# Geometric Interpretation of ML

### Introduction

- Steps:
  - Idea (elevator pitch)
  - Proof (formalize, correctness, existence, unique)
  - Compute (implementation, runtime, Big-O)
  - Useful (real application, type of question)
- Algorithm development
  - Generalization
  - Extension

### Introduction

#### 1. Fit

- Average
- Linear regression
- Piecewise linear regression
- Exponential curve

#### 2. Axis transformation

- PCA
- Kernel methods

#### 3. Separation

- LDA
- Logistic regression
- SVM, perceptron

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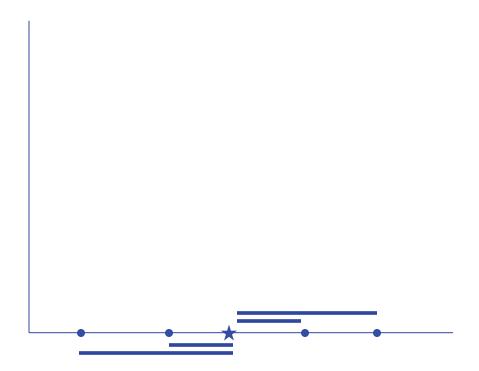
# Fit



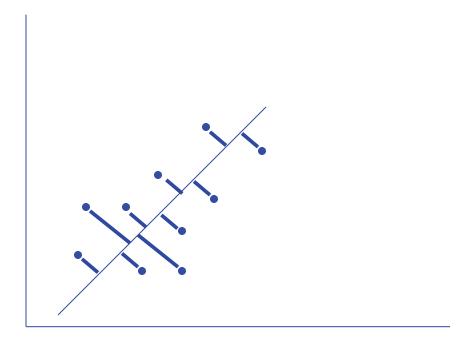
#### Average

### Average

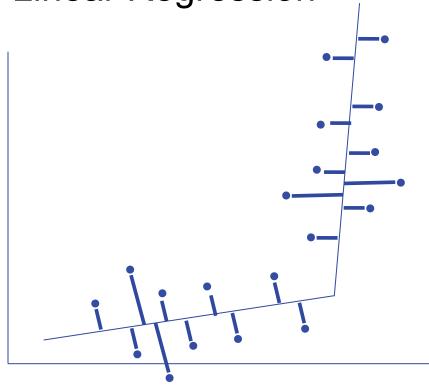
### Average



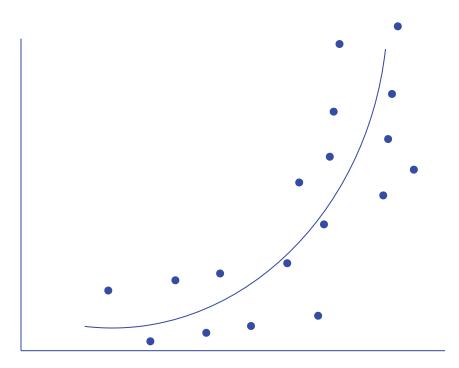
#### Least Squares



Piecewise Linear Regression

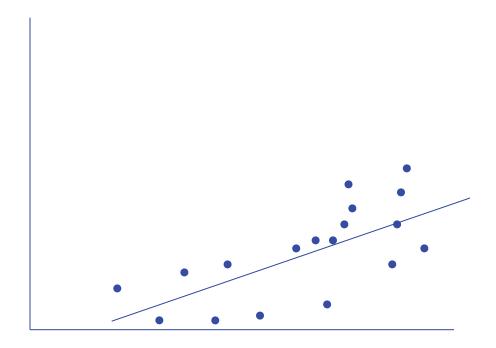


#### Logarithmic Curve

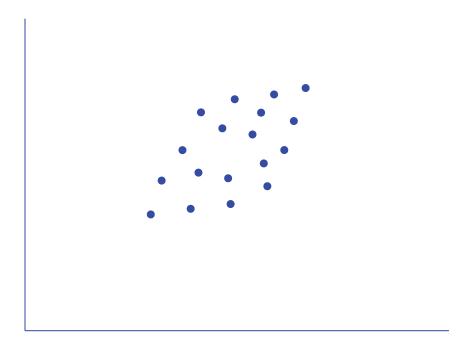


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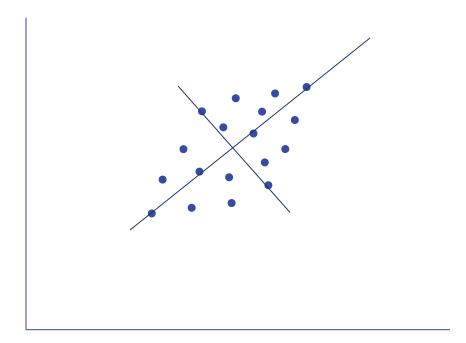
#### Logistic Axis



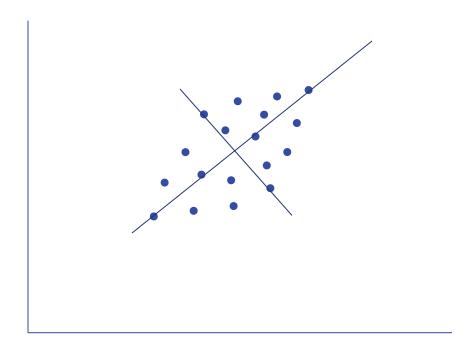




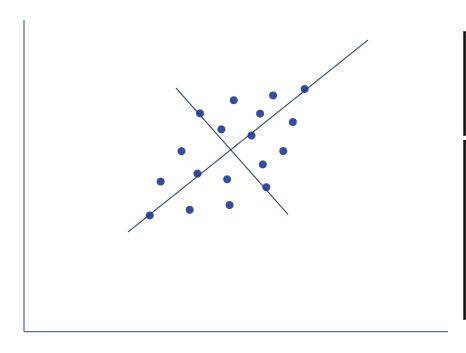
Principle Component Analysis (PCA)



#### Principle Component Analysis



#### Principle Component Analysis (PCA)



```
data = [[1, 5, 10, 300],
[10, 50, 100, 3000],
[2, 10, 20, 600]]
```

```
data = [[1,0,0,2,3,2],

[1,0,0,0,0,0],

[5,0,0,0,0,0],

[3,0,0,4,6,4],

[2,0,0,6,9,6],

[0,0,0,2,3,2],

[8,0,0,2,3,2]]
```

# Motivating Example

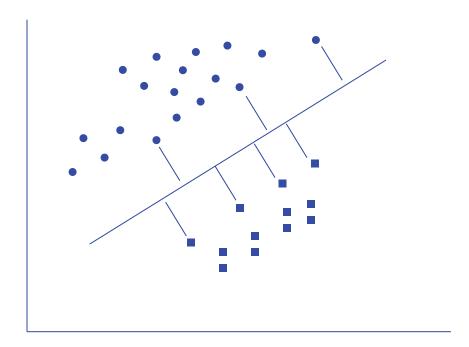


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# Separation

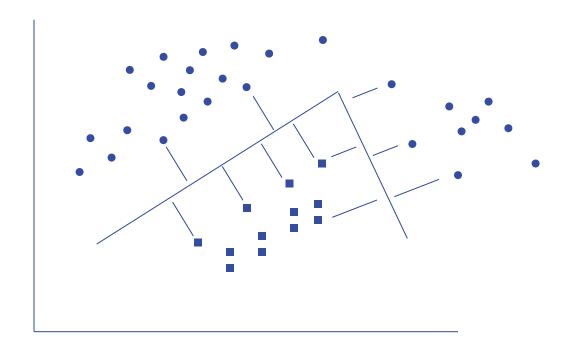
# Separation

#### **Linear Discriminant Analysis**

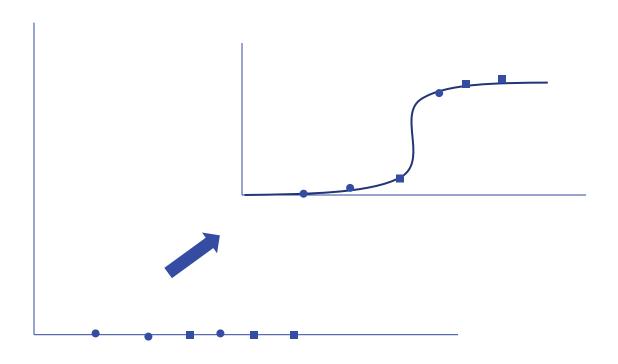


# Separation

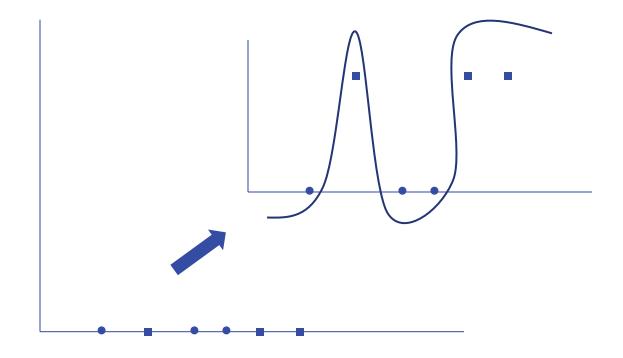
#### **Linear Discriminant Analysis**



#### Logistic Regression



#### Support Vector Machine/Perceptron



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