



# Machine Learning: How Machines Learn from Data

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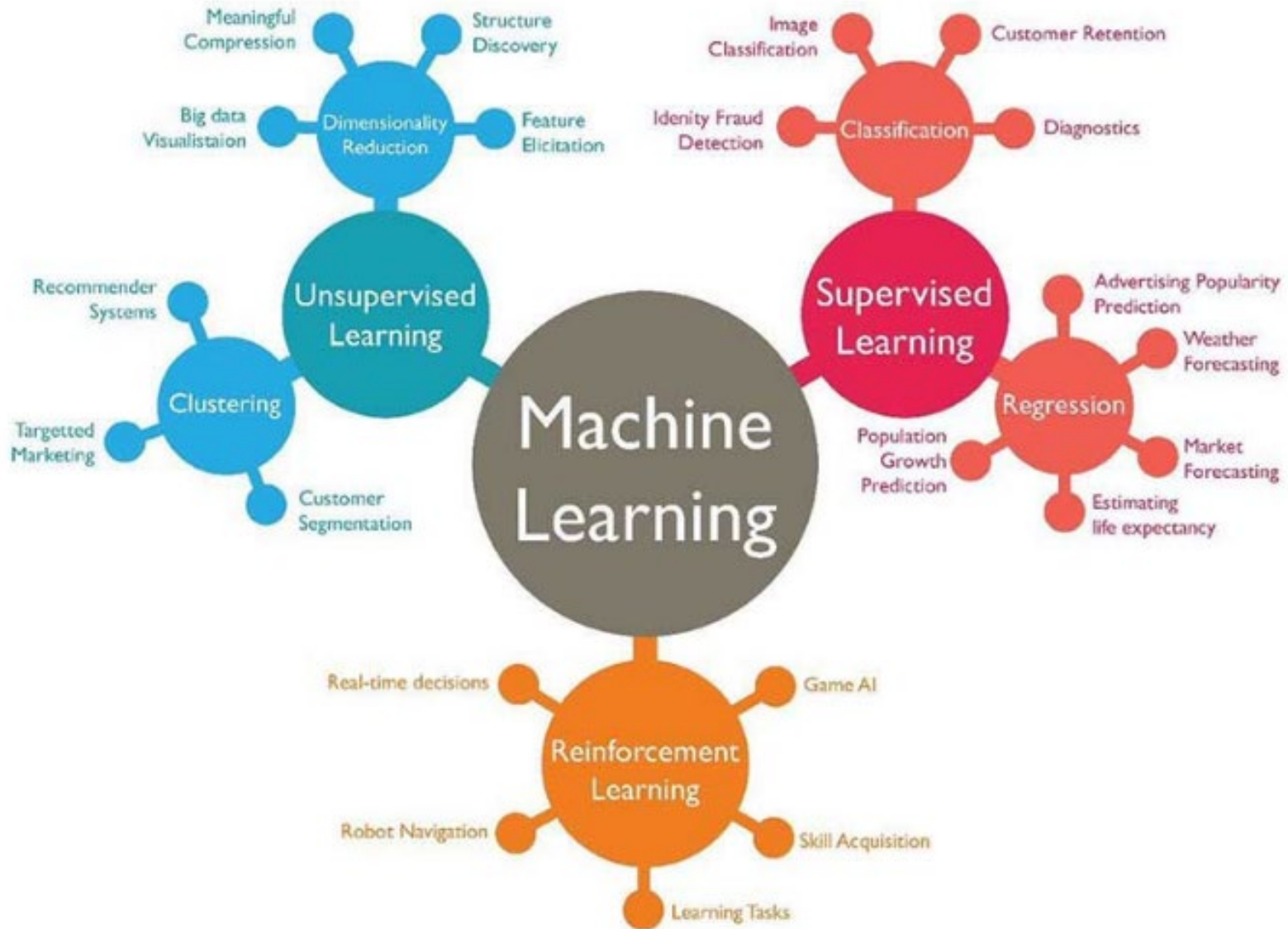
1. Introduction to ML
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  - Classification: Support Vector Machine
3. Unsupervised Learning
  - Clustering: K-Means
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# 1. Introduction to ML

Field of study that gives the computer the ability to **learn** without being explicitly programmed.

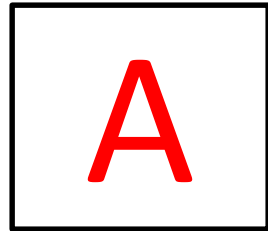
Arthur Samuel, 1959

# Types of ML

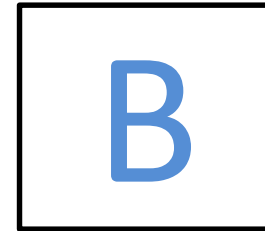


# Supervised Learning

Given



Predict



Input

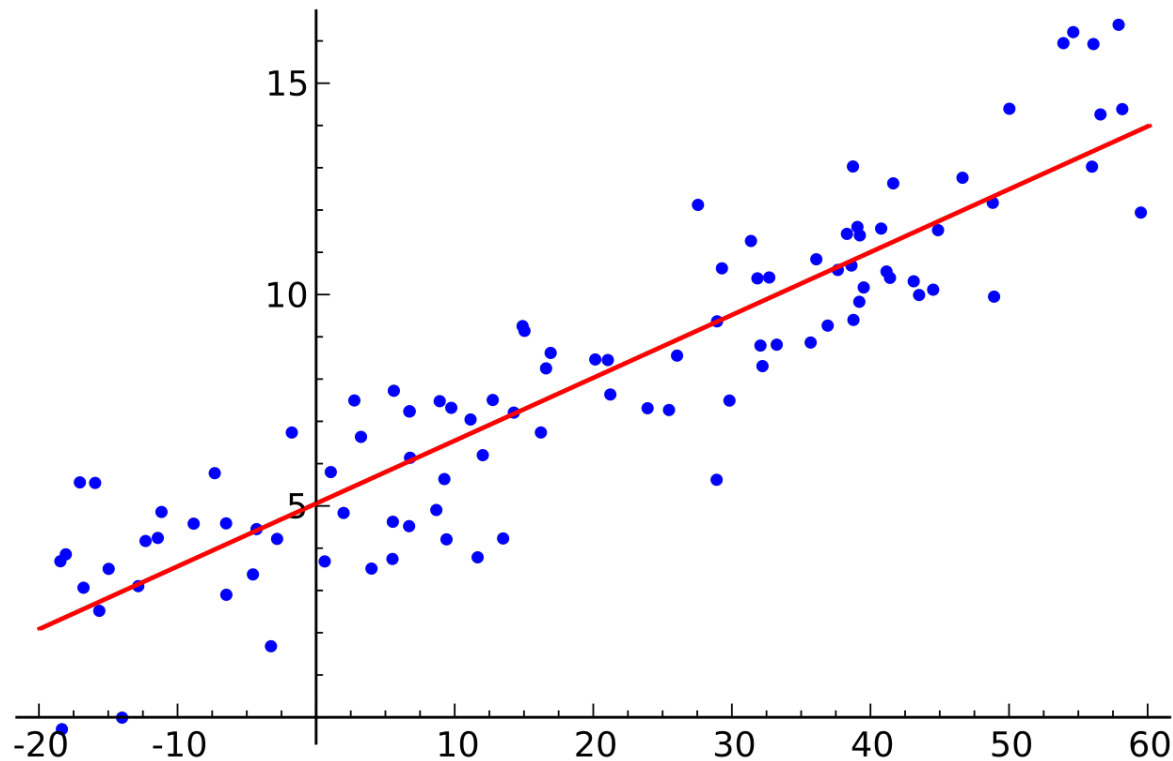
Output

# Supervised Learning

Input (A)	Output (B)	Application
Email	Spam or not	Spam Filter
English	Chinese	Speech translation
Ad, User info	Click/Purchase	Online advertisement
Image of a product	Defect or not	Product inspection

# 2.1 Regression Analysis

- Regression is the supervised learning task for modeling and predicting **continuous, numeric** variables.



# Real Estate Example

Size of House (Sq Ft)	# of BRs	Price (\$1,000)
523	1	115
645	1	150
708	2	210
1034	3	280
2290	4	355
2545	4	440



# Size -> Price



Size of House (Sq Ft)	# of BRs	Price (\$1,000)
523	1	115
645	1	150
708	2	210
1034	3	280
2290	4	355
2545	4	440

# Size, #BR -> Price



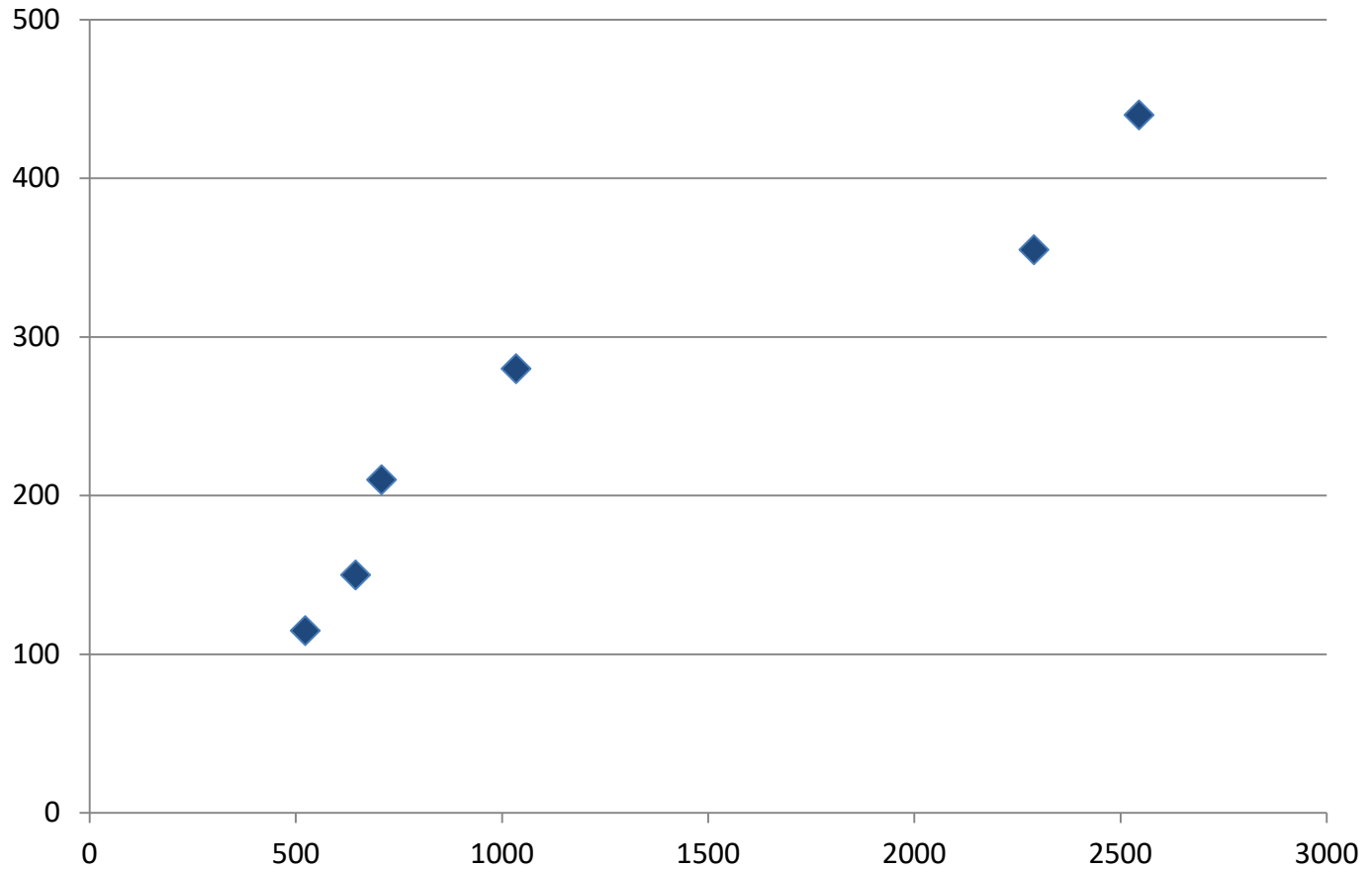
Size of House (Sq Ft)	# of BRs	Price (\$1,000)
523	1	115
645	1	150
708	2	210
1034	3	280
2290	4	355
2545	4	440

# Price -> Sq Ft

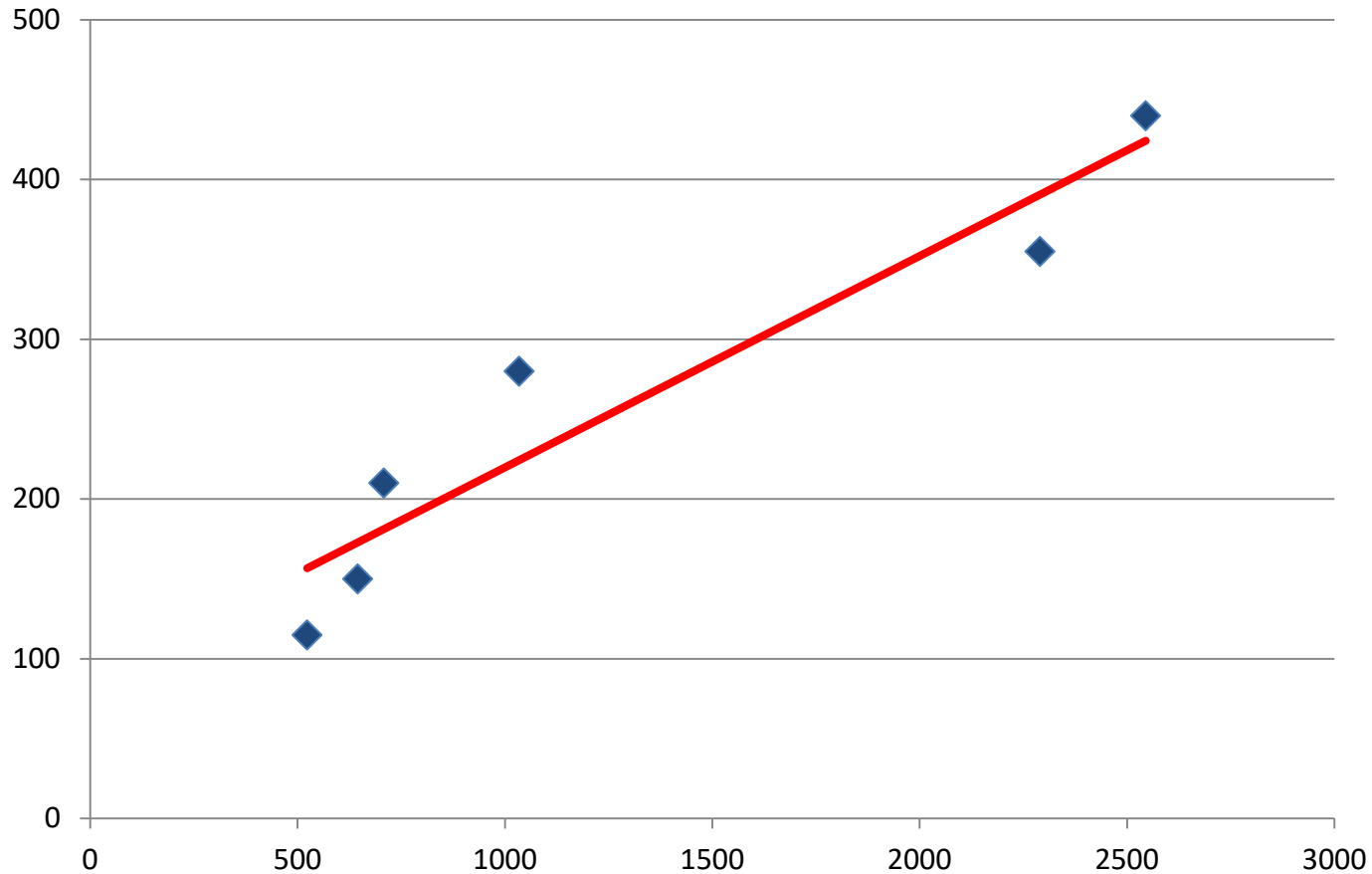


Size of House (Sq Ft)	# of BRs	Price (\$1,000)
523	1	115
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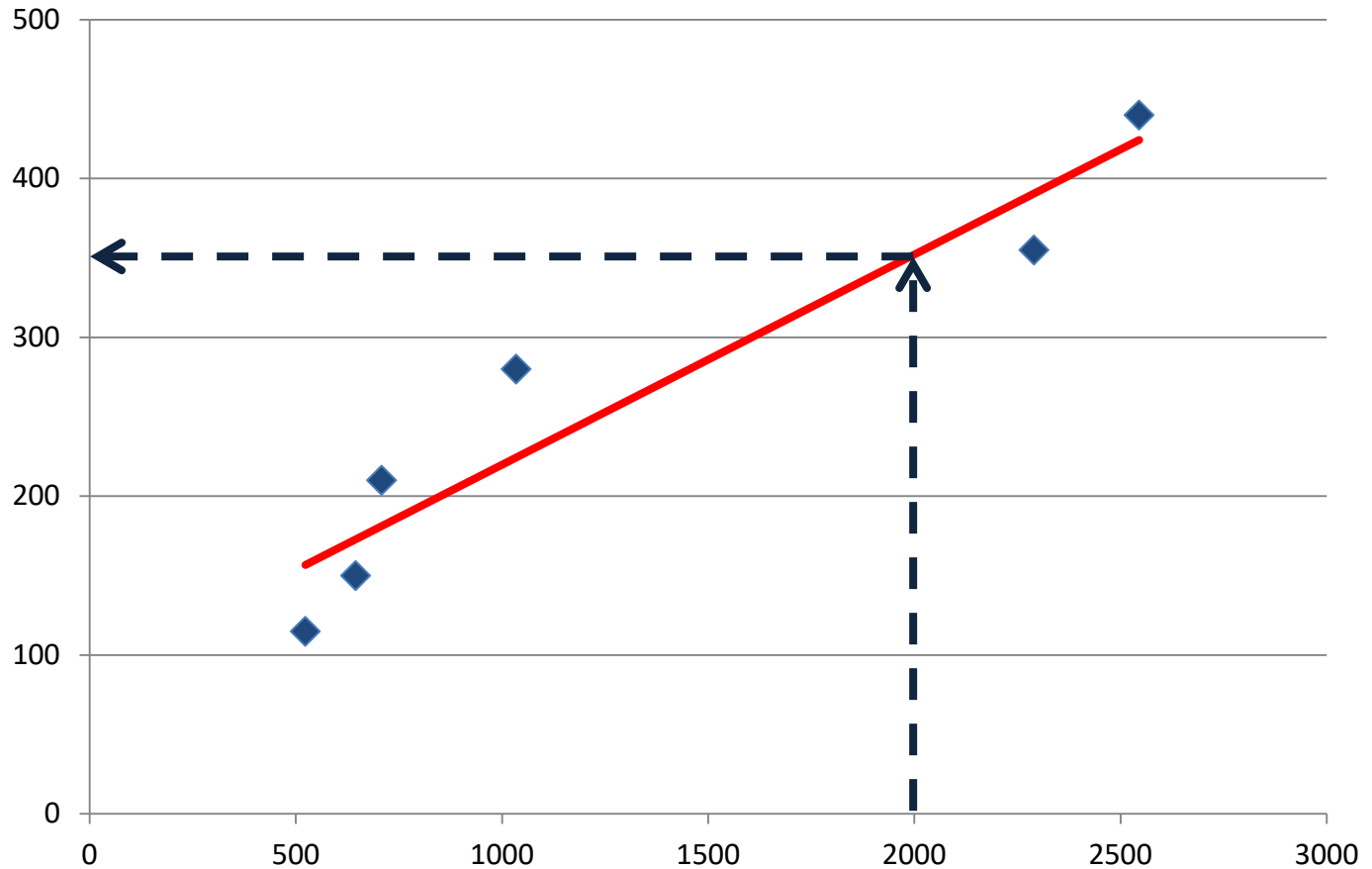
# Regression



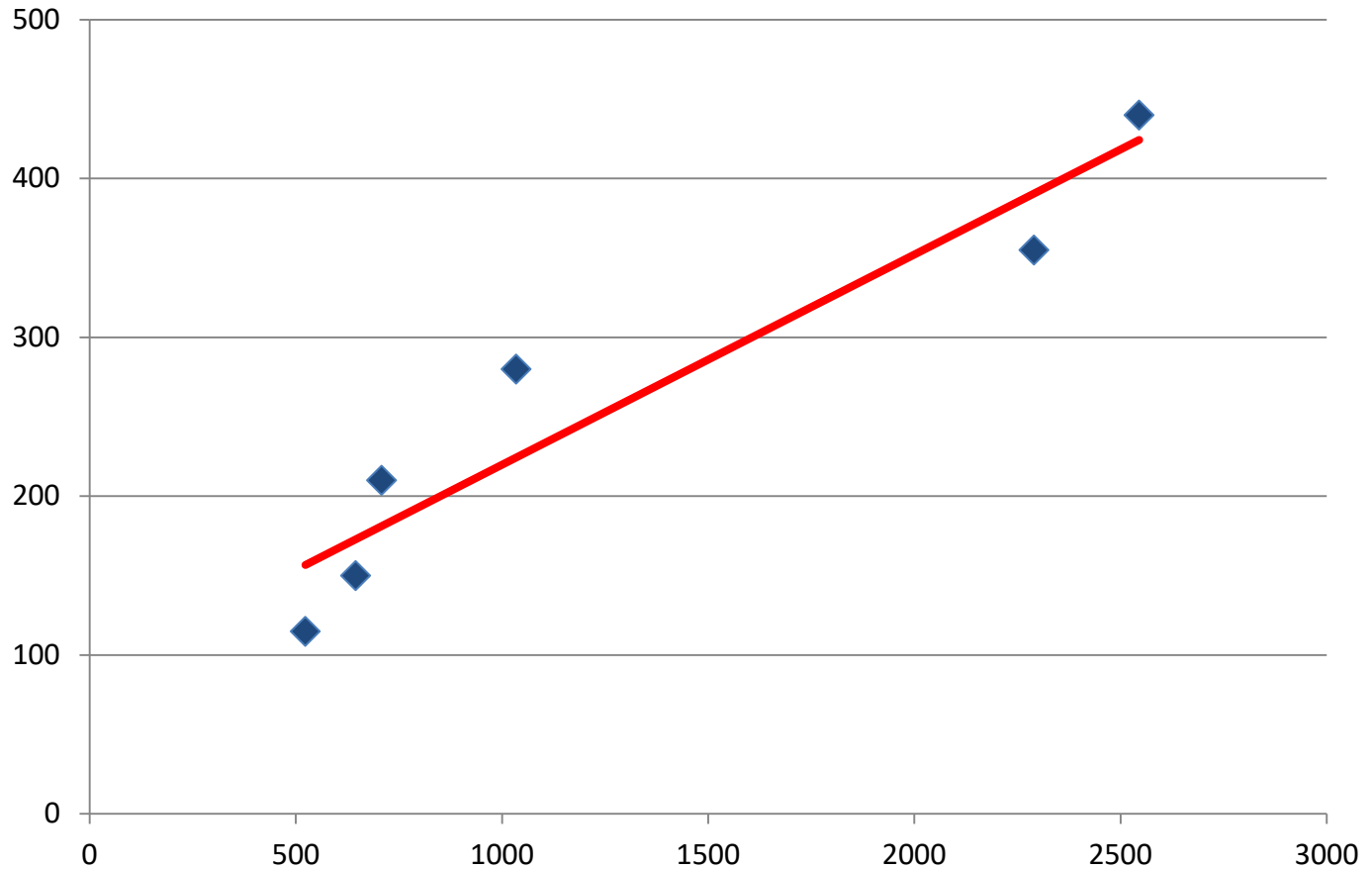
# Linear Regression



# Under-fitting vs Over-fitting



# Under-fitting vs Over-fitting

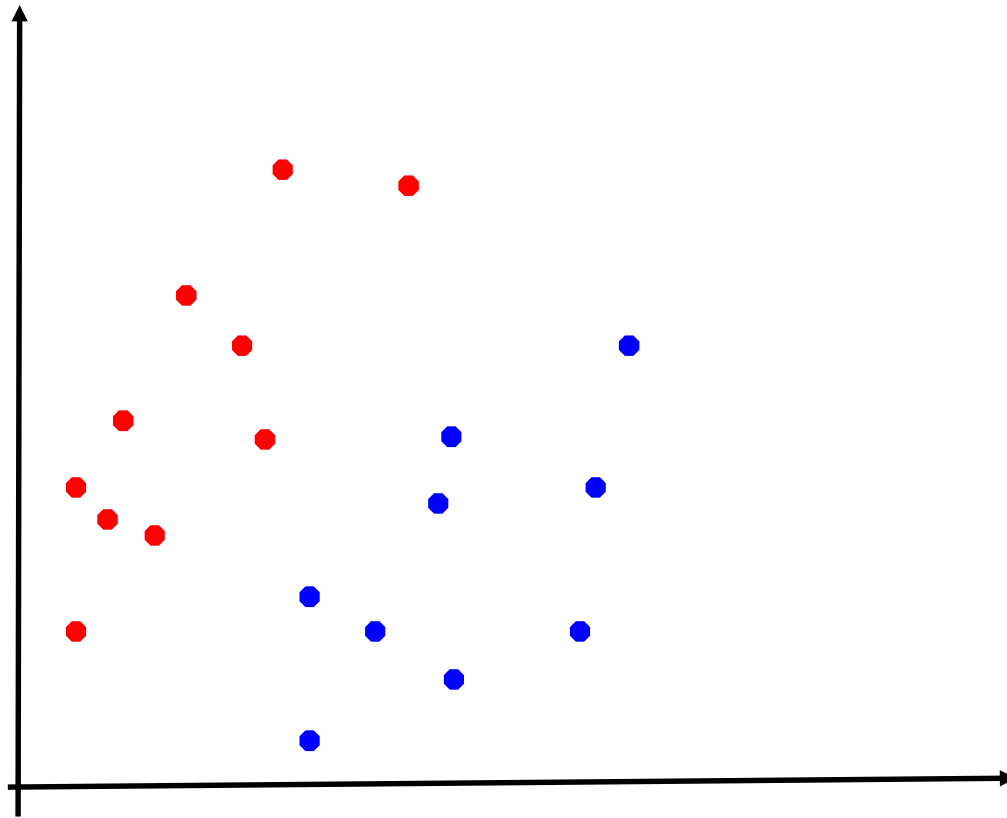


# 2.2 Support Vector Machine

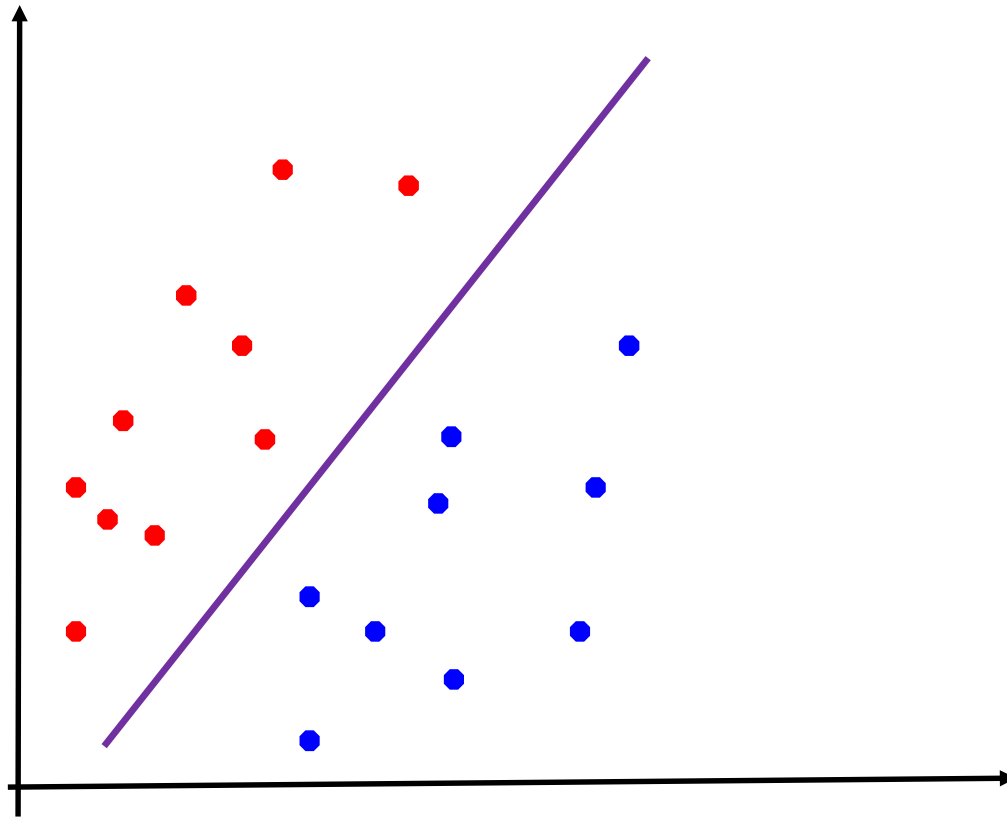
- Classification is the supervised learning task for modeling and predicting **categorical** variables.
- SVM algorithms finds a boundary that maximizes the distance between the closest members of separate classes.



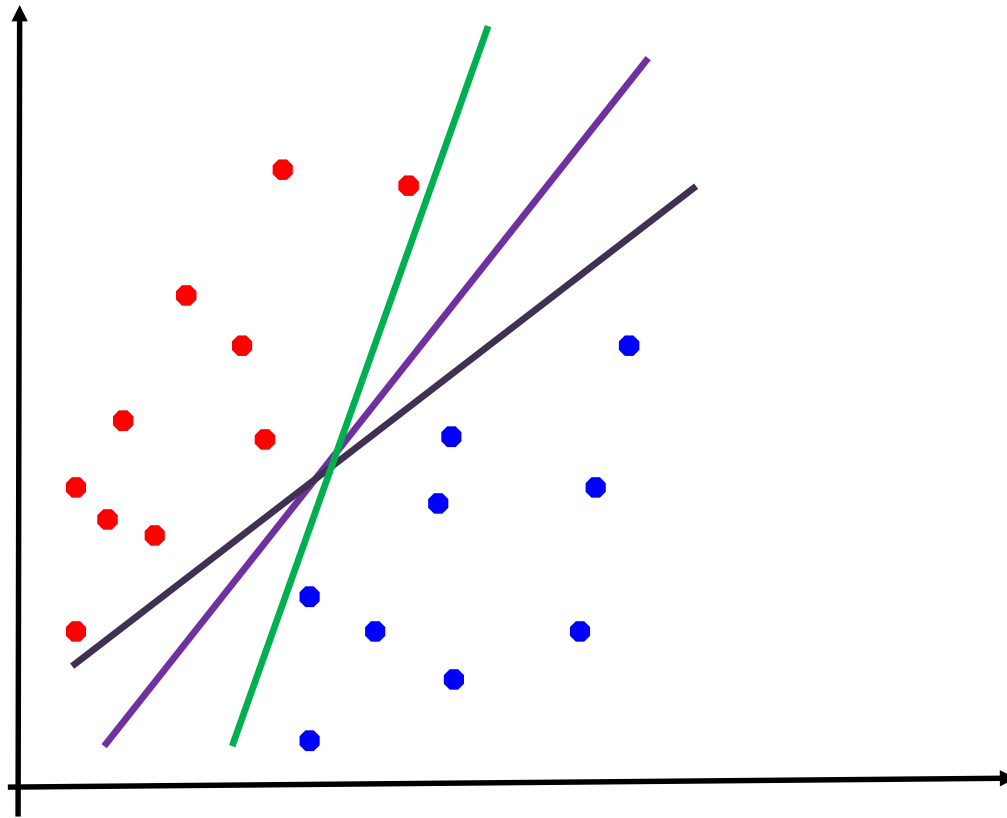
# Two classes of data points



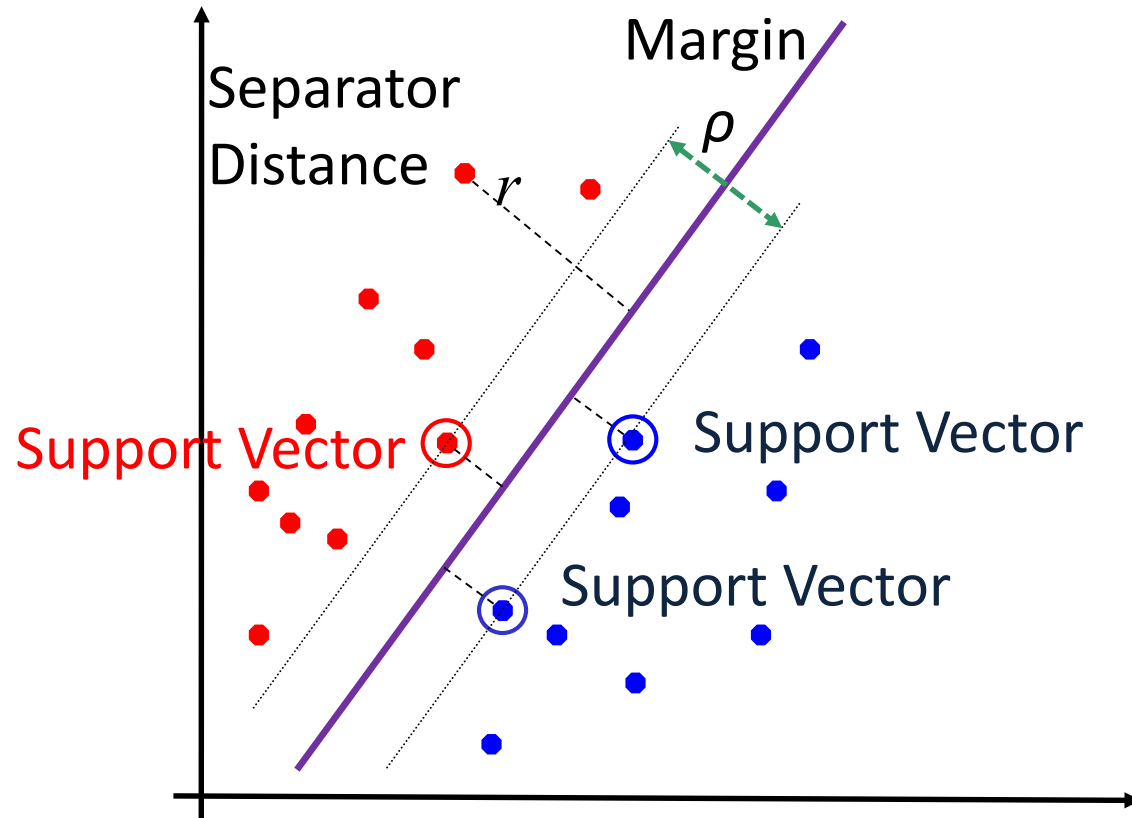
# Linear Separator



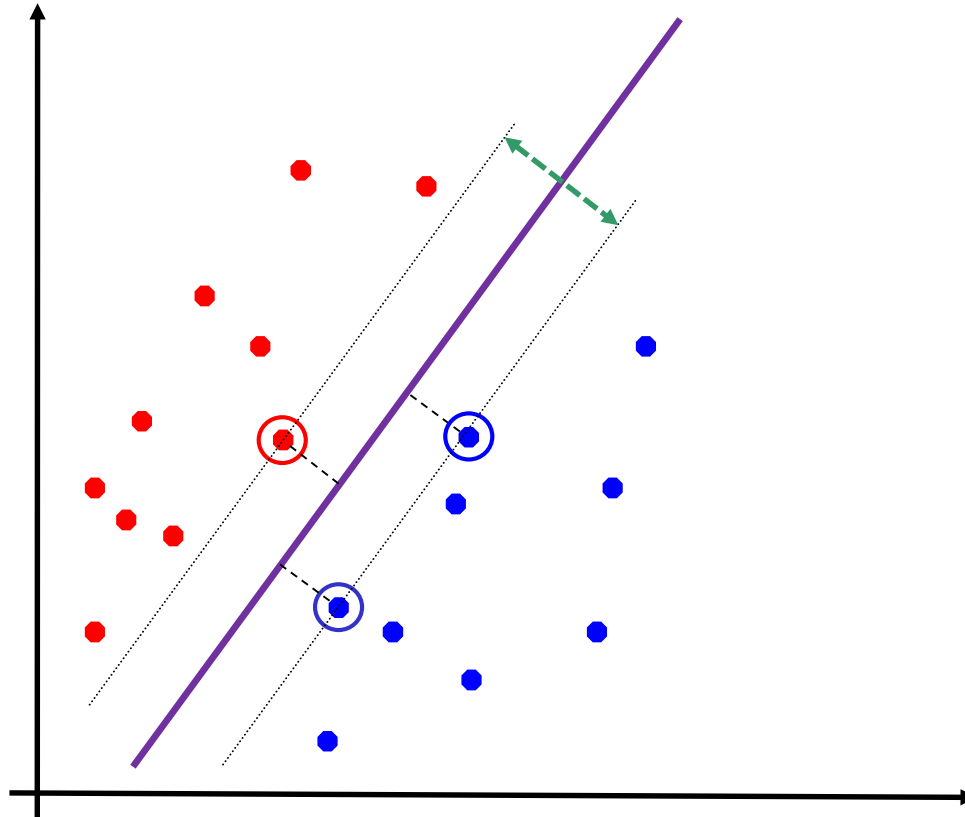
# Linear Separator



# Optimal Separator

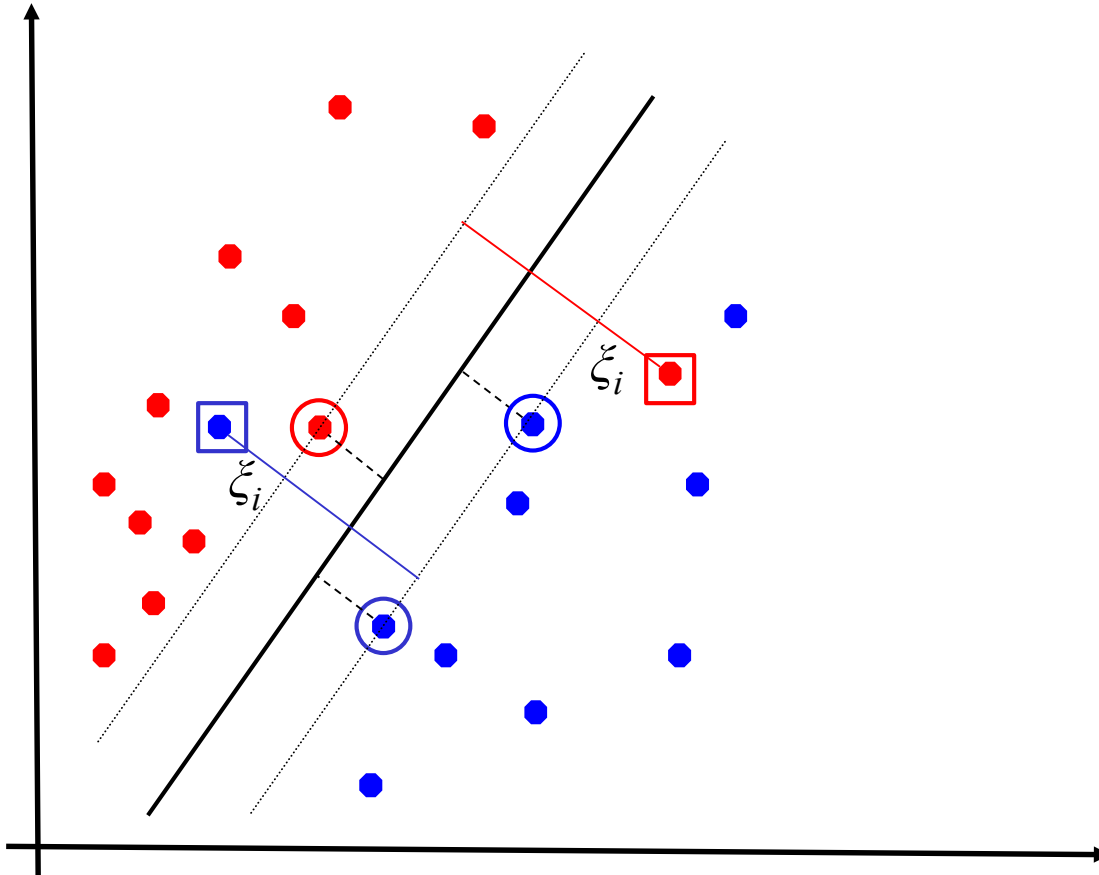


# Optimal Separator



- Maximizing the margin is good intuitively.
- Only support vectors matters

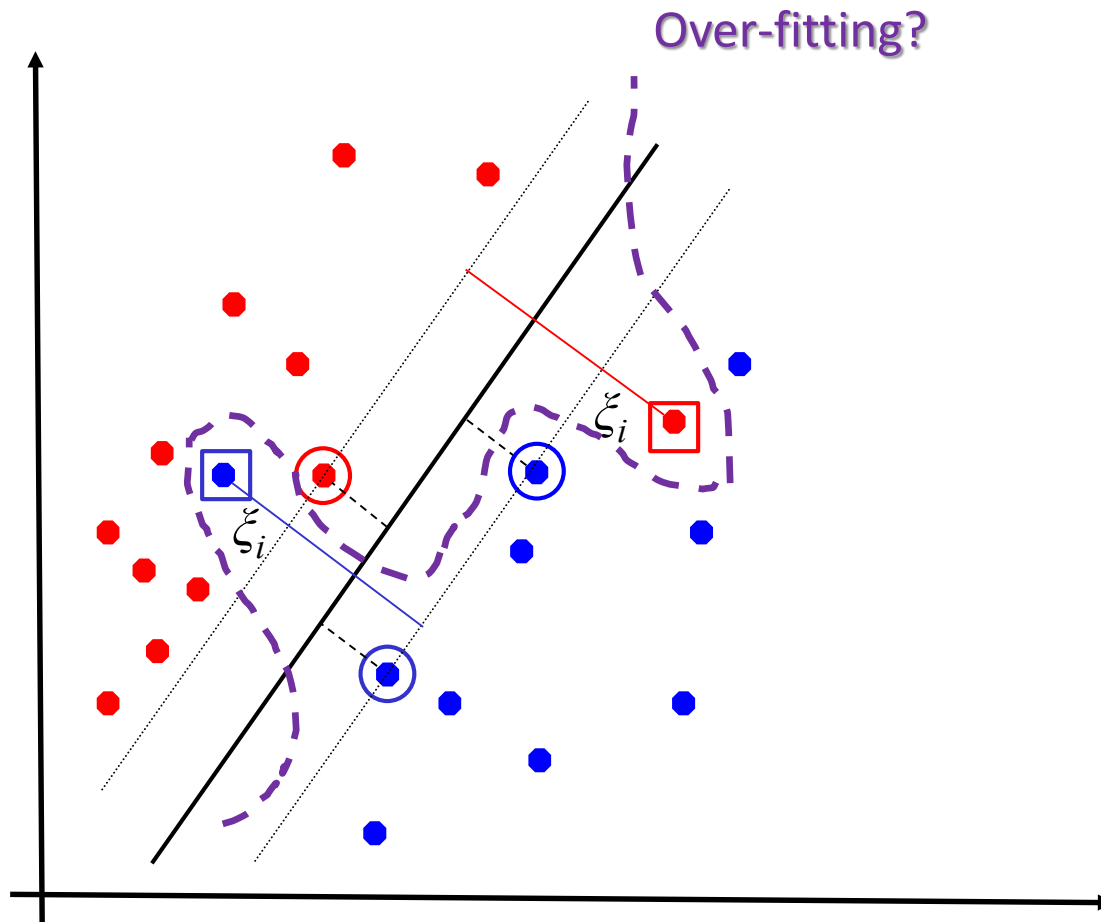
# Soft Margin



Solutions:

- Linear solution with penalty
- Non-linear solutions
- Higher dimensions

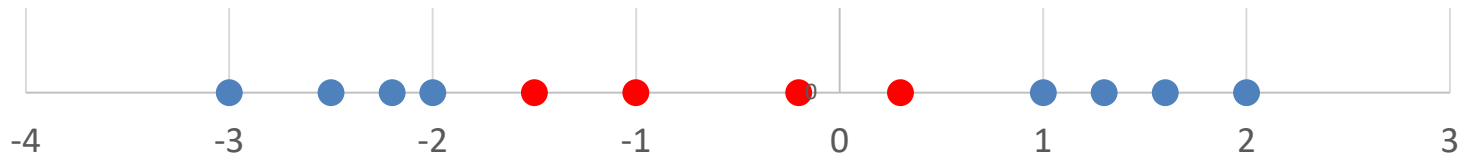
# Soft Margin



Solutions:

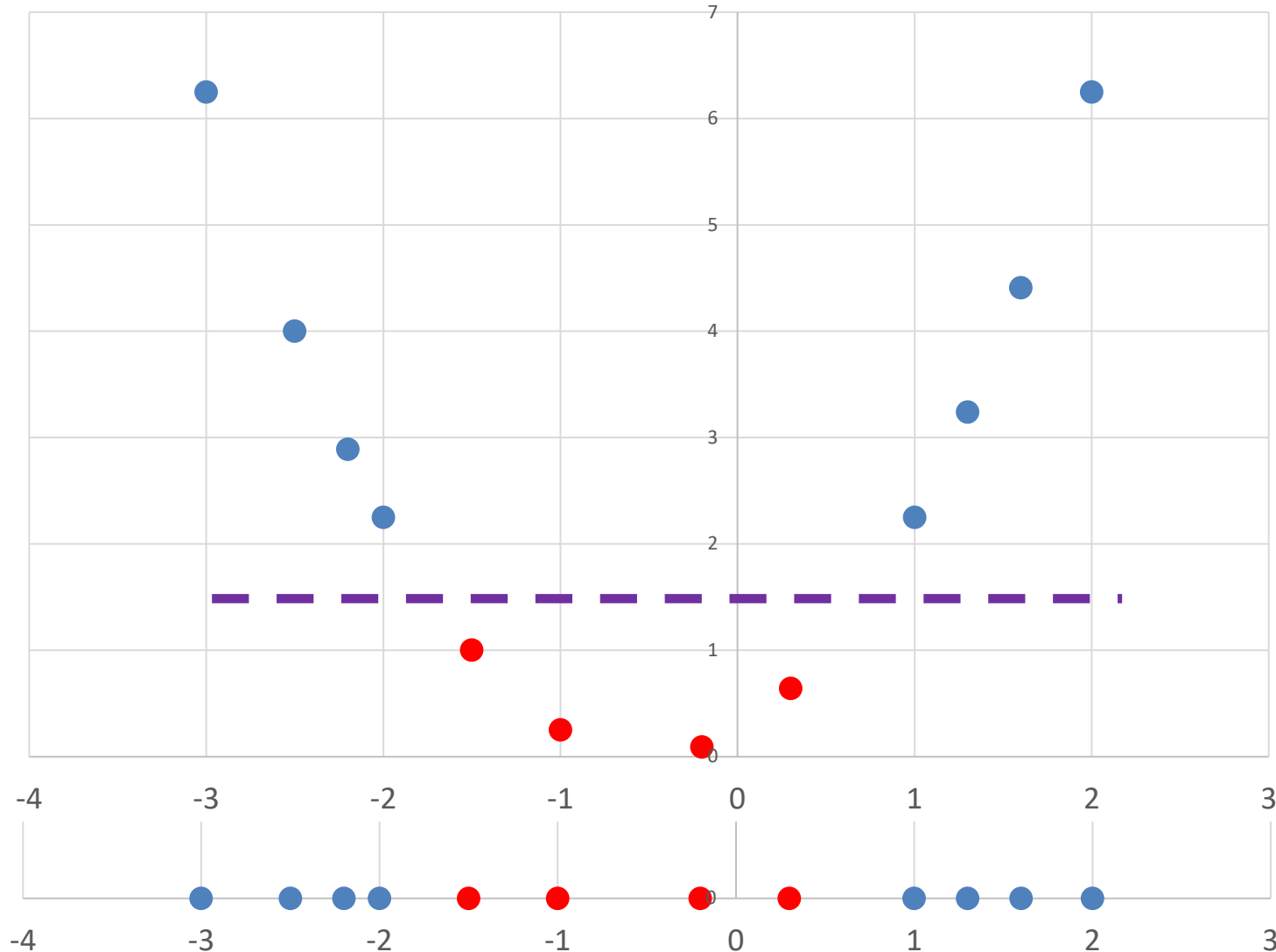
- Linear solution with penalty
- Non-linear solutions
- Higher dimensions

# Feature Space of higher dimension

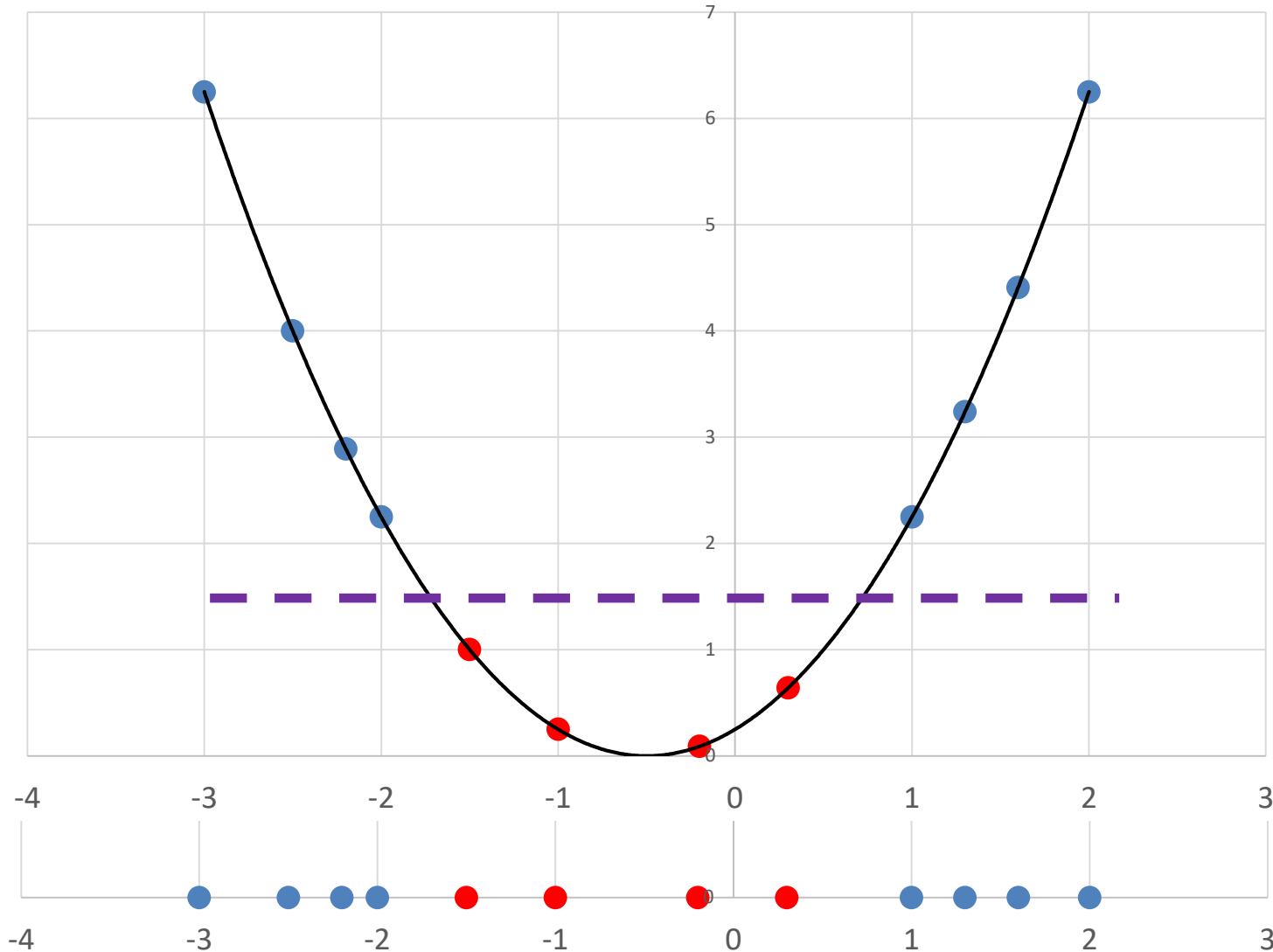




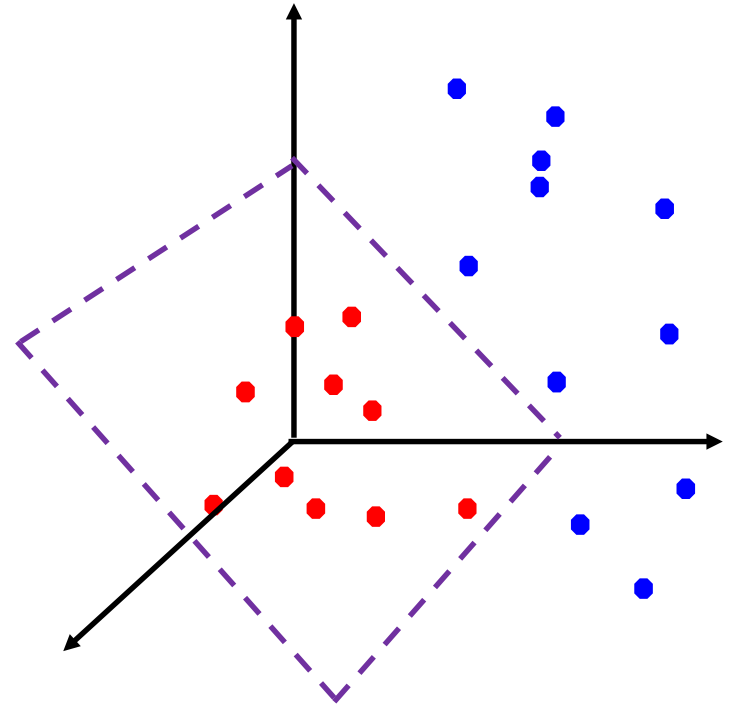
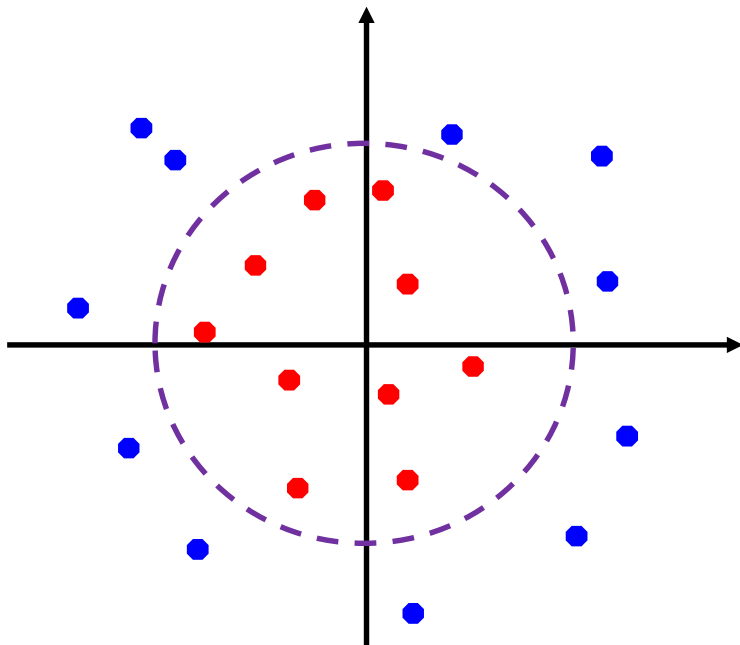
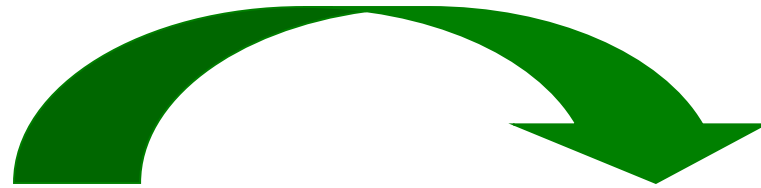
# Feature Space of higher dimension



# Feature Space of higher dimension



# Feature Space of higher dimension



# Example



CUPCAKE

VS



MUFFIN

# Cupcakes




# Muffins



# Collecting Data

Recipe	Flour	Sugar	Butter	Milk	Egg	Baking Powder	Salt	Vanilla
Muffin1	2 cups	1/2 cup	1/4 cup	1 cup	2 eggs	1 tbsp	1/2 tsp	1 tsp
Cupcake1	2 cups	3/4 cup	1/2 cup	1 cup	2 eggs	2 tsp	1/2 tsp	1 tsp
...	...	...	...	...	...	...	...	...

# Cleaning the Data




Recipe	Flour	Sugar	Other
Muffin1	2 cups	1/2 cup	...
Cupcake1	2 cups	3/4 cup	...

Recipe	Flour	Sugar	Other	Total Volume
Muffin1	47%	24%	...	100%
Cupcake1	42%	21%	...	100%



# The Data

Label



Type	Flour	Milk	Sugar	Butter	Egg	Baking Powder	Vanilla	Salt
Muffin	55	28	3	7	5	2	0	0
Muffin	47	24	12	6	9	1	0	0
Muffin	47	23	18	6	4	1	0	0
Muffin	50	25	12	6	5	2	1	0
Muffin	55	27	3	7	5	2	1	0
Muffin	54	27	7	5	5	2	0	0
Muffin	47	26	10	10	4	1	0	0
Muffin	50	17	17	8	6	1	0	0
Muffin	50	17	17	11	4	1	0	0
Cupcake	39	0	26	19	14	1	1	0
Cupcake	34	17	20	20	5	2	1	0
Cupcake	39	13	17	19	10	1	1	0
Cupcake	38	15	23	15	8	0	1	0
Cupcake	42	18	25	9	5	1	0	0
Cupcake	36	14	21	14	11	2	1	0
Cupcake	38	15	31	8	6	1	1	0
Cupcake	36	16	24	12	9	1	1	0
Cupcake	34	17	23	11	13	0	1	0

# The 8 Dimensional Data

Label

Not  
Useful

Type	Flour	Milk	Sugar	Butter	Egg	Baking Powder	Vanilla	Salt
Muffin	55	28	3	7	5	2	0	0
Muffin	47	24	12	6	9	1	0	0
Muffin	47	23	18	6	4	1	0	0
Muffin	50	25	12	6	5	2	1	0
Muffin	55	27	3	7	5	2	1	0
Muffin	54	27	7	5	5	2	0	0
Muffin	47	26	10	10	4	1	0	0
Muffin	50	17	17	8	6	1	0	0
Muffin	50	17	17	11	4	1	0	0
Cupcake	39	0	26	19	14	1	1	0
Cupcake	34	17	20	20	5	2	1	0
Cupcake	39	13	17	19	10	1	1	0
Cupcake	38	15	23	15	8	0	1	0
Cupcake	42	18	25	9	5	1	0	0
Cupcake	36	14	21	14	11	2	1	0
Cupcake	38	15	31	8	6	1	1	0
Cupcake	36	16	24	12	9	1	1	0
Cupcake	34	17	23	11	13	0	1	0

# The 8 Dimensional Data

Label

Maybe

Type	Flour	Milk	Sugar	Butter	Egg	Baking Powder	Vanilla	Salt
Muffin	55	28	3	7	5	2	0	0
Muffin	47	24	12	6	9	1	0	0
Muffin	47	23	18	6	4	1	0	0
Muffin	50	25	12	6	5	2	1	0
Muffin	55	27	3	7	5	2	1	0
Muffin	54	27	7	5	5	2	0	0
Muffin	47	26	10	10	4	1	0	0
Muffin	50	17	17	8	6	1	0	0
Muffin	50	17	17	11	4	1	0	0
Cupcake	39	0	26	19	14	1	1	0
Cupcake	34	17	20	20	5	2	1	0
Cupcake	39	13	17	19	10	1	1	0
Cupcake	38	15	23	15	8	0	1	0
Cupcake	42	18	25	9	5	1	0	0
Cupcake	36	14	21	14	11	2	1	0
Cupcake	38	15	31	8	6	1	1	0
Cupcake	36	16	24	12	9	1	1	0
Cupcake	34	17	23	11	13	0	1	0

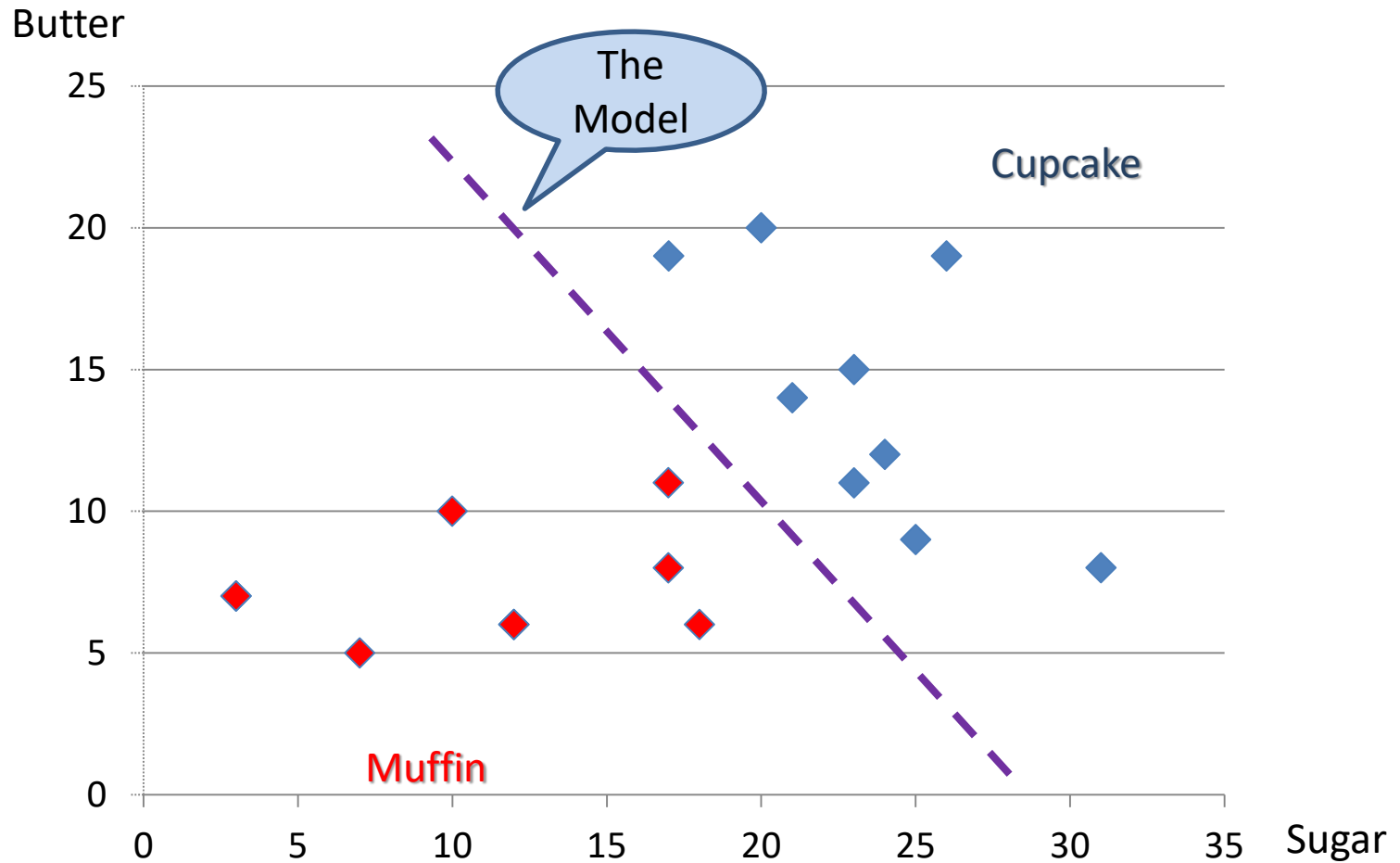
# The 8 Dimensional Data

Label

Our Selection

Type	Flour	Milk	Sugar	Butter	Egg	Baking Powder	Vanilla	Salt
Muffin	55	28	3	7	5	2	0	0
Muffin	47	24	12	6	9	1	0	0
Muffin	47	23	18	6	4	1	0	0
Muffin	50	25	12	6	5	2	1	0
Muffin	55	27	3	7	5	2	1	0
Muffin	54	27	7	5	5	2	0	0
Muffin	47	26	10	10	4	1	0	0
Muffin	50	17	17	8	6	1	0	0
Muffin	50	17	17	11	4	1	0	0
Cupcake	39	0	26	19	14	1	1	0
Cupcake	34	17	20	20	5	2	1	0
Cupcake	39	13	17	19	10	1	1	0
Cupcake	38	15	23	15	8	0	1	0
Cupcake	42	18	25	9	5	1	0	0
Cupcake	36	14	21	14	11	2	1	0
Cupcake	38	15	31	8	6	1	1	0
Cupcake	36	16	24	12	9	1	1	0
Cupcake	34	17	23	11	13	0	1	0

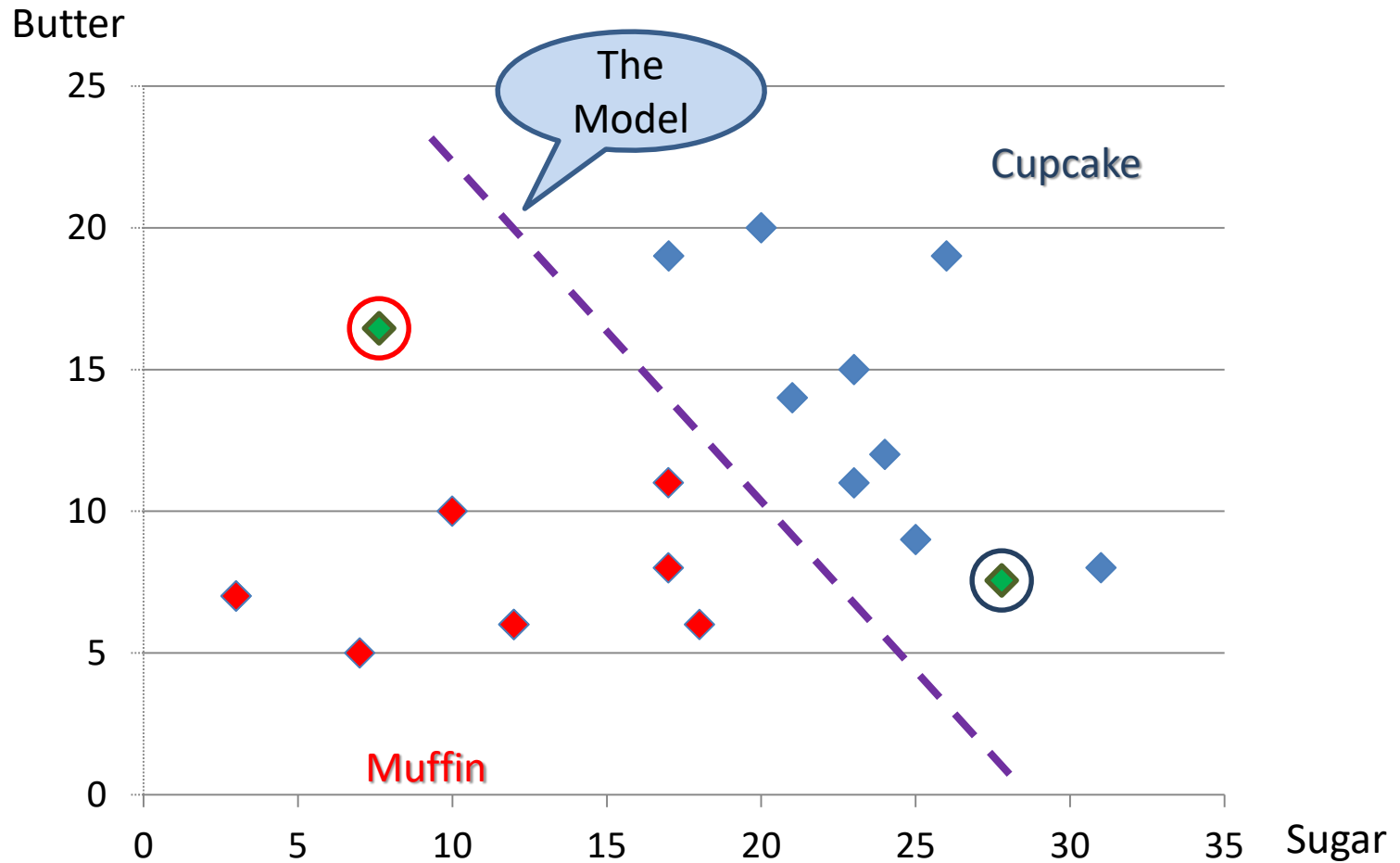
# SVM Training



# Cupcake or Muffin?



# SVM Testing



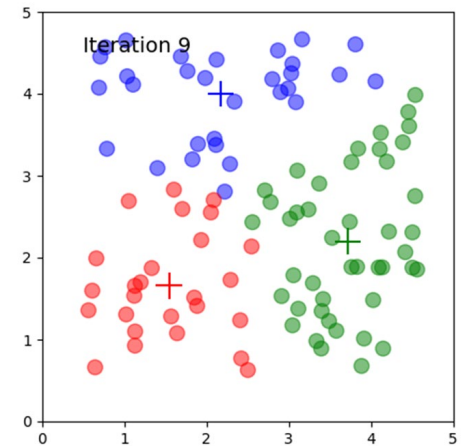
# 3. Unsupervised Learning

- Clustering groups a set of objects in such a way that objects in the same group are more **similar** to each other than to those in other groups.
- $K$ -means clustering aims to partition  $n$  observations into  $k$  clusters in which each observation belongs to the cluster with the nearest mean.



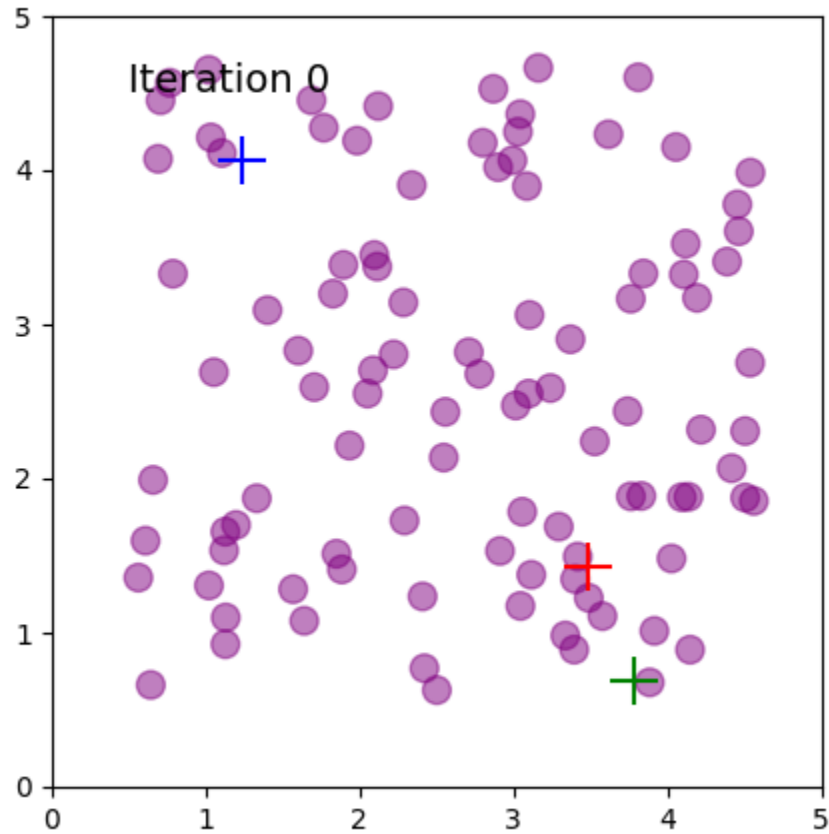
# K-means algorithm

- 1) Pick  $k$  cluster centers randomly.
- 2) Assign every data point to its nearest cluster center.
- 3) Move each cluster center to the mean of its assigned data points.
- 4) Repeat Steps 2-3 until it converges.

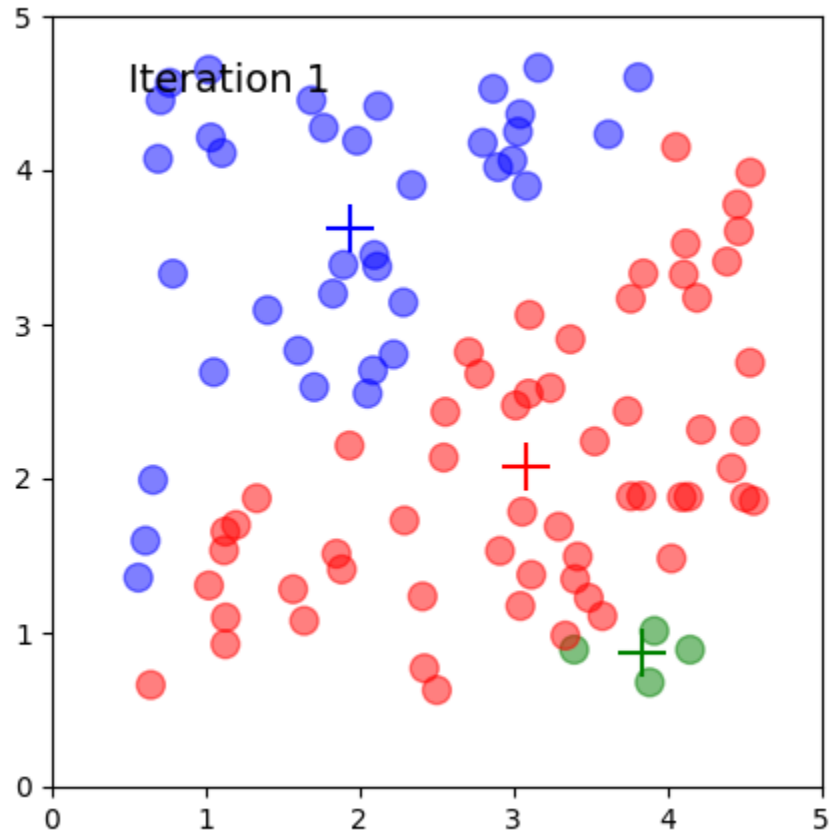


Slides from Wash Univ. BIO5488 lecture, 2004

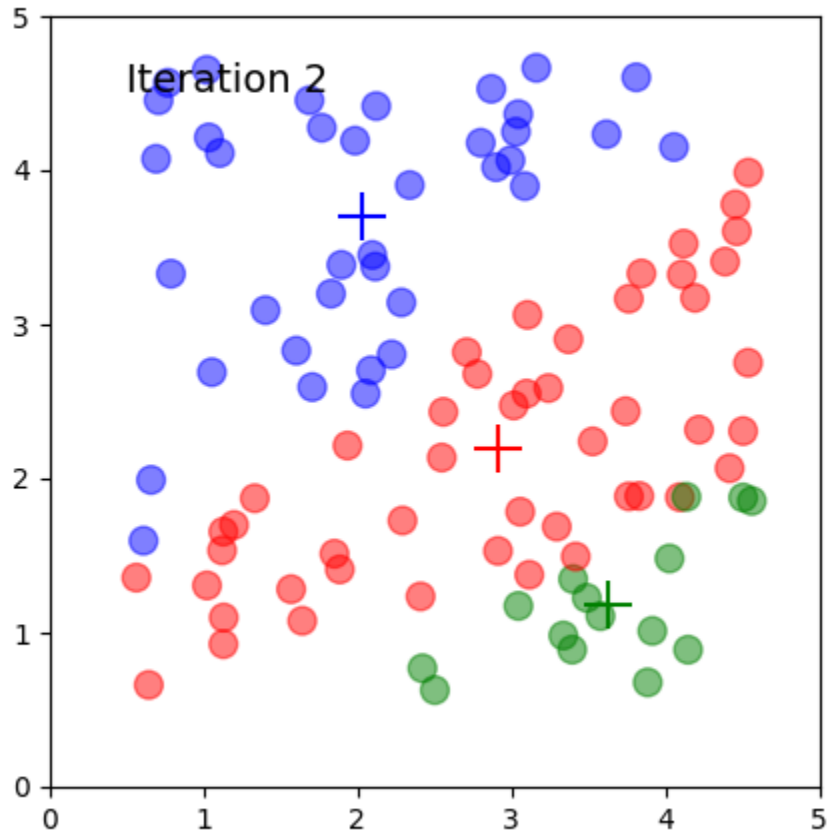
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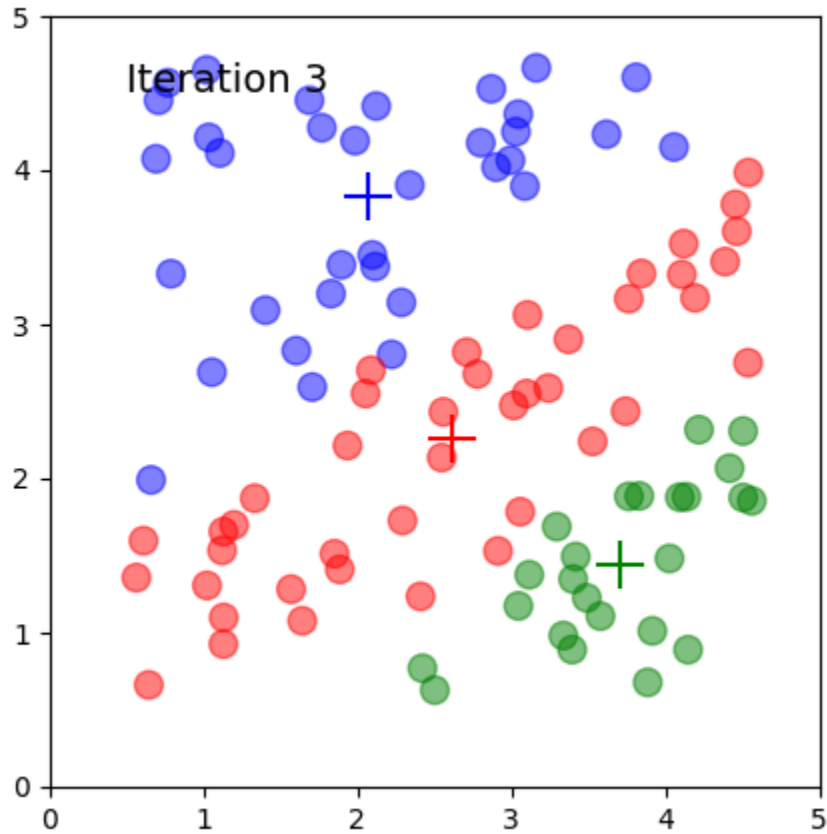
# 1



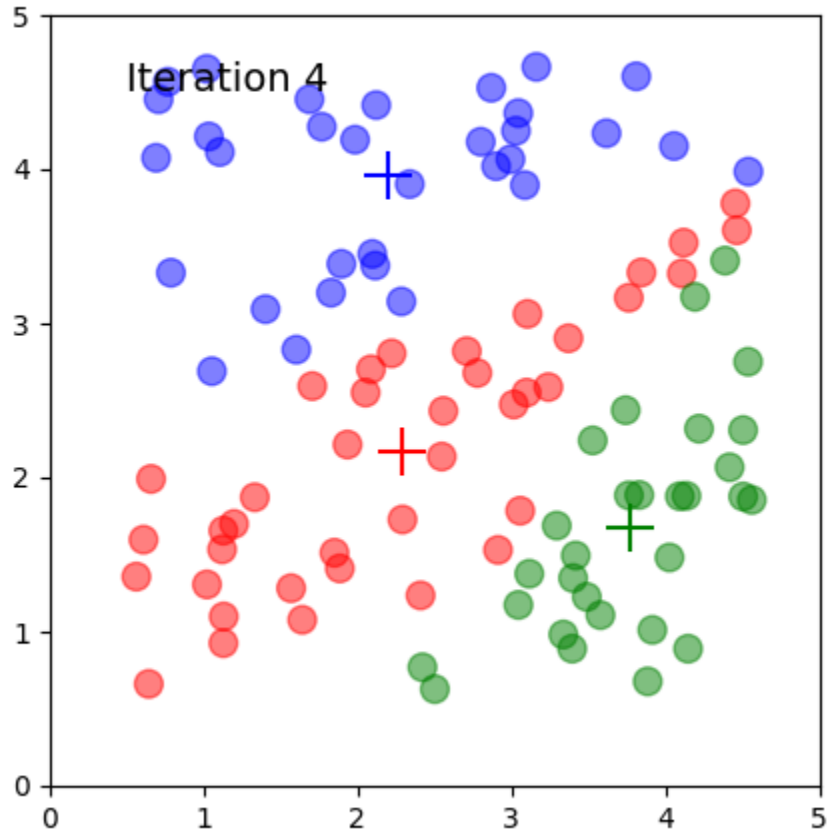
# 2



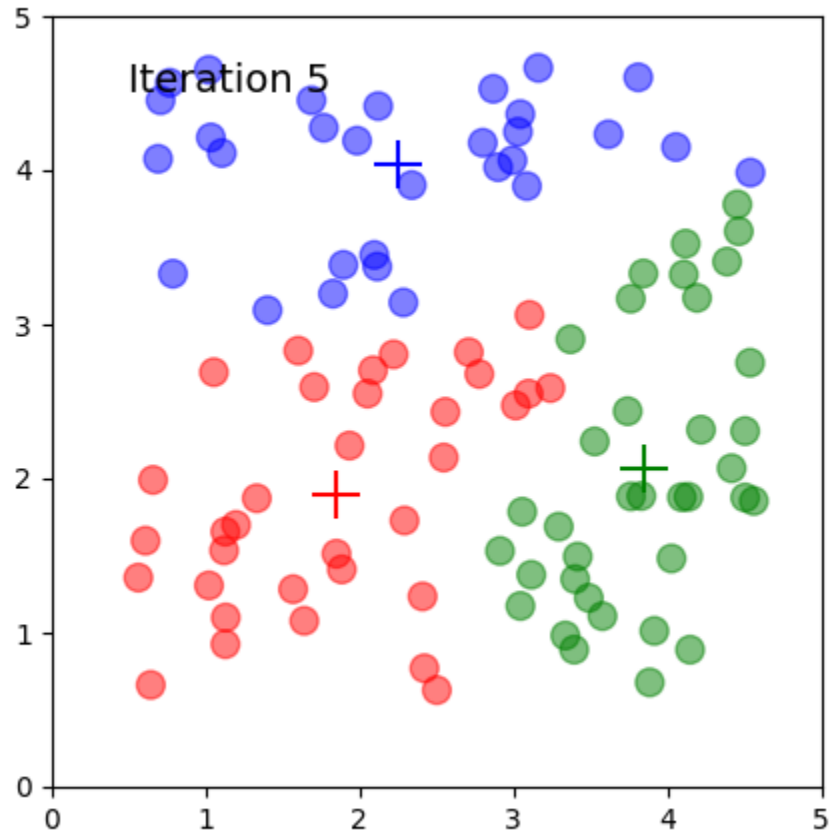
# 3



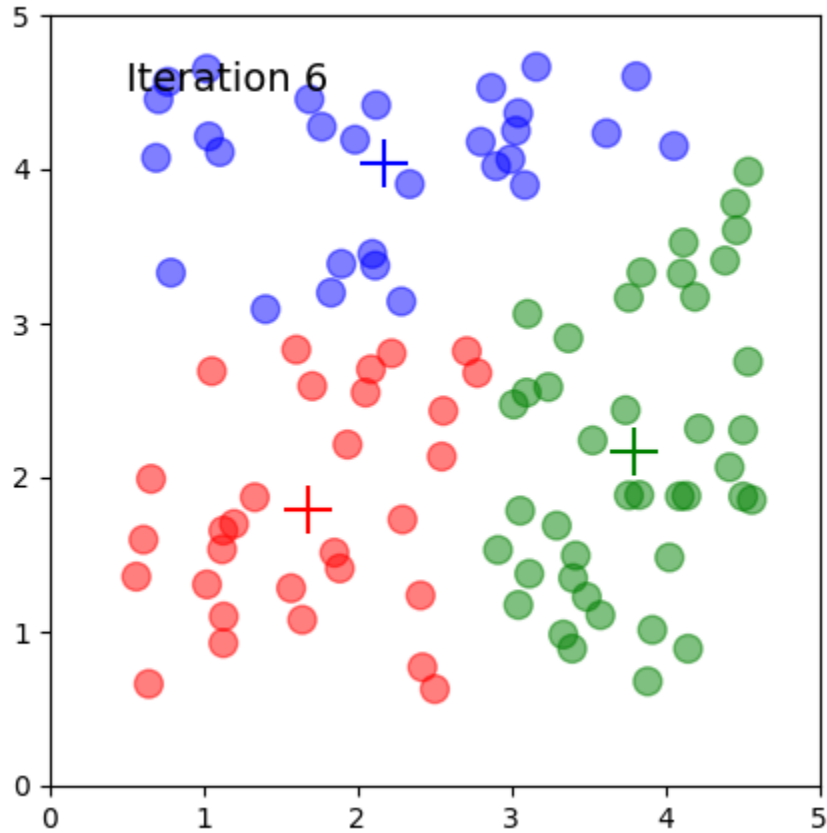
# 4



# 5

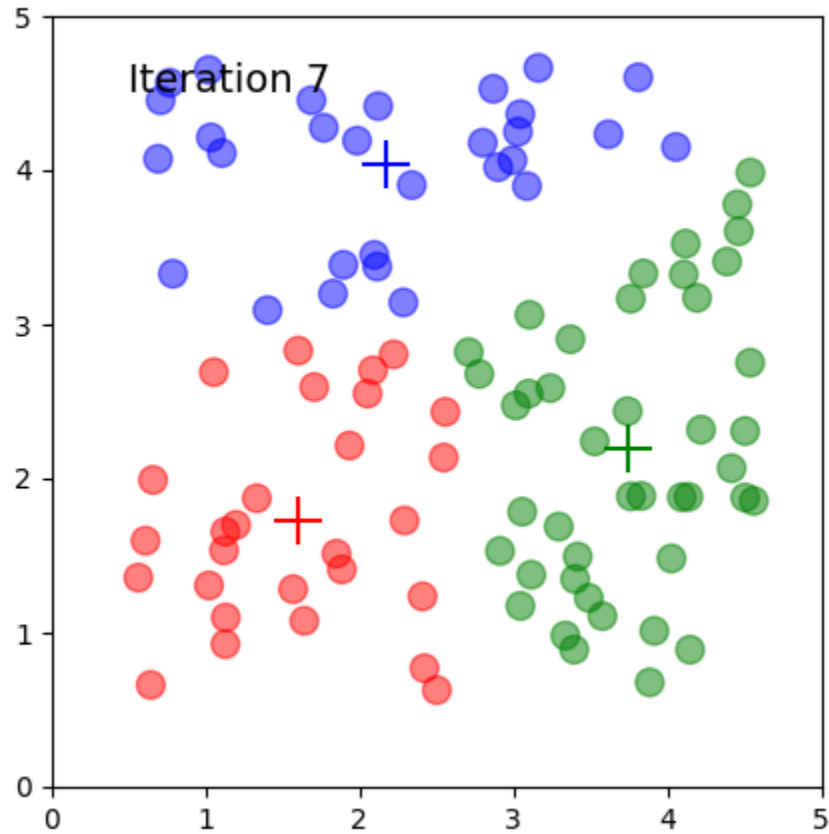


# 6

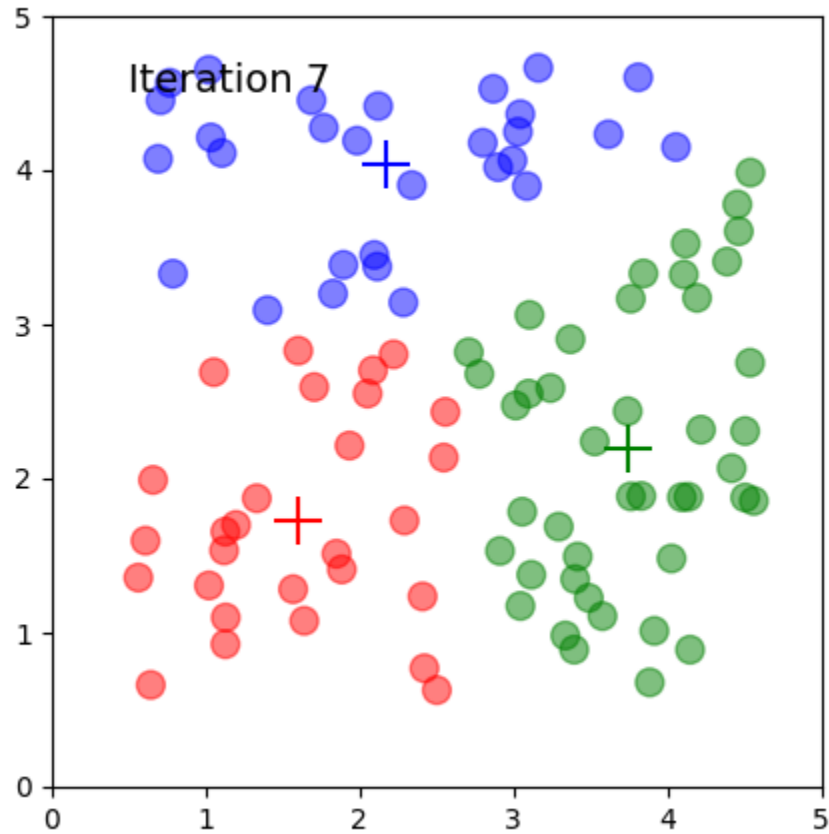




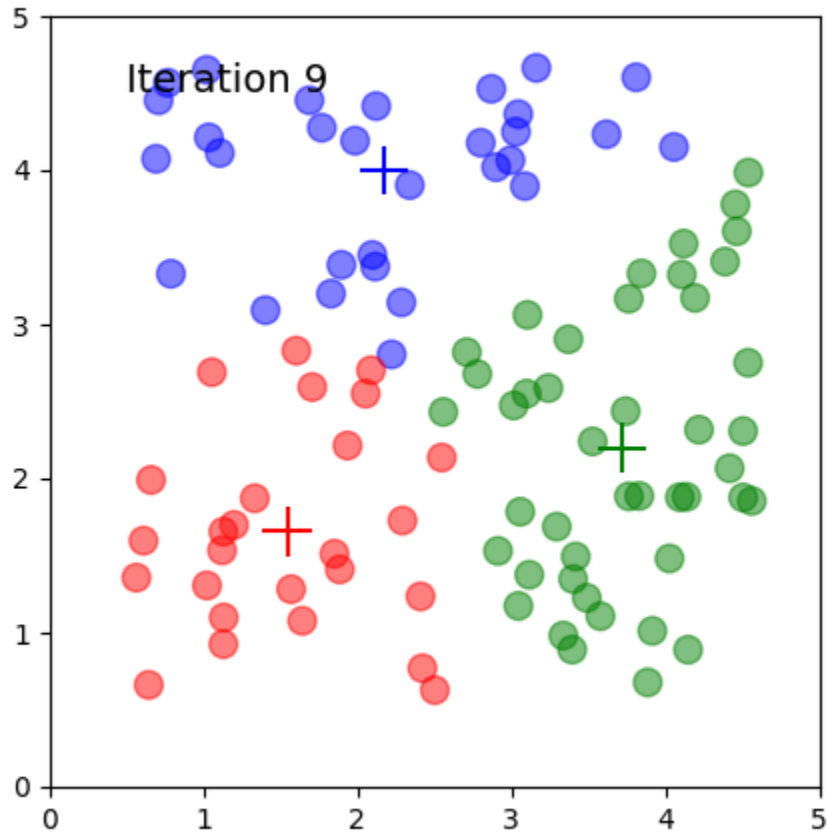
# 7



# 8



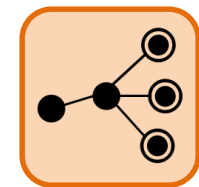
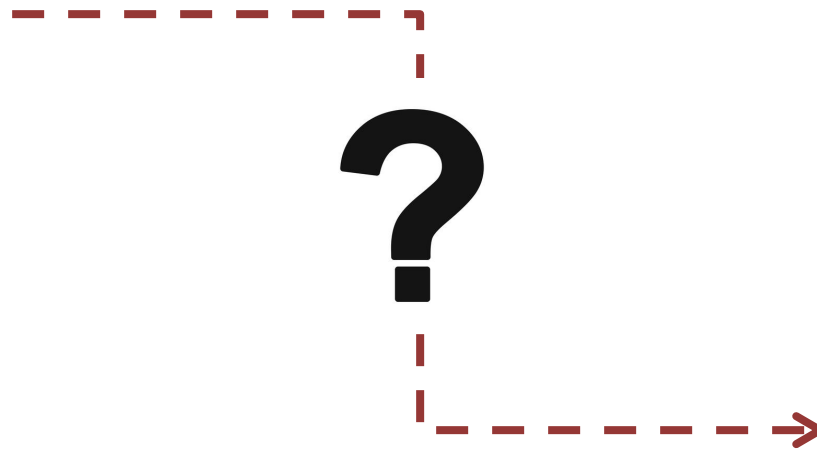
# 9



# 4. Machine Learning Process



Data



Prediction

# Machine Learning Process



Data

# Machine Learning Process



Labels



Data

# Machine Learning Process



Labels

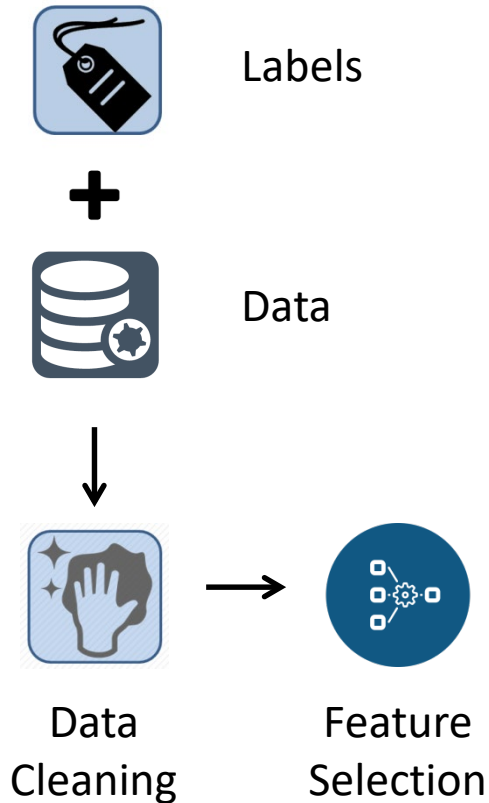


Data



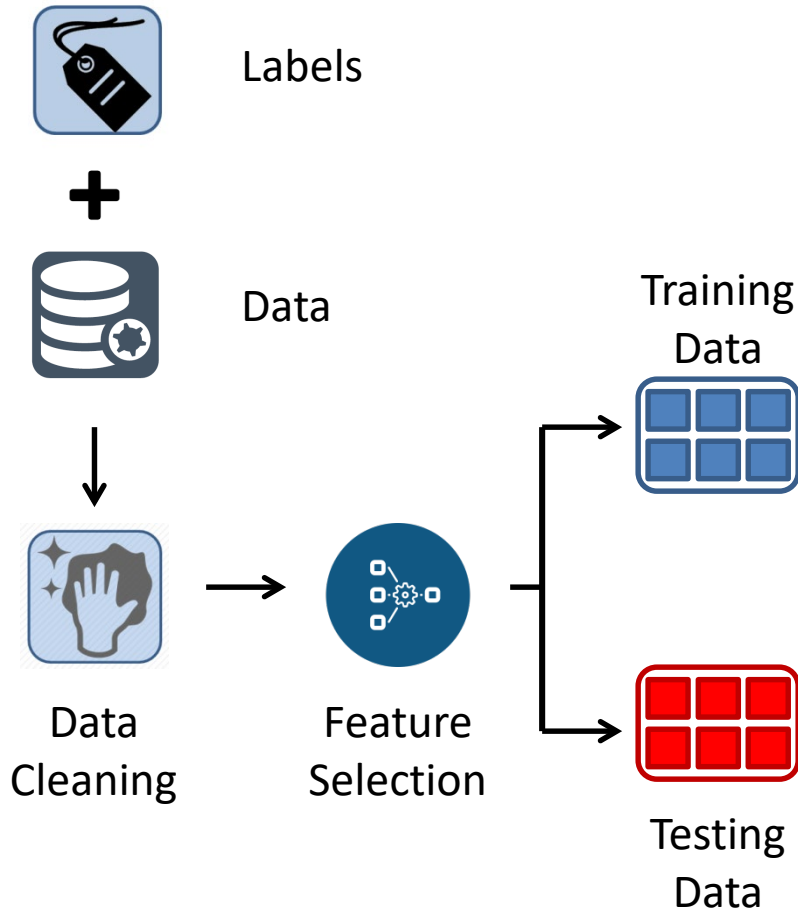
Data  
Cleaning

# Machine Learning Process

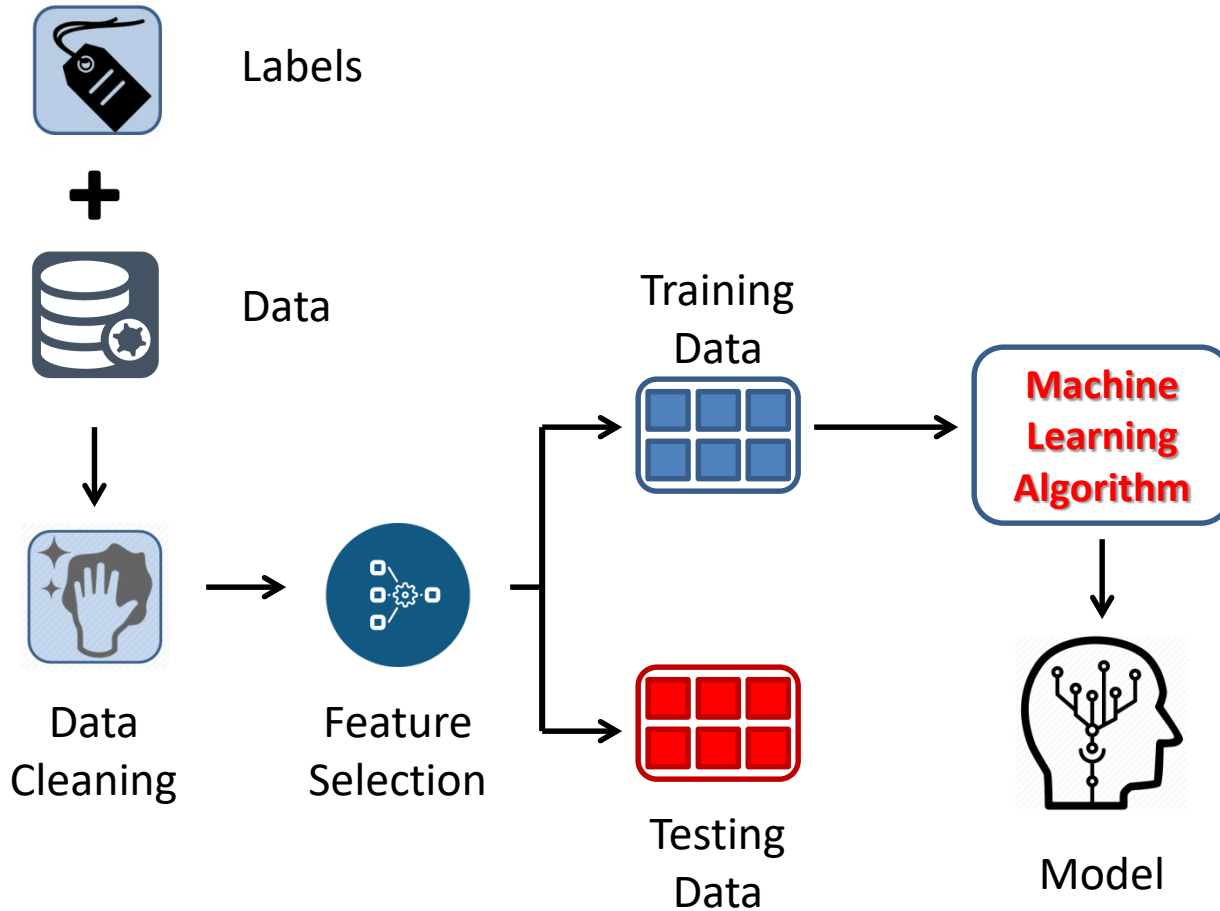




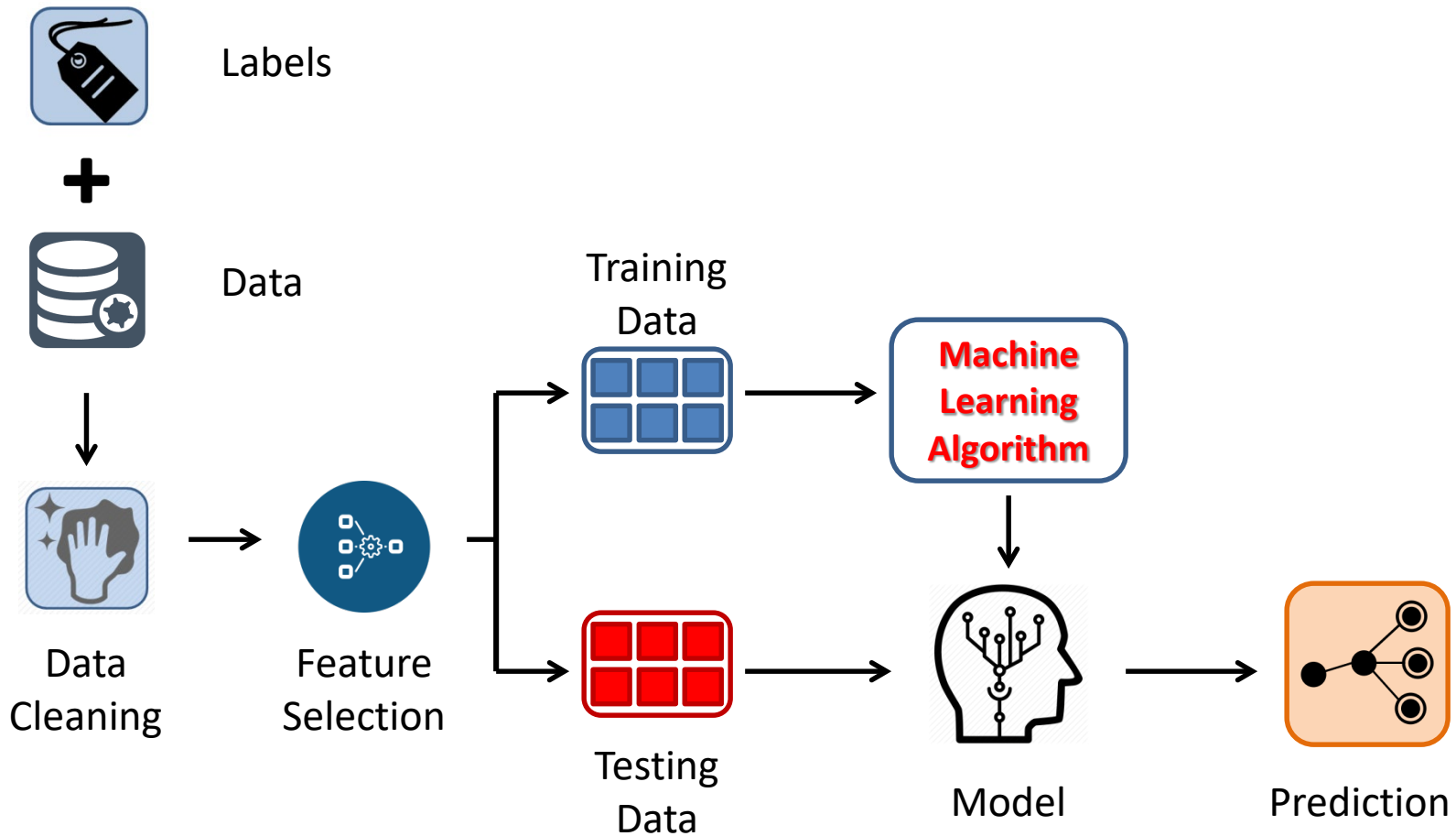
# Machine Learning Process



# Machine Learning Process



# Machine Learning Process



# 5. Summary

- ML tends to work well when:
  - Learning a “simple” concept
  - There is lots of data available
  - There are attributes that distinguish the classes.
- ML tends to work poorly when:
  - Learning complex concepts from small amount of data
  - It is asked to perform on new types of data
- ML Sometimes requires tuning (parameters)

# Credits

- AI for Everyone, Andrew Ng,  
<https://www.coursera.org/learn/ai-for-everyone>,
- Support Vector Machines: A Visual Explanation with Sample Python Code, Alice Zhao,  
<https://www.youtube.com/watch?v=N1vOgolbjSc>
- K-means illustrator, Wesam Elshamy,  
<https://github.com/welshamy/k-means-illustrator>