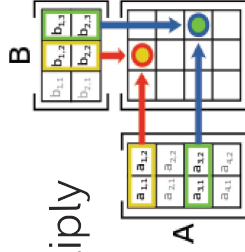
# Target audience



# Math background

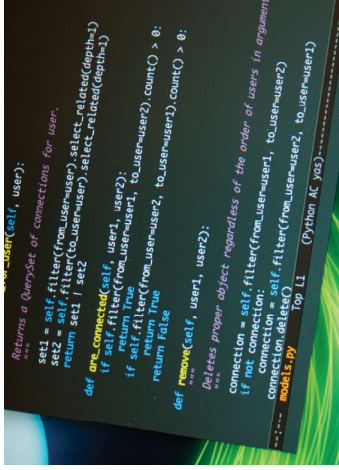


- Basic calculus
  - Concept of derivatives
- Basic linear algebra
  - Vectors
  - Matrices
  - Matrix multiply



# Programming experience

- Basic Python used
  - Can pick up along the way if knowledge of other language



```
def are_connected(self, user1, user2):
    """Returns a QuerySet of connections for user.
    set1 = self.filter(from_user=user).select_related(depth=1)
    set2 = self.filter(to_user=user).select_related(depth=1)
    return set1 | set2
    """
    def are_connected(self, user1, user2):
        return self.filter(from_user=user1, to_user=user2).count() > 0
    return True
    return False

def remove(self, user1, user2):
    """Deletes proper object regardless of the order of users in argument
    connection = self.filter(from_user=user1, to_user=user2)
    if not connection:
        connection.delete()
    """
    models.py Top 11 (Python 3.6.1)
```



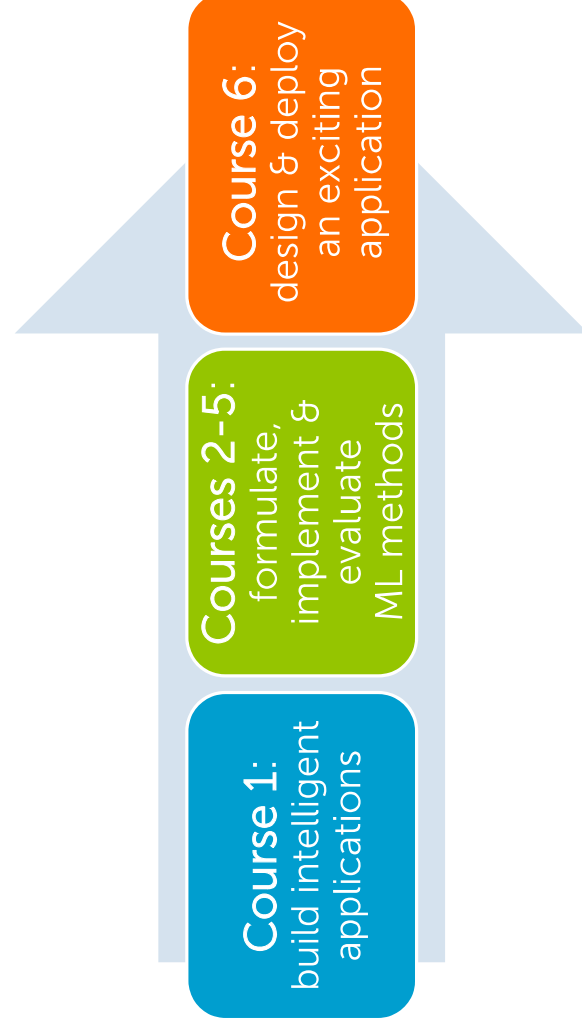
# Computing needs

- Basic desktop or laptop
  - Install and run Python
  - Store a few GB of data
- Access to internet
- Ability to:



You'll be able to do  
amazing things....

## Our journey together...



# The Capstone Project:

## *Build and deploy an intelligent application with deep learning*

# An intelligent recommender

## using images & text



We will do something even more exciting...

