



K Nearest Neighbor

Cesar Acosta Ph.D.

Department of Industrial and Systems Engineering
University of Southern California



K-Nearest Neighbors

*K-Nearest neighbors for classification
predicts the category using
the k -closest observations*



K-Nearest Neighbors – Example (K=1)

Y	X
0	13
1	32
0	19
0	14
1	28
1	22



K-Nearest Neighbors – Example (K=1)

Y	X		distance
0	13		
1	32		19
0	19		6
0	14		1
1	28		15
1	22		9



K-Nearest Neighbors – Example (K=1)

Y	X	Yhat	distance	
0	13			
1	32		19	
0	19		6	
0	14	0	1	min
1	28		15	
1	22		9	



K-Nearest Neighbors – Example (K=1)

Y	X	distance
0	13	19
1	32	
0	19	13
0	14	18
1	28	4
1	22	10



K-Nearest Neighbors – Example (K=1)

Y	X	Yhat	distance	
0	13		19	
1	32			
0	19		13	
0	14	0	18	
1	28	1	4	min
1	22		10	



K-Nearest Neighbors – Example (K=1)

Y	X1	X2
0	13	29
1	32	8
0	19	21
0	12	33
1	28	14
1	22	12



K-Nearest Neighbors – Example (K=1)

Y	X1	X2		distance
0	13	29		
1	32	8		28.32
0	19	21		10.00
0	12	33		4.12
1	28	14		21.21
1	22	12		19.24



K-Nearest Neighbors – Example (K=1)

Y	X1	X2	distance	
0	13	29		
1	32	8	28.32	(13,29) to (32,8)
0	19	21	10.00	(13,29) to (19,21)
0	12	33	4.12	(13,29) to (12,33)
1	28	14	21.21	(13,29) to (28,14)
1	22	12	19.24	(13,29) to (22,12)



K-Nearest Neighbors – Example (K=1)

Y	X1	X2	Yhat	distance			
0	13	29					
1	32	8		28.32		(13,29) to (32,8)	
0	19	21		10.00		(13,29) to (19,21)	
0	12	33	0	4.12	min	(13,29) to (12,33)	
1	28	14		21.21		(13,29) to (28,14)	
1	22	12		19.24		(13,29) to (22,12)	



K-Nearest Neighbors – Example (K=1)

Y	X1	X2	distance
0	13	29	28.32
1	32	8	
0	19	21	18.38
0	12	33	32.02
1	28	14	7.21
1	22	12	10.77



K-Nearest Neighbors – Example (K=1)

Y	X1	X2	distance	
0	13	29	28.32	(32,8) to (13,29)
1	32	8		
0	19	21	18.38	(32,8) to (19,21)
0	12	33	32.02	(32,8) to (12,33)
1	28	14	7.21	(32,8) to (28,14)
1	22	12	10.77	(32,8) to (22,12)



K-Nearest Neighbors – Example (K=1)

Y	X1	X2	Yhat	distance		
0	13	29		28.32		(32,8) to (13,29)
1	32	8				
0	19	21		18.38		(32,8) to (19,21)
0	12	33	0	32.02		(32,8) to (12,33)
1	28	14	1	7.21	min	(32,8) to (28,14)
1	22	12		10.77		(32,8) to (22,12)

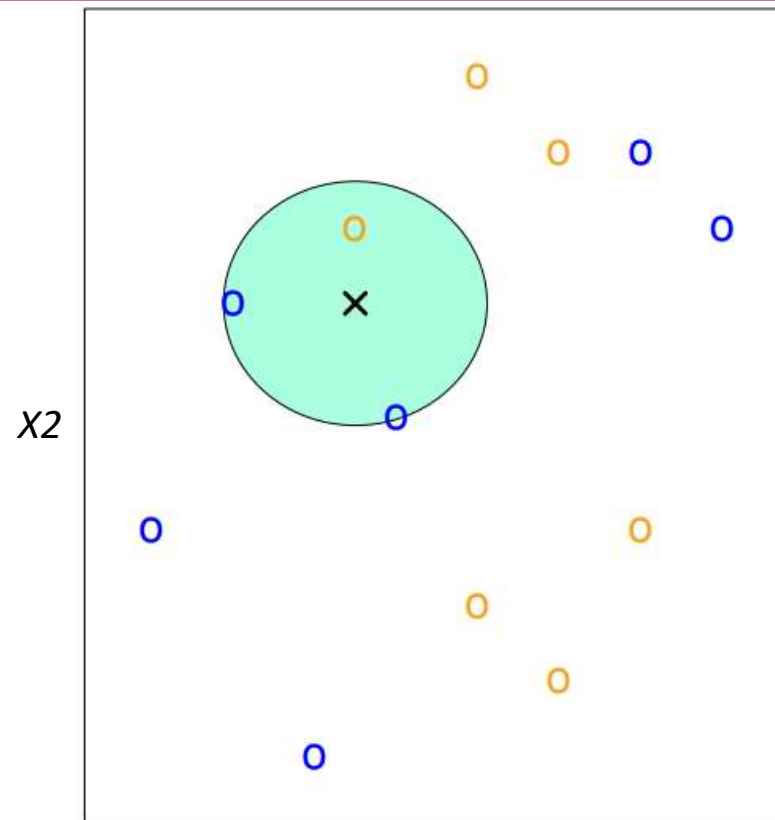


K-Nearest Neighbors – Example (K=1)

Y	X1	X2	Yhat	distance		
0	13	29		28.32		(32,8) to (13,29)
1	32	8				
0	19	21		18.38		(32,8) to (19,21)
0	12	33	0	32.02		(32,8) to (12,33)
1	28	14	1	7.21	min	(32,8) to (28,14)
1	22	12		10.77		(32,8) to (22,12)

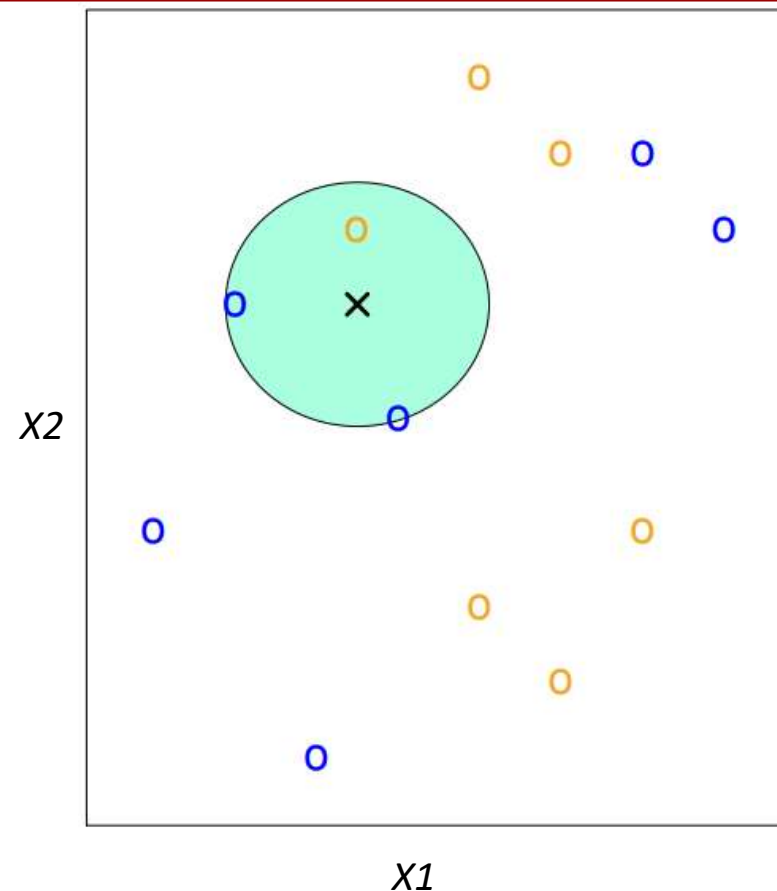


K-Nearest Neighbors – Example (K=3)





K-Nearest Neighbors – Example (K=3)



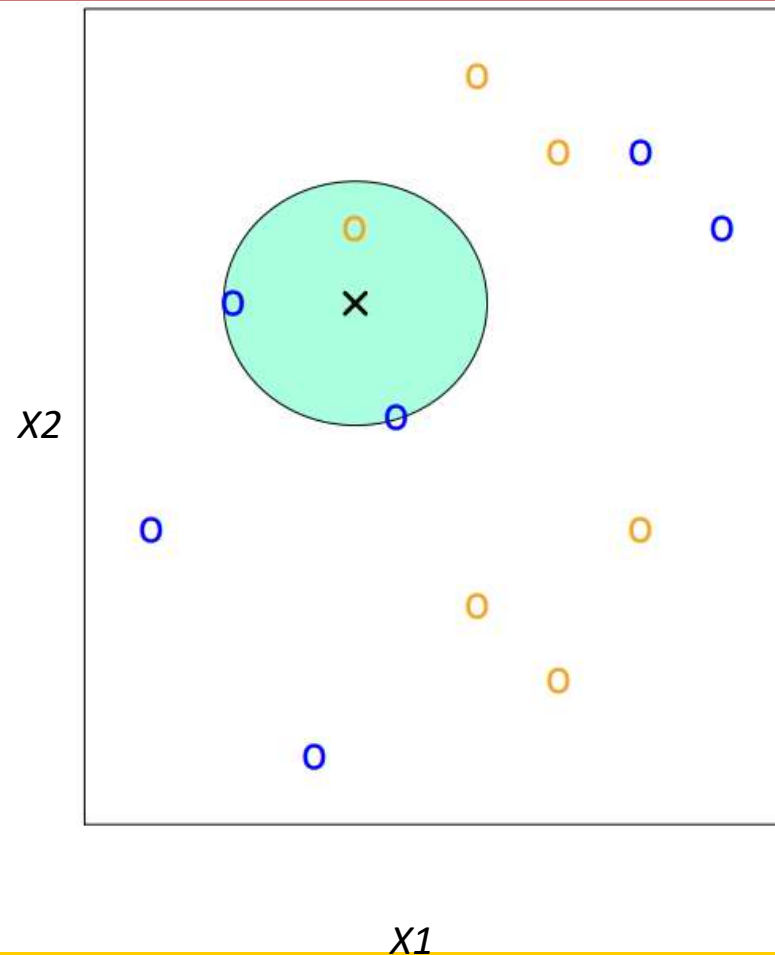
*Identify $k=3$ observations
closest to x*

*There are two
from category 1 (blue)
and one from
category 0 (orange)*

*Hence x is
predicted from
category 1 (blue)*



K-Nearest Neighbors – Example (K=3)



Identify $k=3$ observations closest to x

There are two from category 1 (blue) and one from

category 0 (orange)

Hence x is

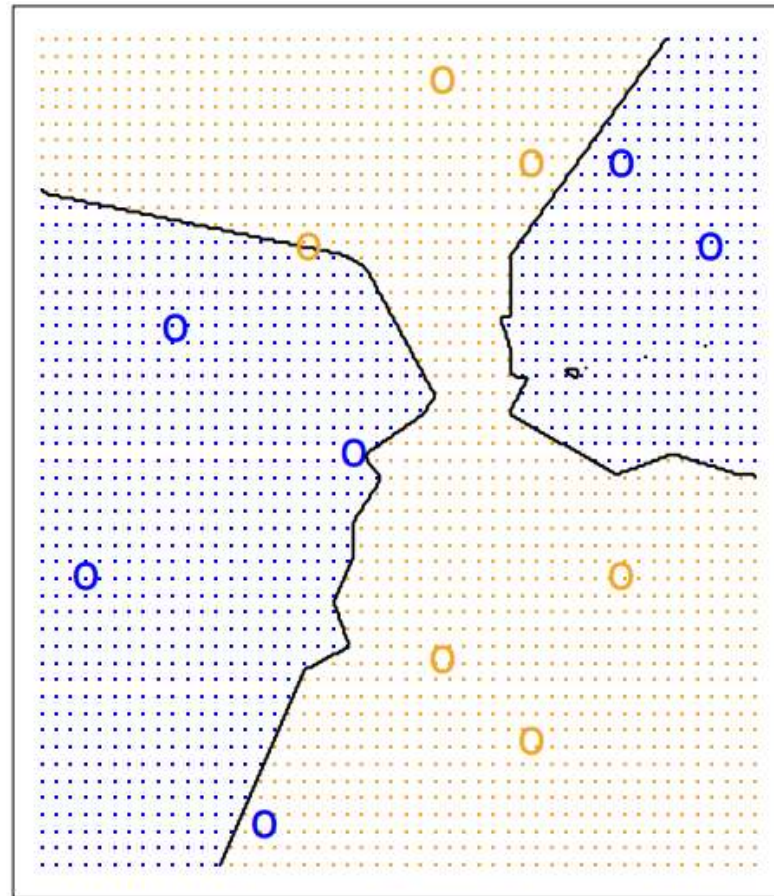
predicted from category 1 (blue)

with probability $2/3$



K-Nearest Neighbors – Example (K=3)

KNN (k=3) DECISION BOUNDARY



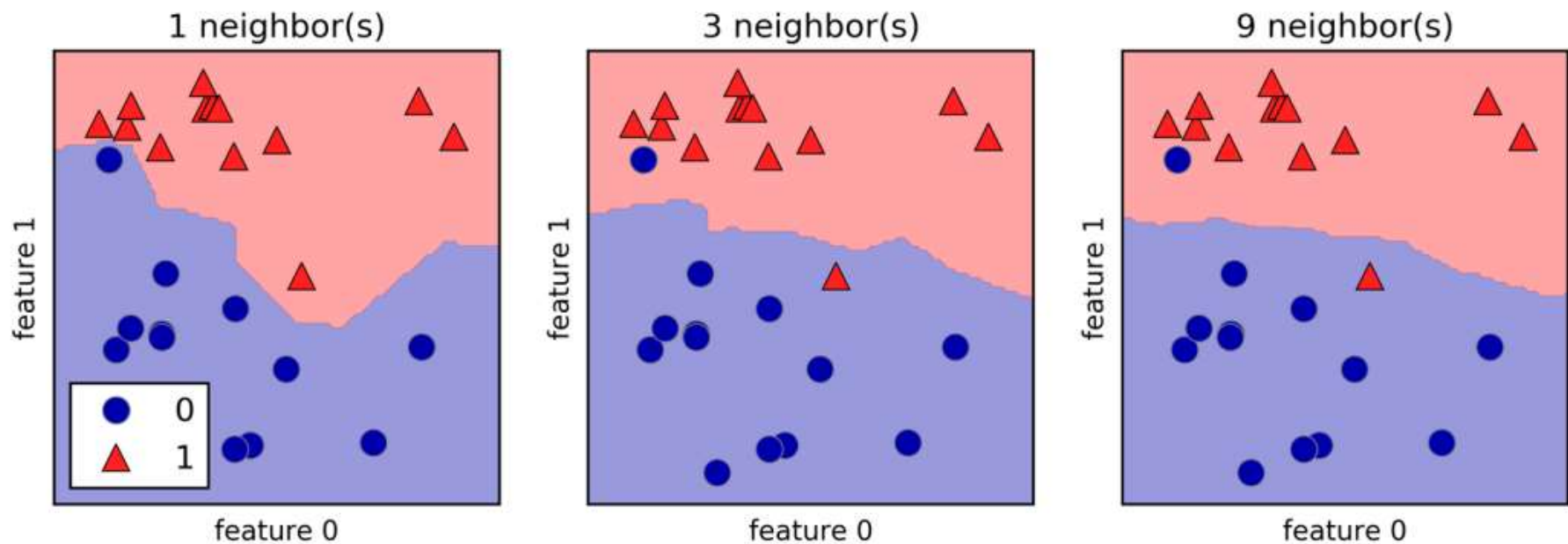


K-Nearest Neighbors – Assumptions

- All predictors are numeric
- KNN is a distance-based method (predictions may change with a different scale)
- large k results in a smooth decision boundary
- small k results in a more variable decision boundary



K-Nearest Neighbors, $k = 1, 3, 9$



Use Cross validation to select best value for k