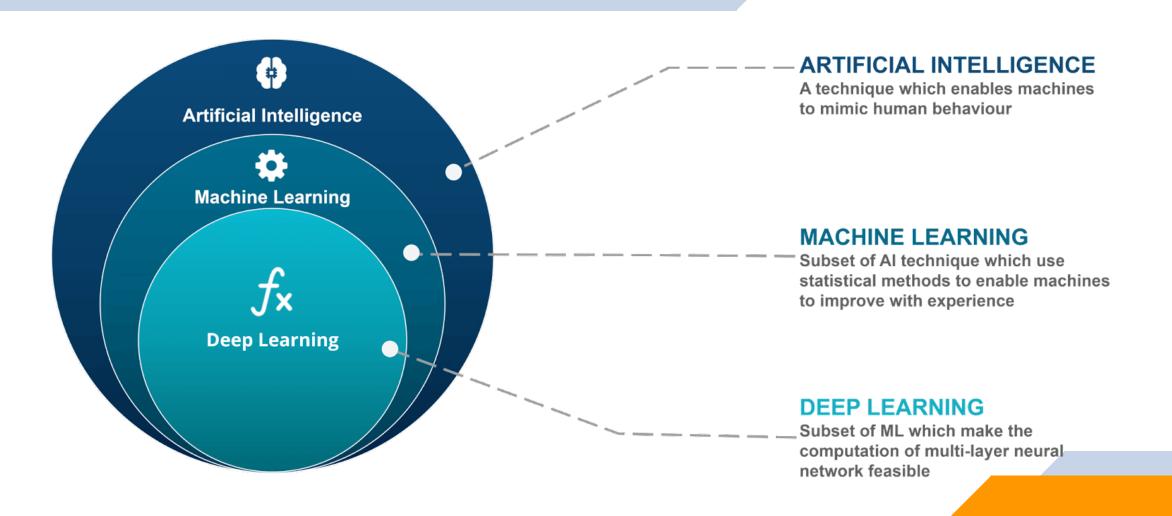
Session

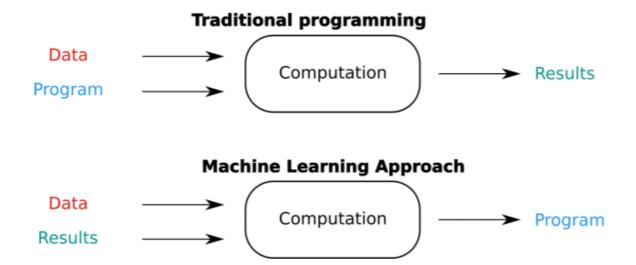
Agenda

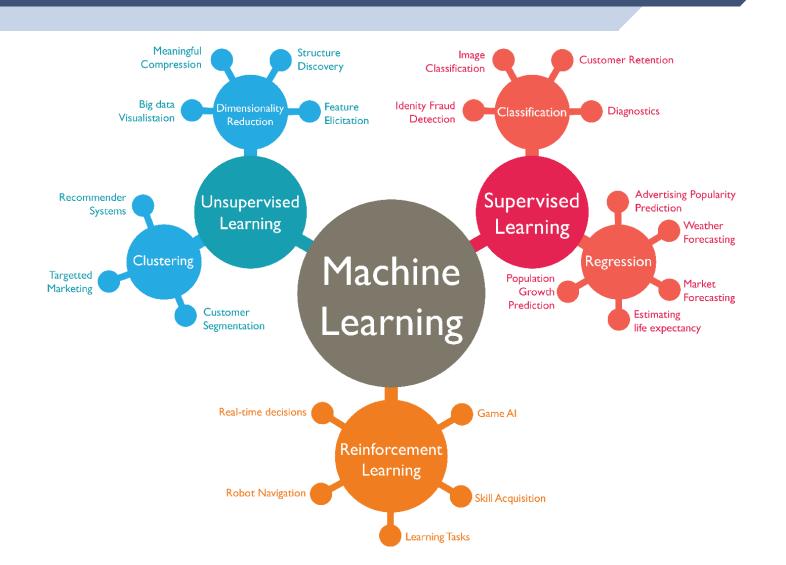
- Introduction to machine learning
- Supervised Learning
 - Linear Regression
 - Gradient Descent
 - Linear Regression vs Logistic Regression



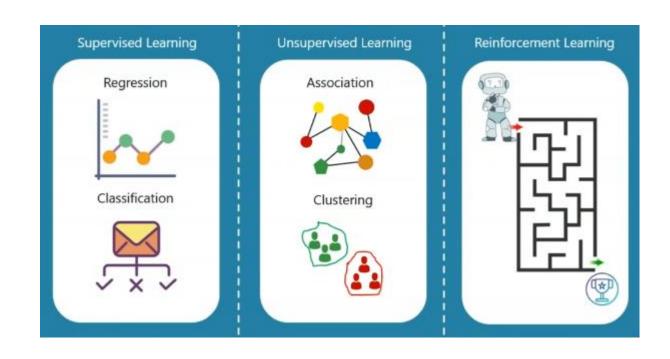
Machine learning is an application of artificial intelligence (AI) that provides systems the ability to automatically learn and improve from experience without being explicitly programmed

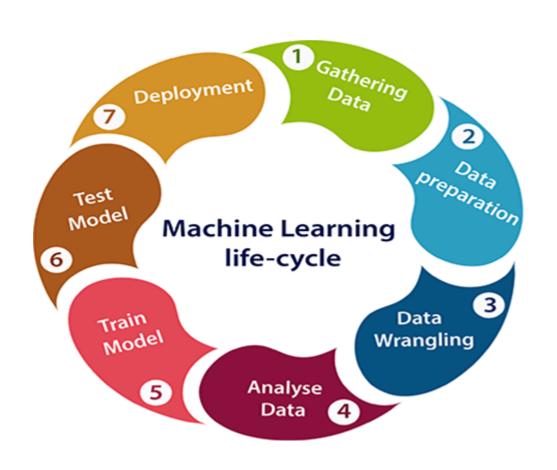






- Supervised Learning
 - Regression
 - Classification
- Unsupervised Learning
 - Clustering
 - Association
- Reinforcement Learning





- Programming Language
 - Python
 - C++
 - R
 - •
- Many Libraries
 - Scikit-learn ←
 - PyTorch
 - TensorFlow
 - Keras

•

Classical machine learning

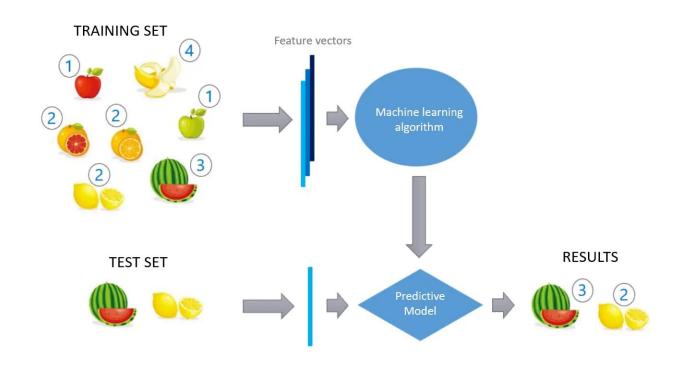
Deep learning frameworks

Agenda

- Introduction to machine learning
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Supervised Learning

Apply what has been learned in the past to new data using labeled examples to predict future events.



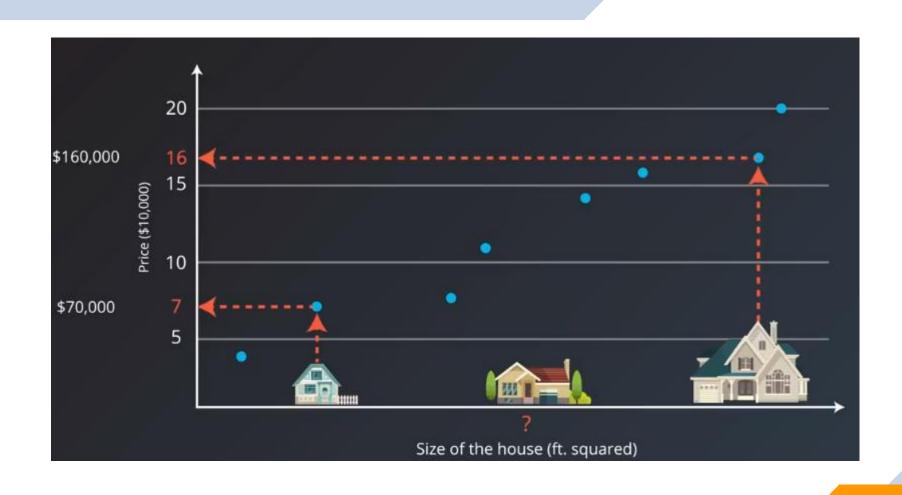
Agenda

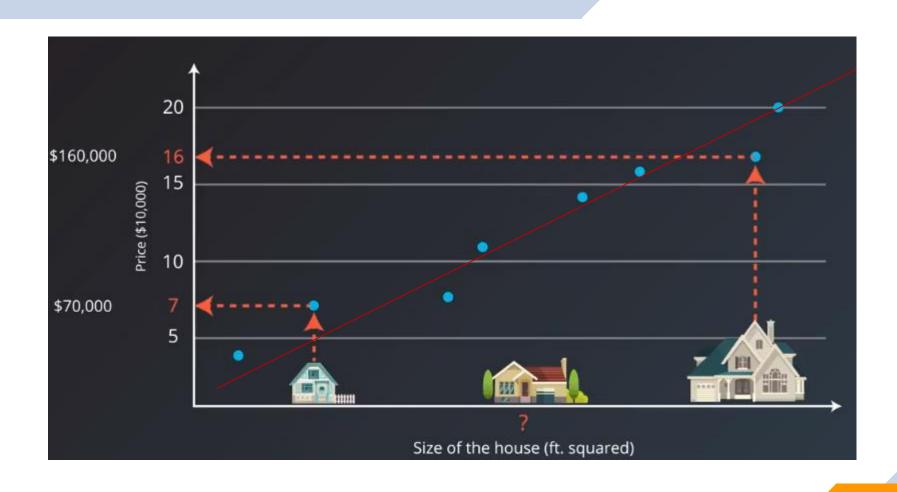
- Introduction to machine learning
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- Let's say we want to estimate the price of the medium sized house
- What do we do?

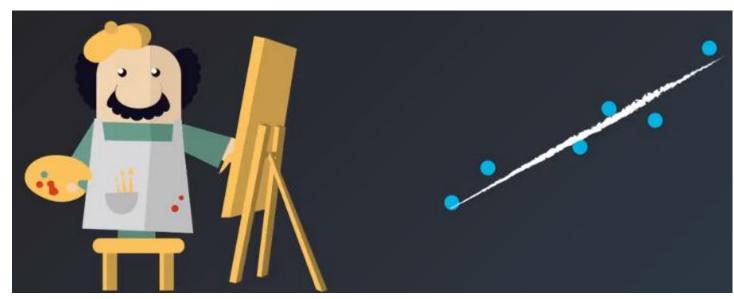




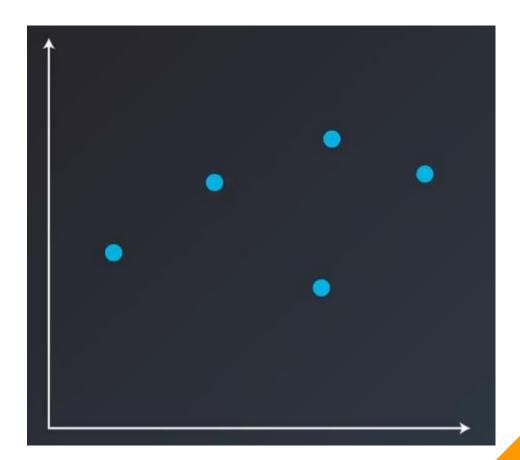




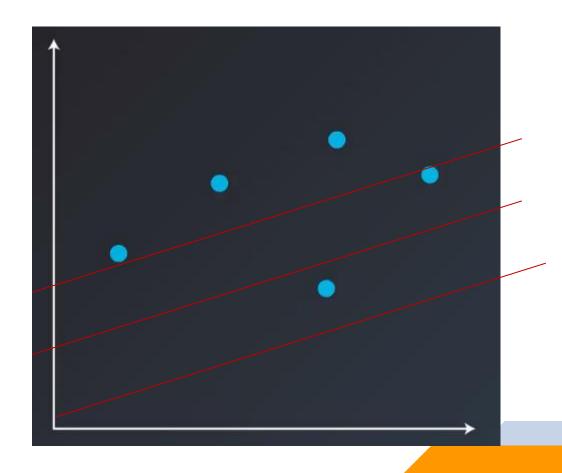
You can think of Linear Regression as a painter drawing best fit line through your data



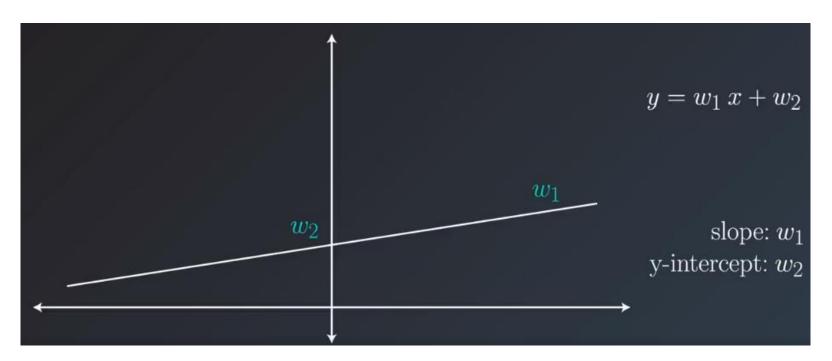
Now how do we find that line?



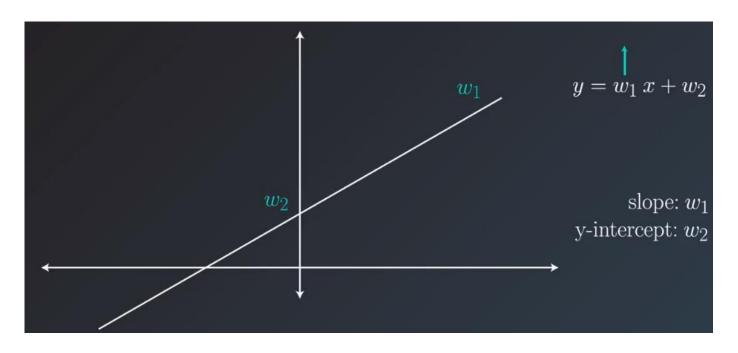
Now how do we find that line?



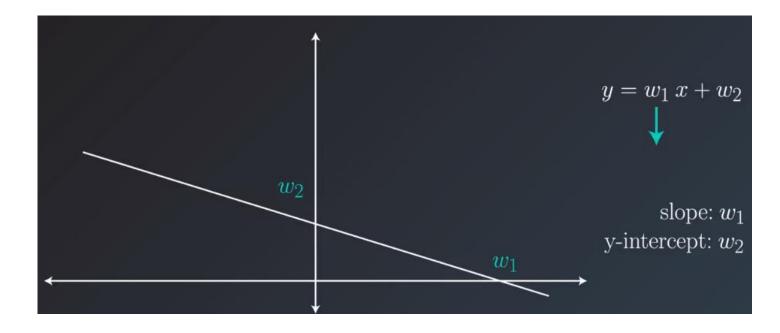
- Now...
- How do we move that line (in mathematical terms)



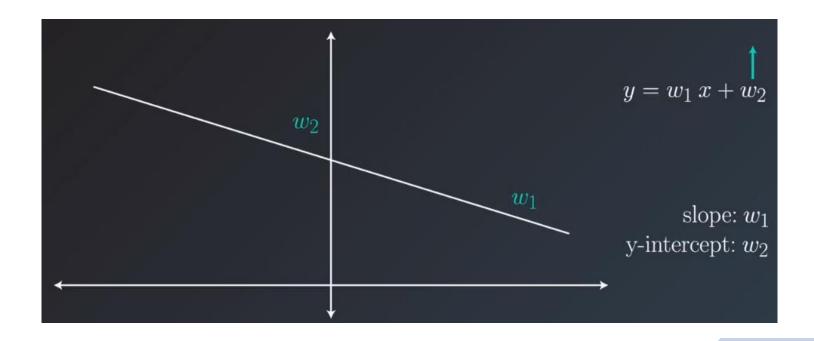
Increase the slope (w1)



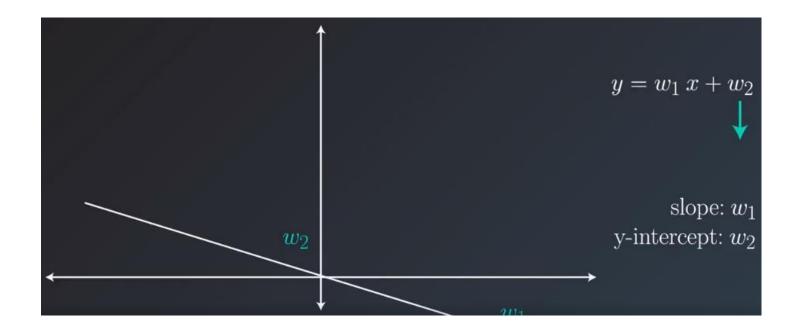
Decrease the slope (w1)



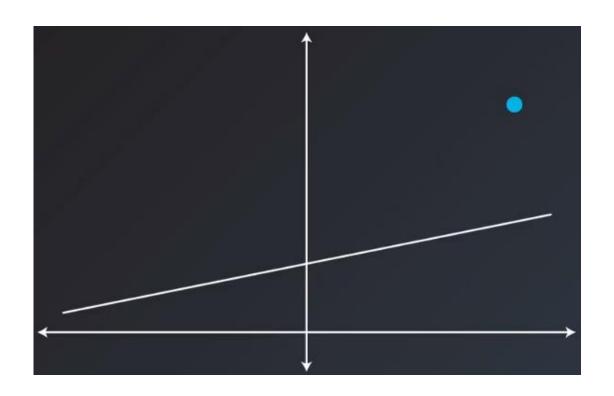
Increase y-intercept (w2)



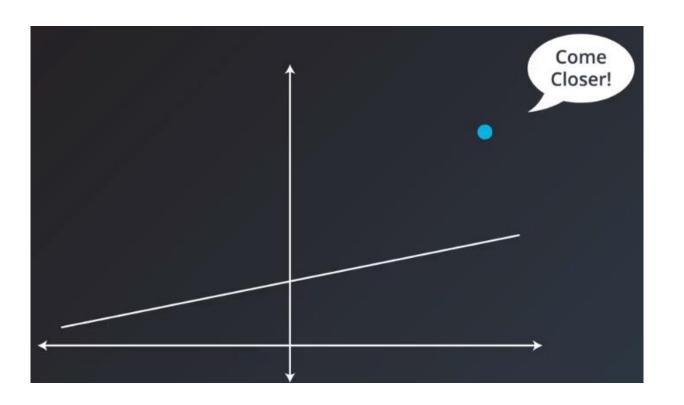
Decrease y-intercept (w2)

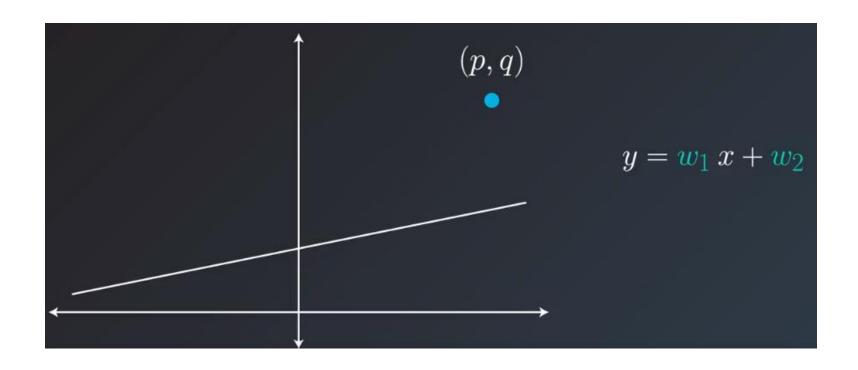


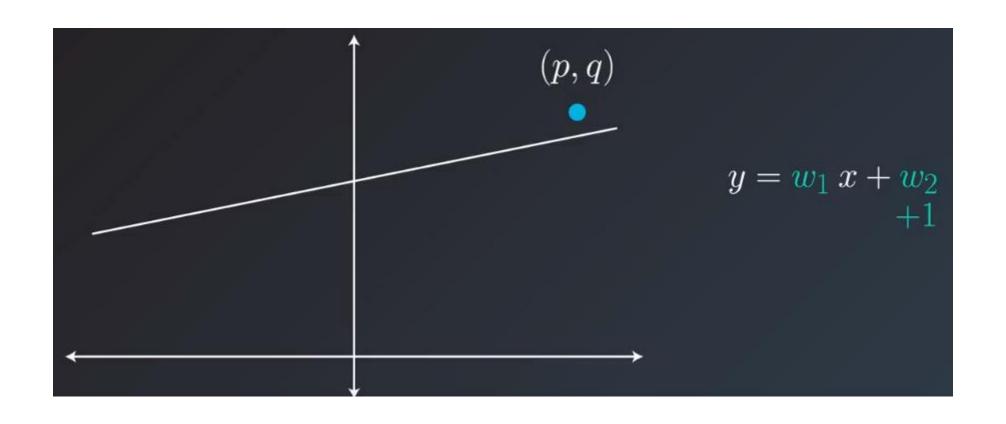
- Now that we know how to move the line.
- Its time to move the line towards the point
- For that we have two techniques:
 - Absolute trick
 - Square trick

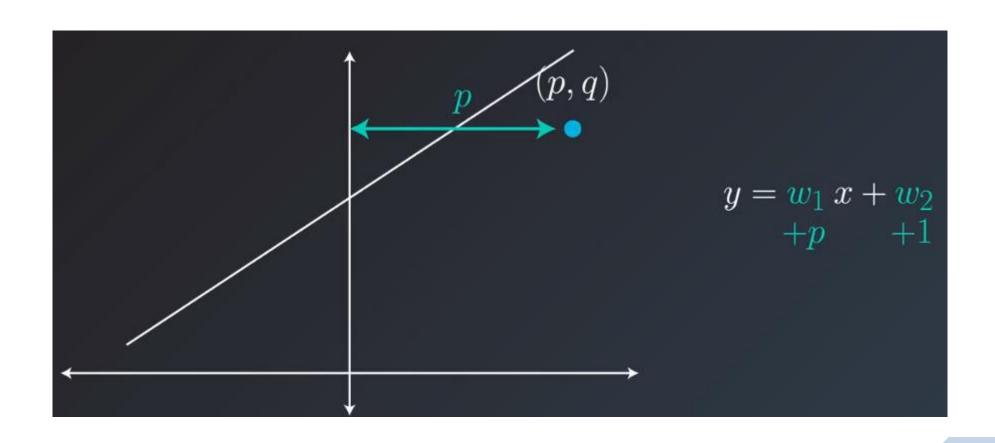


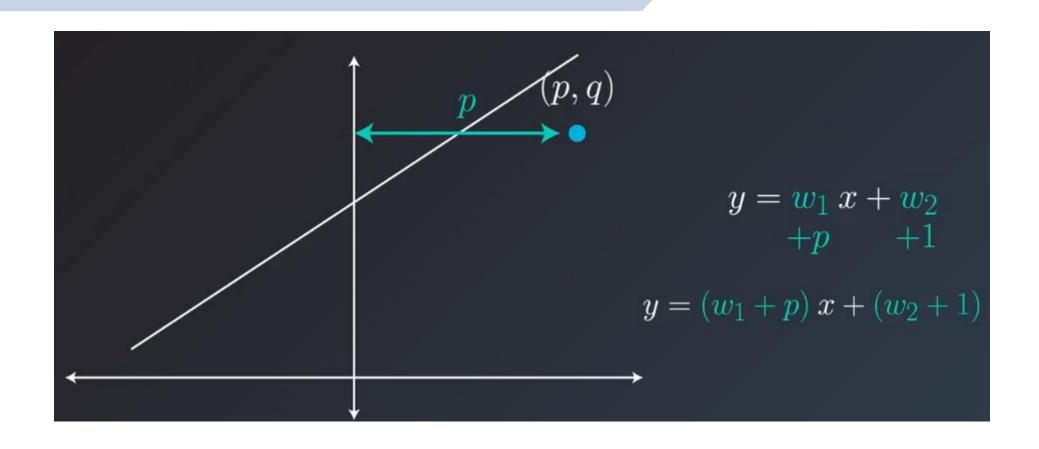
- Absolute trick
- Square trick



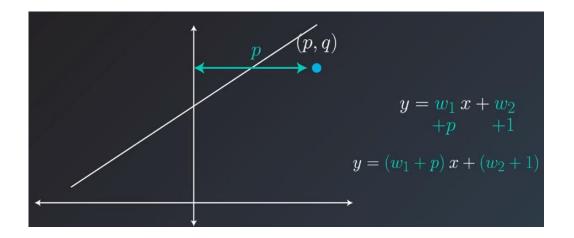


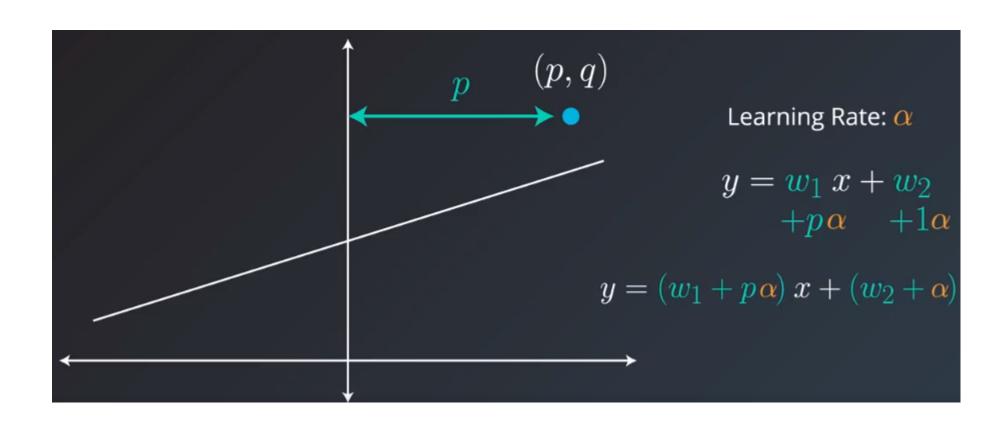




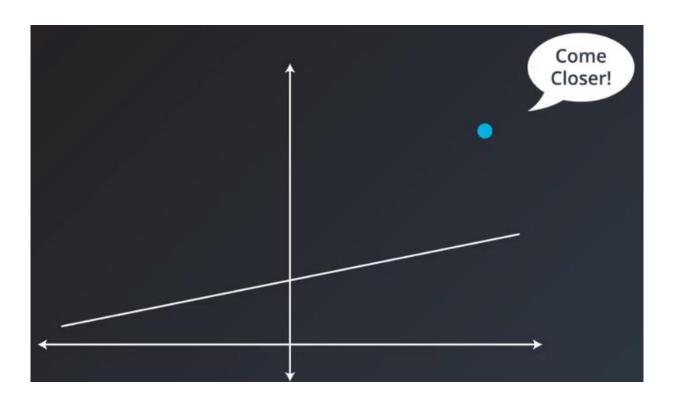


- We successfully moved the line towards the point.
- Notice however that we moved the line too much. And the line has now moved past the point
- We don't like to do that in machine learning. So we have to take smaller steps

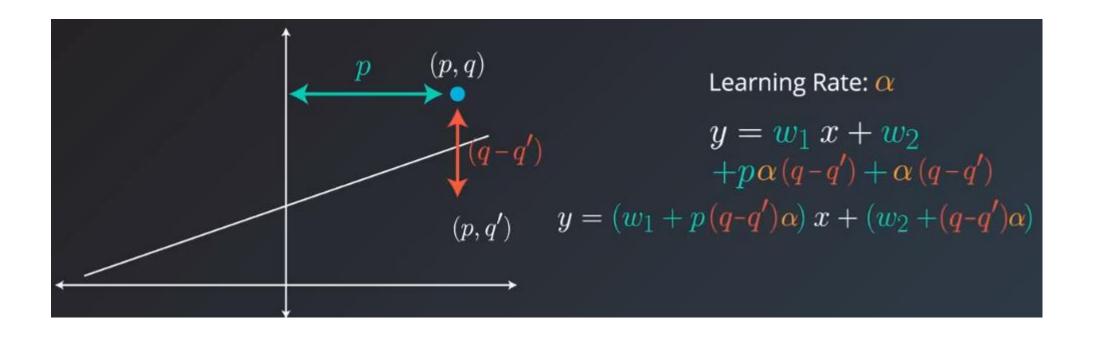




- Absolute trick
- Square trick



- The absolute trick doesn't consider how far or close the point is from the line.
- Taking the same step everytime



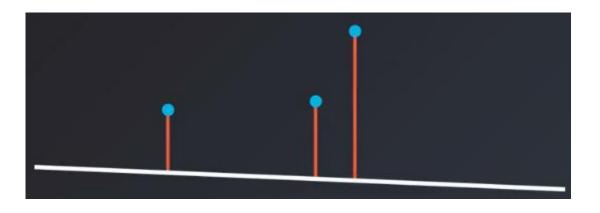
Agenda

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Gradient Descent

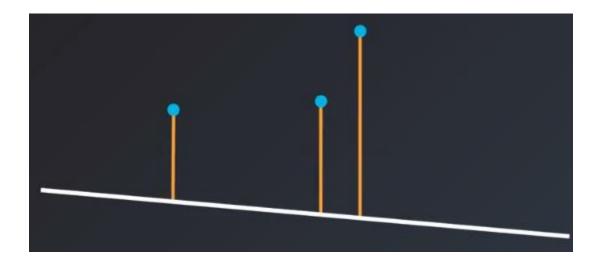


Gradient Descent



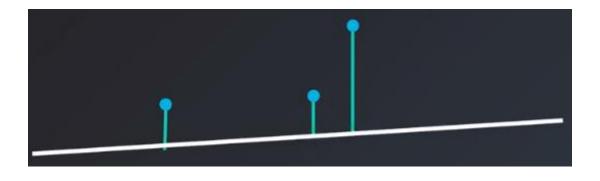
Error:

Gradient Descent



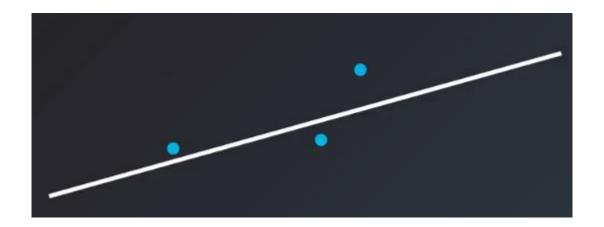
Error:

Gradient Descent

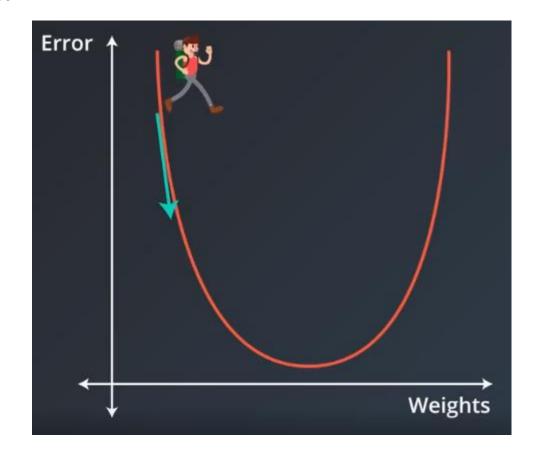


Error:

Gradient Descent



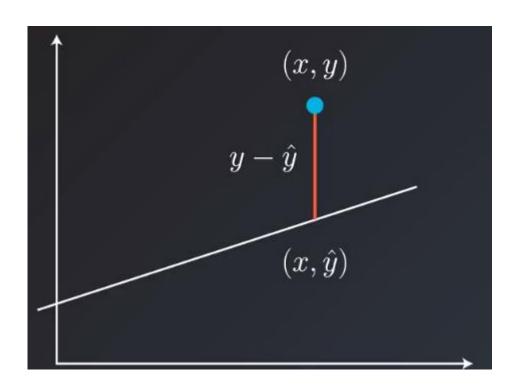
Gradient Descent



Error Function - Gradient of Error Function $w_i o w_i - lpha rac{\partial}{\partial w_i} Error$

- Error Functions:
 - Mean Absolute Error
 - Mean Squared Error

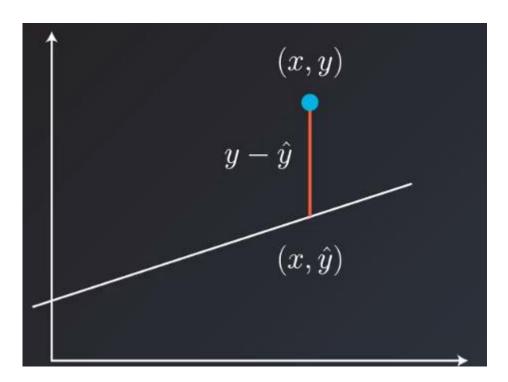
- Error Functions:
 - Mean Absolute Error
 - Mean Squared Error



Error Functions:

- Mean Absolute Error
- Mean Squared Error

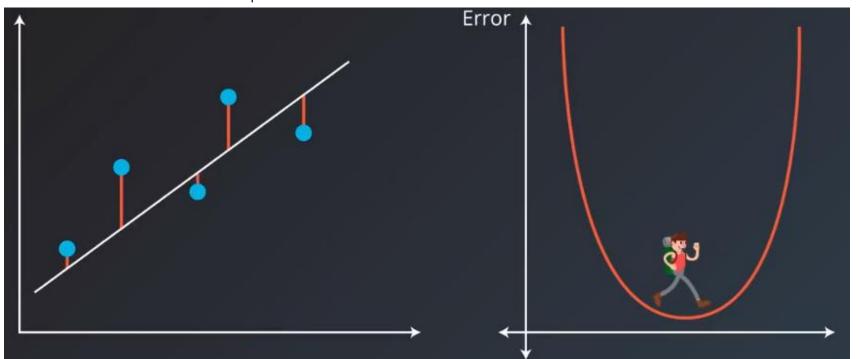
$$Error = \frac{1}{m} \sum_{i=1}^{m} |y - \hat{y}|$$



- Error Functions:
 - Mean Absolute Error
 - Mean Squared Error



- Error Functions:
 - Mean Absolute Error
 - Mean Squared Error



- Error Functions:
 - Mean Absolute Error
 - Mean Squared Error

$$Error = \frac{1}{2m} \sum_{i=1}^{m} (y - \hat{y})^2$$

- Error Functions:
 - Mean Absolute Error
 - Mean Squared Error

1) Absolute Error Function

$$Error = \frac{1}{m} \sum_{i=1}^{m} |y - \hat{y}|$$

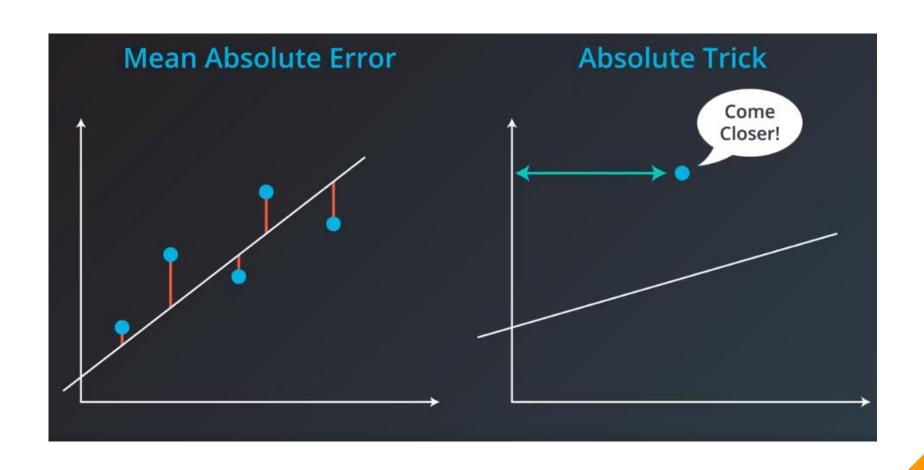
2) Squared Error Function

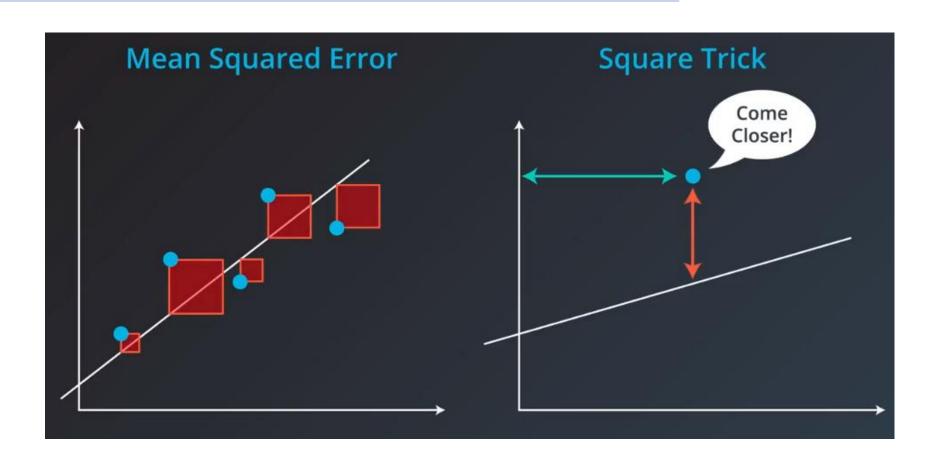
$$Error = \frac{1}{2m} \sum_{i=1}^{m} (y - \hat{y})^2$$

- We have learned 2 methods to decrease the error. Tricks, and Error functions.
- Turns out they are both the exact same thing

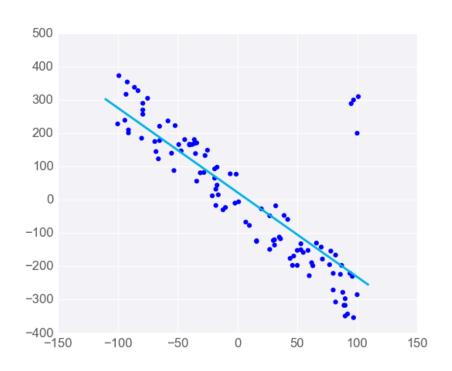
Tricks

Error Functions

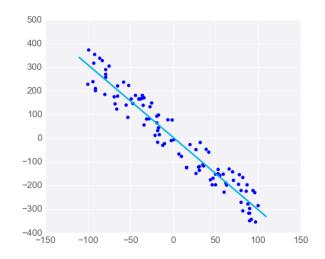


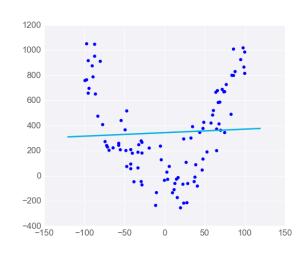


- Linear regression rules:
 - Affected by outliers.
 - □ It works best when data is linear



- Linear regression rules:
 - Affected by outliers.
 - It works best when data is linear





Linear Regression in scikit-learn

```
>>> from sklearn.linear_model import LinearRegression
>>> model = LinearRegression()
>>> model.fit(x_values, y_values)
```

```
>>> print(model.predict([ [127], [248] ]))
[[ 438.94308857, 127.14839521]]
```

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Supervised Learning: Linear Regression vs Logistic Regression



Regression

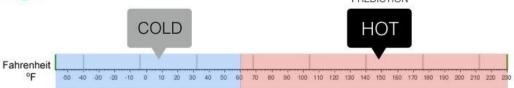
What is the temperature going to be tomorrow?





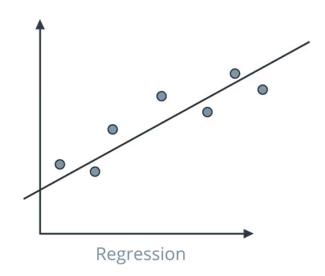
Classification

Will it be Cold or Hot tomorrow?

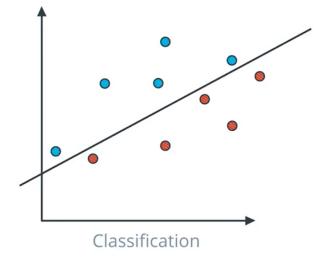


Supervised Learning: Linear Regression vs Logistic Regression

REGRESSION AND CLASSIFICATION



Regression returns a numeric value
We use Linear Regression



Classification returns a numeric state.
We use Logistic Regression

Supervised Learning: Linear Regression vs Logistic Regression

Logistic Regression in scikit-learn

```
# import the class
from sklearn.linear_model import LogisticRegression
# instantiate the model (using the default parameters)
logreg = LogisticRegression()
```