

CS 171: Intro to ML and DM

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Slide Set 2: Probabilities



- From UC Riverside
 - ▶ CS 171: Introduction to Machine Learning and Data Mining
 - ▶ Professor Christian Shelton
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 - ▶ Elements of Statistical Learning (Hastie, et al.)
 - ▶ Pattern Recognition and Machine Learning (Bishop)
 - ▶ An Introduction to Machine Learning (Kubat)
 - ▶ Machine Learning: A Probabilistic Perspective (Murphy)
 - ▶ For use only by enrolled students in the course

A toy problem

x_1 (weekend?)	x_2 (weather)	x_3 (game?)	y (faster route)
no	sunny	no	Rte-113
no	sunny	yes	Rte-113
no	cloudy	yes	Rte-75
yes	sunny	no	Rte-113
no	raining	no	Rte-113
no	raining	yes	Rte-75
yes	cloudy	yes	Rte-75
no	sunny	no	Rte-113
yes	sunny	no	Rte-75
no	cloudy	no	Rte-113

A toy problem

x_1 (weekend?)	x_2 (weather)	x_3 (game?)	y (faster route)
no	sunny	no	Rte-113
no	sunny	yes	Rte-113
no	cloudy	yes	Rte-75
yes	sunny	no	Rte-113
no	raining	no	Rte-113
no	raining	yes	Rte-75
yes	cloudy	yes	Rte-75
no	sunny	no	Rte-113
yes	sunny	no	Rte-75
no	cloudy	no	Rte-113

A toy problem

x_1 (weekend?)	x_2 (weather)	x_3 (game?)	y (faster route)
no	sunny	no	Rte-113
no	sunny	yes	Rte-113
no	cloudy	yes	Rte-75
yes	sunny	no	Rte-113
no	raining	no	Rte-113
no	raining	yes	Rte-75
yes	cloudy	yes	Rte-75
no	sunny	no	Rte-113
yes	sunny	no	Rte-75
no	cloudy	no	Rte-113

$x_2 = \text{cloudy} \rightarrow y = \text{???}$

A toy problem

x_1 (weekend?)	x_2 (weather)	x_3 (game?)	y (faster route)
no	sunny	no	Rte-113
no	sunny	yes	Rte-113
no	cloudy	yes	Rte-75
yes	sunny	no	Rte-113
no	raining	no	Rte-113
no	raining	yes	Rte-75
yes	cloudy	yes	Rte-75
no	sunny	no	Rte-113
yes	sunny	no	Rte-75
no	cloudy	no	Rte-113

$x_2 = \text{cloudy} \rightarrow y = \text{????}$

$x_2 = \text{raining} \rightarrow y = \text{????}$

$x_2 = \text{sunny} \rightarrow y = \text{????}$

A toy problem

x_1 (weekend?)	x_2 (weather)	x_3 (game?)	y (faster route)
no	sunny	no	Rte-113
no	sunny	yes	Rte-113
no	cloudy	yes	Rte-75
yes	sunny	no	Rte-113
no	raining	no	Rte-113
no	raining	yes	Rte-75
yes	cloudy	yes	Rte-75
no	sunny	no	Rte-113
yes	sunny	no	Rte-75
no	cloudy	no	Rte-113

counts	Rte-113	Rte-75
cloudy	1	2
raining	1	1
sunny	4	1

$x_2 = \text{cloudy} \rightarrow y = \text{Rte-75}$

$x_2 = \text{raining} \rightarrow y = \text{????}$

$x_2 = \text{sunny} \rightarrow y = \text{Rte-113}$

A toy problem

x_1 (weekend?)	x_2 (weather)	x_3 (game?)	y (faster route)
no	sunny	no	Rte-113
no	sunny	yes	Rte-113
no	cloudy	yes	Rte-75
yes	sunny	no	Rte-113
no	raining	no	Rte-113
no	raining	yes	Rte-75
yes	cloudy	yes	Rte-75
no	sunny	no	Rte-113
yes	sunny	no	Rte-75
no	cloudy	no	Rte-113

$x_2 = \text{cloudy} \rightarrow y = \text{Rte-75}$

$x_2 = \text{raining} \rightarrow y = \text{????}$

$x_2 = \text{sunny} \rightarrow y = \text{Rte-113}$

counts	Rte-113	Rte-75
cloudy	1	2
raining	1	1
sunny	4	1

$\hat{P}(x_2, y)$	Rte-113	Rte-75
cloudy	0.1	0.2
raining	0.1	0.1
sunny	0.4	0.1

Bayes' rule

$\hat{P}(x_2, y)$	Rte-113	Rte-75
cloudy	0.1	0.2
raining	0.1	0.1
sunny	0.4	0.1

Bayes' rule

$\hat{P}(x_2, y)$	Rte-113	Rte-75		$\hat{P}(x_2)$
cloudy	0.1	0.2	$\left[\begin{array}{c} \\ \\ \end{array} \right]$	cloudy $\left[\begin{array}{c} 0.3 \\ \\ \end{array} \right]$
raining	0.1	0.1		raining $\left[\begin{array}{c} 0.2 \\ \\ \end{array} \right]$
sunny	0.4	0.1		sunny $\left[\begin{array}{c} 0.5 \\ \\ \end{array} \right]$

Bayes' rule

$\hat{P}(x_2, y)$	Rte-113	Rte-75		$\hat{P}(x_2)$
cloudy	0.1	0.2	$\left[\begin{array}{c} \\ \\ \end{array} \right]$	cloudy $\left[\begin{array}{c} 0.3 \\ \\ \end{array} \right]$
raining	0.1	0.1		raining $\left[\begin{array}{c} \\ 0.2 \\ \end{array} \right]$
sunny	0.4	0.1		sunny $\left[\begin{array}{c} \\ \\ 0.5 \end{array} \right]$

$\hat{P}(y)$	Rte-113	Rte-75
	$\left[\begin{array}{cc} 0.6 & 0.4 \end{array} \right]$	

Bayes' rule

$\hat{P}(x_2, y)$	Rte-113	Rte-75	$\hat{P}(x_2)$	$\hat{P}(y x_2)$	Rte-113	Rte-75
cloudy	0.1	0.2	cloudy	0.3	0.33	0.67
raining	0.1	0.1	raining	0.2	0.50	0.50
sunny	0.4	0.1	sunny	0.5	0.80	0.20

$\hat{P}(y)$	Rte-113	Rte-75
	0.6	0.4

Bayes' rule

$\hat{P}(x_2, y)$	Rte-113	Rte-75	$\hat{P}(x_2)$	$\hat{P}(y x_2)$	Rte-113	Rte-75
cloudy	0.1	0.2	cloudy	0.3	0.33	0.67
raining	0.1	0.1	raining	0.2	0.50	0.50
sunny	0.4	0.1	sunny	0.5	0.80	0.20

$\hat{P}(y)$	Rte-113	Rte-75
	0.6	0.4

$\hat{P}(x_2 y)$	Rte-113	Rte-75
cloudy	0.17	0.50
raining	0.17	0.25
sunny	0.67	0.25

Bayes' rule

$\hat{P}(x_2,y)$	Rte-113	Rte-75	$\hat{P}(x_2)$	$\hat{P}(y x_2)$	Rte-113	Rte-75	
cloudy	0.1	0.2	cloudy	0.3	cloudy	0.33	0.67
raining	0.1	0.1	raining	0.2	raining	0.50	0.50
sunny	0.4	0.1	sunny	0.5	sunny	0.80	0.20

$\hat{P}(y)$	Rte-113	Rte-75
	0.6	0.4

$$P(A=a) = \sum_b P(A=a, B=b)$$

$\hat{P}(x_2 y)$	Rte-113	Rte-75
cloudy	0.17	0.50
raining	0.17	0.25
sunny	0.67	0.25

Bayes' rule

$\hat{P}(x_2,y)$	Rte-113	Rte-75	$\hat{P}(x_2)$	$\hat{P}(y x_2)$	Rte-113	Rte-75	
cloudy	0.1	0.2	cloudy	0.3	cloudy	0.33	0.67
raining	0.1	0.1	raining	0.2	raining	0.50	0.50
sunny	0.4	0.1	sunny	0.5	sunny	0.80	0.20

$\hat{P}(y)$	Rte-113	Rte-75
	0.6	0.4

$$P(A=a) = \sum_b P(A=a, B=b)$$

$\hat{P}(x_2 y)$	Rte-113	Rte-75
cloudy	0.17	0.50
raining	0.17	0.25
sunny	0.67	0.25

$$\begin{aligned} P(A=a, B=b) &= P(A=a)P(B=b \mid A=a) \\ &= P(B=b)P(A=a \mid B=b) \end{aligned}$$

Bayes' rule

$\hat{P}(x_2, y)$	Rte-113	Rte-75
cloudy	0.1	0.2
raining	0.1	0.1
sunny	0.4	0.1

$\hat{P}(x_2)$		$\hat{P}(y x_2)$	Rte-113	Rte-75
cloudy	$\begin{bmatrix} 0.3 \\ 0.2 \\ 0.5 \end{bmatrix}$	cloudy	$\begin{bmatrix} 0.33 \\ 0.50 \\ 0.80 \end{bmatrix}$	$\begin{bmatrix} 0.67 \\ 0.50 \\ 0.20 \end{bmatrix}$
raining		raining		
sunny		sunny		

$\hat{P}(y)$	Rte-113	Rte-75
	$\begin{bmatrix} 0.6 \\ 0.4 \end{bmatrix}$	

$$P(A=a) = \sum_b P(A=a, B=b)$$

$\hat{P}(x_2 y)$	Rte-113	Rte-75
cloudy	0.17	0.50
raining	0.17	0.25
sunny	0.67	0.25

$$\begin{aligned} P(A=a, B=b) &= P(A=a)P(B=b | A=a) \\ &= P(B=b)P(A=a | B=b) \end{aligned}$$

$$P(A=a | B=b) = \frac{P(A=a, B=b)}{P(B=b)}$$

Bayes' rule

$\hat{P}(x_2, y)$	Rte-113	Rte-75
cloudy	0.1	0.2
raining	0.1	0.1
sunny	0.4	0.1

$\hat{P}(x_2)$		$\hat{P}(y x_2)$	Rte-113	Rte-75
cloudy	$\begin{bmatrix} 0.3 \\ 0.2 \\ 0.5 \end{bmatrix}$	cloudy	$\begin{bmatrix} 0.33 \\ 0.50 \\ 0.80 \end{bmatrix}$	$\begin{bmatrix} 0.67 \\ 0.50 \\ 0.20 \end{bmatrix}$
raining		raining		
sunny		sunny		

$\hat{P}(y)$	Rte-113	Rte-75
	$\begin{bmatrix} 0.6 \\ 0.4 \end{bmatrix}$	

$$P(A=a) = \sum_b P(A=a, B=b)$$

$\hat{P}(x_2 y)$	Rte-113	Rte-75
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Bayes' rule

$\hat{P}(x_2, y)$	Rte-113	Rte-75
cloudy	0.1	0.2
raining	0.1	0.1
sunny	0.4	0.1

$\hat{P}(x_2)$		$\hat{P}(y x_2)$	Rte-113	Rte-75
cloudy	$\begin{bmatrix} 0.3 \\ 0.2 \\ 0.5 \end{bmatrix}$	cloudy	$\begin{bmatrix} 0.33 \\ 0.50 \\ 0.80 \end{bmatrix}$	$\begin{bmatrix} 0.67 \\ 0.50 \\ 0.20 \end{bmatrix}$
raining		raining		
sunny		sunny		

$\hat{P}(y)$	Rte-113	Rte-75
	$\begin{bmatrix} 0.6 \\ 0.4 \end{bmatrix}$	

$$P(A=a) = \sum_b P(A=a, B=b)$$

$\hat{P}(x_2 y)$	Rte-113	Rte-75
cloudy	0.17	0.50
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$$\begin{aligned} P(A=a, B=b) &= P(A=a)P(B=b | A=a) \\ &= P(B=b)P(A=a | B=b) \end{aligned}$$

$$P(A=a | B=b) = \frac{P(A=a, B=b)}{P(B=b)}$$

$$P(A=a | B=b) = \frac{P(B=b|A=a)P(A=a)}{P(B=b)} \quad (\text{Bayes' Theorem})$$

Bayes' rule

$\hat{P}(x_2, y)$	Rte-113	Rte-75
cloudy	0.1	0.2
raining	0.1	0.1
sunny	0.4	0.1

$\hat{P}(x_2)$		$\hat{P}(y x_2)$	Rte-113	Rte-75
cloudy	$\begin{bmatrix} 0.3 \\ 0.2 \\ 0.5 \end{bmatrix}$	cloudy	$\begin{bmatrix} 0.33 \\ 0.50 \\ 0.80 \end{bmatrix}$	$\begin{bmatrix} 0.67 \\ 0.50 \\ 0.20 \end{bmatrix}$
raining		raining		
sunny		sunny		

$\hat{P}(y)$	Rte-113	Rte-75
	$\begin{bmatrix} 0.6 \\ 0.4 \end{bmatrix}$	

$$P(A=a) = \sum_b P(A=a, B=b)$$

$\hat{P}(x_2 y)$	Rte-113	Rte-75
cloudy	$\begin{bmatrix} 0.17 \\ 0.17 \\ 0.67 \end{bmatrix}$	$\begin{bmatrix} 0.50 \\ 0.25 \\ 0.25 \end{bmatrix}$
raining		
sunny		

$$\begin{aligned} P(A=a, B=b) &= P(A=a)P(B=b | A=a) \\ &= P(B=b)P(A=a | B=b) \end{aligned}$$

$$P(A=a | B=b) = \frac{P(A=a, B=b)}{P(B=b)}$$

$$P(A=a | B=b) = \frac{P(B=b|A=a)P(A=a)}{P(B=b)} \quad (\text{Bayes' Theorem})$$

$$\text{Example: } \hat{P}(\text{Rte-113} | \text{sunny}) = \frac{\hat{P}(\text{sunny}|\text{Rte-113})\hat{P}(\text{Rte-113})}{\hat{P}(\text{sunny})} = \frac{0.67 \times 0.6}{0.5} = 0.8$$

Bayes-optimal classifier

$(x_1 = \text{no}, x_2 = \text{sunny}, x_3 = \text{no}) \rightarrow y = \text{???}$

Bayes-optimal classifier

$(x_1 = \text{no}, x_2 = \text{sunny}, x_3 = \text{no}) \rightarrow y = \text{???}$

Bayes-optimal classifier: Given $x = (x_1, x_2, \dots, x_n)$, predict $y = \arg \max_c P(y = c \mid x)$

Bayes-optimal classifier

$(x_1 = \text{no}, x_2 = \text{sunny}, x_3 = \text{no}) \rightarrow y = \text{????}$

Bayes-optimal classifier: Given $x = (x_1, x_2, \dots, x_n)$, predict $y = \arg \max_c P(y = c \mid x)$

Using estimated probabilities: Given $x = (x_1, x_2, \dots, x_n)$, predict $y = \arg \max_c \hat{P}(y = c \mid x)$

Bayes-optimal classifier

x_1 (weekend?)	x_2 (weather)	x_3 (game?)	y (faster route)
no	sunny	no	Rte-113
no	sunny	yes	Rte-113
no	cloudy	yes	Rte-75
yes	sunny	no	Rte-113
no	raining	no	Rte-113
no	raining	yes	Rte-75
yes	cloudy	yes	Rte-75
no	sunny	no	Rte-113
yes	sunny	no	Rte-75
no	cloudy	no	Rte-113

$(x_1 = \text{no}, x_2 = \text{sunny}, x_3 = \text{no}) \rightarrow y = \text{????}$

Bayes-optimal classifier: Given $x = (x_1, x_2, \dots, x_n)$, predict $y = \arg \max_c P(y = c \mid x)$

Using estimated probabilities: Given $x = (x_1, x_2, \dots, x_n)$, predict $y = \arg \max_c \hat{P}(y = c \mid x)$

Bayes-optimal classifier

x_1 (weekend?)	x_2 (weather)	x_3 (game?)	y (faster route)	$\hat{P}(x,y)$	Rte-113	Rte-75
no	sunny	no	Rte-113	(no,cloudy,no)	0.1	0.0
no	sunny	yes	Rte-113	(no,cloudy,yes)	0.0	0.1
no	cloudy	yes	Rte-75	(no,raining,no)	0.1	0.0
yes	sunny	no	Rte-113	(no,raining,yes)	0.0	0.1
no	raining	no	Rte-113	(no,sunny,no)	0.2	0.0
no	raining	yes	Rte-75	(no,sunny,yes)	0.1	0.0
yes	cloudy	yes	Rte-75	(yes,cloudy,no)	0.0	0.0
no	sunny	no	Rte-113	(yes,cloudy,yes)	0.0	0.1
yes	sunny	no	Rte-75	(yes,raining,no)	0.0	0.0
no	cloudy	no	Rte-113	(yes,raining,yes)	0.0	0.0
				(yes,sunny,no)	0.1	0.1
				(yes,sunny,yes)	0.0	0.0

$(x_1 = \text{no}, x_2 = \text{sunny}, x_3 = \text{no}) \rightarrow y = \text{????}$

Bayes-optimal classifier: Given $x = (x_1, x_2, \dots, x_n)$, predict $y = \arg \max_c P(y = c \mid x)$

Using estimated probabilities: Given $x = (x_1, x_2, \dots, x_n)$, predict $y = \arg \max_c \hat{P}(y = c \mid x)$

Naïve Bayes

Assume: $P(x_1, x_2, \dots, x_n \mid y) = \prod_{i=1}^n P(x_i \mid y)$

$[X_i \text{ is independent of } X_j, \text{ given } Y]$

Naïve Bayes

Assume: $P(x_1, x_2, \dots, x_n \mid y) = \prod_{i=1}^n P(x_i \mid y)$

$[X_i \text{ is independent of } X_j, \text{ given } Y]$

Then use $\hat{P}(x_i \mid y)$ instead of $P(x_i \mid y)$.

Naïve Bayes

Assume: $P(x_1, x_2, \dots, x_n \mid y) = \prod_{i=1}^n P(x_i \mid y)$

$[X_i \text{ is independent of } X_j, \text{ given } Y]$

Then use $\hat{P}(x_i \mid y)$ instead of $P(x_i \mid y)$.

$$\begin{aligned}\hat{P}(y \mid x) &= \frac{\hat{P}(x \mid y) \hat{P}(y)}{\hat{P}(x)} \\ &= \frac{\hat{P}(y) \prod_{i=1}^n \hat{P}(x_i \mid y)}{\hat{P}(x)}\end{aligned}$$

Naïve Bayes

Assume: $P(x_1, x_2, \dots, x_n \mid y) = \prod_{i=1}^n P(x_i \mid y)$

$[X_i \text{ is independent of } X_j, \text{ given } Y]$

Then use $\hat{P}(x_i \mid y)$ instead of $P(x_i \mid y)$.

$$\begin{aligned}\hat{P}(y \mid x) &= \frac{\hat{P}(x \mid y) \hat{P}(y)}{\hat{P}(x)} \\ &= \frac{\hat{P}(y) \prod_{i=1}^n \hat{P}(x_i \mid y)}{\hat{P}(x)}\end{aligned}$$

Naïve Bayes rule:

Given $x = (x_1, x_2, \dots, x_n)$,

Pick c that maximizes $\hat{P}(y=c) \prod_{i=1}^n \hat{P}(x_i \mid y=c)$

Naïve Bayes Example

x_1 (weekend?)	x_2 (weather)	x_3 (game?)	y (faster route)
no	sunny	no	Rte-113
no	sunny	yes	Rte-113
no	cloudy	yes	Rte-75
yes	sunny	no	Rte-113
no	raining	no	Rte-113
no	raining	yes	Rte-75
yes	cloudy	yes	Rte-75
no	sunny	no	Rte-113
yes	sunny	no	Rte-75
no	cloudy	no	Rte-113

Naïve Bayes Example

x_1 (weekend?)	x_2 (weather)	x_3 (game?)	y (faster route)
no	sunny	no	Rte-113
no	sunny	yes	Rte-113
no	cloudy	yes	Rte-75
yes	sunny	no	Rte-113
no	raining	no	Rte-113
no	raining	yes	Rte-75
yes	cloudy	yes	Rte-75
no	sunny	no	Rte-113
yes	sunny	no	Rte-75
no	cloudy	no	Rte-113

$\hat{P}(y)$	Rte-113	Rte-75
	[0.6	0.4]

$\hat{P}(x_1 y)$	Rte-113	Rte-75
no	[0.83	0.50]
yes	[0.17	0.50]

$\hat{P}(x_2 y)$	Rte-113	Rte-75
cloudy	[0.17	0.50]
raining	[0.17	0.25]
sunny	[0.67	0.25]

$\hat{P}(x_3 y)$	Rte-113	Rte-75
no	[0.83	0.25]
yes	[0.17	0.75]

Naïve Bayes Example

Input: $x = (\text{no}, \text{cloudy}, \text{no})$

$$\hat{P}(y) \quad \begin{array}{cc} \text{Rte-113} & \text{Rte-75} \\ \left[\begin{array}{cc} 0.6 & 0.4 \end{array} \right] \end{array}$$

$$\hat{P}(x_1|y) \quad \begin{array}{cc} \text{Rte-113} & \text{Rte-75} \\ \begin{array}{c} \text{no} \\ \text{yes} \end{array} \left[\begin{array}{cc} 0.83 & 0.50 \\ 0.17 & 0.50 \end{array} \right] \end{array}$$

$$\hat{P}(x_2|y) \quad \begin{array}{cc} \text{Rte-113} & \text{Rte-75} \\ \begin{array}{c} \text{cloudy} \\ \text{raining} \\ \text{sunny} \end{array} \left[\begin{array}{cc} 0.17 & 0.50 \\ 0.17 & 0.25 \\ 0.67 & 0.25 \end{array} \right] \end{array}$$

$$\hat{P}(x_3|y) \quad \begin{array}{cc} \text{Rte-113} & \text{Rte-75} \\ \begin{array}{c} \text{no} \\ \text{yes} \end{array} \left[\begin{array}{cc} 0.83 & 0.25 \\ 0.17 & 0.75 \end{array} \right] \end{array}$$

Naïve Bayes Example

Input: $x = (\text{no}, \text{cloudy}, \text{no})$

Possibility: $y = \text{Rte-113}$

$$\hat{P}(y) \quad \begin{array}{cc} \text{Rte-113} & \text{Rte-75} \\ \left[\begin{array}{cc} 0.6 & 0.4 \end{array} \right] \end{array}$$

$$\hat{P}(x_1|y) \quad \begin{array}{cc} \text{Rte-113} & \text{Rte-75} \\ \begin{array}{c} \text{no} \\ \text{yes} \end{array} \left[\begin{array}{cc} 0.83 & 0.50 \\ 0.17 & 0.50 \end{array} \right] \end{array}$$

$$\hat{P}(x_2|y) \quad \begin{array}{cc} \text{Rte-113} & \text{Rte-75} \\ \begin{array}{c} \text{cloudy} \\ \text{raining} \\ \text{sunny} \end{array} \left[\begin{array}{cc} 0.17 & 0.50 \\ 0.17 & 0.25 \\ 0.67 & 0.25 \end{array} \right] \end{array}$$

$$\hat{P}(x_3|y) \quad \begin{array}{cc} \text{Rte-113} & \text{Rte-75} \\ \begin{array}{c} \text{no} \\ \text{yes} \end{array} \left[\begin{array}{cc} 0.83 & 0.25 \\ 0.17 & 0.75 \end{array} \right] \end{array}$$

Naïve Bayes Example

Input: $x = (\text{no}, \text{cloudy}, \text{no})$

Possibility: $y = \text{Rte-113}$

$\hat{P}(y=\text{Rte-113})\hat{P}(x_1=\text{no} \mid y=\text{Rte-113})\hat{P}(x_2=\text{cloudy} \mid y=\text{Rte-113})\hat{P}(x_3=\text{no} \mid y=\text{Rte-113})$

$\hat{P}(y)$	Rte-113	Rte-75
	0.6	0.4

$\hat{P}(x_1 y)$	Rte-113	Rte-75
no	0.83	0.50
yes	0.17	0.50

$\hat{P}(x_2 y)$	Rte-113	Rte-75
cloudy	0.17	0.50
raining	0.17	0.25
sunny	0.67	0.25

$\hat{P}(x_3 y)$	Rte-113	Rte-75
no	0.83	0.25
yes	0.17	0.75

Naïve Bayes Example

Input: $x = (\text{no}, \text{cloudy}, \text{no})$

Possibility: $y = \text{Rte-113}$

$$\hat{P}(y=\text{Rte-113})\hat{P}(x_1=\text{no} \mid y=\text{Rte-113})\hat{P}(x_2=\text{cloudy} \mid y=\text{Rte-113})\hat{P}(x_3=\text{no} \mid y=\text{Rte-113}) \\ = 0.6 \times 0.83 \times 0.17 \times 0.83 = 0.0694$$

$$\hat{P}(y) \quad \begin{array}{cc} \text{Rte-113} & \text{Rte-75} \\ \left[\begin{array}{cc} 0.6 & 0.4 \end{array} \right] \end{array}$$

$$\hat{P}(x_1|y) \quad \begin{array}{cc} \text{Rte-113} & \text{Rte-75} \\ \text{no} & \left[\begin{array}{cc} 0.83 & 0.50 \end{array} \right] \\ \text{yes} & \left[\begin{array}{cc} 0.17 & 0.50 \end{array} \right] \end{array}$$

$$\hat{P}(x_2|y) \quad \begin{array}{cc} \text{Rte-113} & \text{Rte-75} \\ \text{cloudy} & \left[\begin{array}{cc} 0.17 & 0.50 \end{array} \right] \\ \text{raining} & \left[\begin{array}{cc} 0.17 & 0.25 \end{array} \right] \\ \text{sunny} & \left[\begin{array}{cc} 0.67 & 0.25 \end{array} \right] \end{array}$$

$$\hat{P}(x_3|y) \quad \begin{array}{cc} \text{Rte-113} & \text{Rte-75} \\ \text{no} & \left[\begin{array}{cc} 0.83 & 0.25 \end{array} \right] \\ \text{yes} & \left[\begin{array}{cc} 0.17 & 0.75 \end{array} \right] \end{array}$$

Naïve Bayes Example

Input: $x = (\text{no}, \text{cloudy}, \text{no})$

Possibility: $y = \text{Rte-113}$

$$\hat{P}(y=\text{Rte-113})\hat{P}(x_1=\text{no} \mid y=\text{Rte-113})\hat{P}(x_2=\text{cloudy} \mid y=\text{Rte-113})\hat{P}(x_3=\text{no} \mid y=\text{Rte-113}) \\ = 0.6 \times 0.83 \times 0.17 \times 0.83 = 0.0694$$

Possibility: $y = \text{Rte-75}$

$$\hat{P}(y) \quad \begin{array}{cc} \text{Rte-113} & \text{Rte-75} \\ \left[\begin{array}{cc} 0.6 & 0.4 \end{array} \right] \end{array}$$

$$\hat{P}(x_1|y) \quad \begin{array}{cc} \text{Rte-113} & \text{Rte-75} \\ \begin{array}{c} \text{no} \\ \text{yes} \end{array} \left[\begin{array}{cc} 0.83 & 0.50 \\ 0.17 & 0.50 \end{array} \right] \end{array}$$

$$\hat{P}(x_2|y) \quad \begin{array}{cc} \text{Rte-113} & \text{Rte-75} \\ \begin{array}{c} \text{cloudy} \\ \text{raining} \\ \text{sunny} \end{array} \left[\begin{array}{cc} 0.17 & 0.50 \\ 0.17 & 0.25 \\ 0.67 & 0.25 \end{array} \right] \end{array}$$

$$\hat{P}(x_3|y) \quad \begin{array}{cc} \text{Rte-113} & \text{Rte-75} \\ \begin{array}{c} \text{no} \\ \text{yes} \end{array} \left[\begin{array}{cc} 0.83 & 0.25 \\ 0.17 & 0.75 \end{array} \right] \end{array}$$

Naïve Bayes Example

Input: $x = (\text{no}, \text{cloudy}, \text{no})$

Possibility: $y = \text{Rte-113}$

$$\hat{P}(y=\text{Rte-113})\hat{P}(x_1=\text{no} \mid y=\text{Rte-113})\hat{P}(x_2=\text{cloudy} \mid y=\text{Rte-113})\hat{P}(x_3=\text{no} \mid y=\text{Rte-113}) \\ = 0.6 \times 0.83 \times 0.17 \times 0.83 = 0.0694$$

Possibility: $y = \text{Rte-75}$

$$\hat{P}(y=\text{Rte-75})\hat{P}(x_1=\text{no} \mid y=\text{Rte-75})\hat{P}(x_2=\text{cloudy} \mid y=\text{Rte-75})\hat{P}(x_3=\text{no} \mid y=\text{Rte-75})$$

$\hat{P}(y)$	Rte-113	Rte-75
	0.6	0.4

$\hat{P}(x_1 y)$	Rte-113	Rte-75
no	0.83	0.50
yes	0.17	0.50

$\hat{P}(x_2 y)$	Rte-113	Rte-75
cloudy	0.17	0.50
raining	0.17	0.25
sunny	0.67	0.25

$\hat{P}(x_3 y)$	Rte-113	Rte-75
no	0.83	0.25
yes	0.17	0.75

Naïve Bayes Example

Input: $x = (\text{no}, \text{cloudy}, \text{no})$

Possibility: $y = \text{Rte-113}$

$$\hat{P}(y=\text{Rte-113})\hat{P}(x_1=\text{no} \mid y=\text{Rte-113})\hat{P}(x_2=\text{cloudy} \mid y=\text{Rte-113})\hat{P}(x_3=\text{no} \mid y=\text{Rte-113}) \\ = 0.6 \times 0.83 \times 0.17 \times 0.83 = 0.0694$$

Possibility: $y = \text{Rte-75}$

$$\hat{P}(y=\text{Rte-75})\hat{P}(x_1=\text{no} \mid y=\text{Rte-75})\hat{P}(x_2=\text{cloudy} \mid y=\text{Rte-75})\hat{P}(x_3=\text{no} \mid y=\text{Rte-75}) \\ = 0.4 \times 0.50 \times 0.50 \times 0.25 = 0.025$$

$$\hat{P}(y) \quad \begin{array}{cc} \text{Rte-113} & \text{Rte-75} \\ \left[\begin{array}{cc} 0.6 & 0.4 \end{array} \right] \end{array}$$

$$\hat{P}(x_1|y) \quad \begin{array}{cc} \text{Rte-113} & \text{Rte-75} \\ \begin{array}{c} \text{no} \\ \text{yes} \end{array} \left[\begin{array}{cc} 0.83 & 0.50 \\ 0.17 & 0.50 \end{array} \right] \end{array}$$

$$\hat{P}(x_2|y) \quad \begin{array}{cc} \text{Rte-113} & \text{Rte-75} \\ \begin{array}{c} \text{cloudy} \\ \text{raining} \\ \text{sunny} \end{array} \left[\begin{array}{cc} 0.17 & 0.50 \\ 0.17 & 0.25 \\ 0.67 & 0.25 \end{array} \right] \end{array}$$

$$\hat{P}(x_3|y) \quad \begin{array}{cc} \text{Rte-113} & \text{Rte-75} \\ \begin{array}{c} \text{no} \\ \text{yes} \end{array} \left[\begin{array}{cc} 0.83 & 0.25 \\ 0.17 & 0.75 \end{array} \right] \end{array}$$

Naïve Bayes Example

Input: $x = (\text{no}, \text{cloudy}, \text{no})$

Possibility: $y = \text{Rte-113}$

$$\hat{P}(y=\text{Rte-113})\hat{P}(x_1=\text{no} \mid y=\text{Rte-113})\hat{P}(x_2=\text{cloudy} \mid y=\text{Rte-113})\hat{P}(x_3=\text{no} \mid y=\text{Rte-113}) \\ = 0.6 \times 0.83 \times 0.17 \times 0.83 = 0.0694$$

Possibility: $y = \text{Rte-75}$

$$\hat{P}(y=\text{Rte-75})\hat{P}(x_1=\text{no} \mid y=\text{Rte-75})\hat{P}(x_2=\text{cloudy} \mid y=\text{Rte-75})\hat{P}(x_3=\text{no} \mid y=\text{Rte-75}) \\ = 0.4 \times 0.50 \times 0.50 \times 0.25 = 0.025$$

Therefore, the rule reports Rte-113

$\hat{P}(y)$	Rte-113	Rte-75
	0.6	0.4

$\hat{P}(x_1 y)$	Rte-113	Rte-75
no	0.83	0.50
yes	0.17	0.50

$\hat{P}(x_2 y)$	Rte-113	Rte-75
cloudy	0.17	0.50
raining	0.17	0.25
sunny	0.67	0.25

$\hat{P}(x_3 y)$	Rte-113	Rte-75
no	0.83	0.25
yes	0.17	0.75

Naïve Bayes Example II

Input: $x = (\text{yes}, \text{raining}, \text{no})$

Possibility: $y = \text{Rte-113}$

$$\hat{P}(y=\text{Rte-113})\hat{P}(x_1=\text{yes} \mid y=\text{Rte-113})\hat{P}(x_2=\text{raining} \mid y=\text{Rte-113})\hat{P}(x_3=\text{no} \mid y=\text{Rte-113}) \\ = 0.6 \times 0.17 \times 0.17 \times 0.83 = 0.0139$$

Possibility: $y = \text{Rte-75}$

$$\hat{P}(y=\text{Rte-75})\hat{P}(x_1=\text{yes} \mid y=\text{Rte-75})\hat{P}(x_2=\text{raining} \mid y=\text{Rte-75})\hat{P}(x_3=\text{no} \mid y=\text{Rte-75}) \\ = 0.4 \times 0.50 \times 0.25 \times 0.25 = 0.0125$$

Therefore, the rule reports Rte-113

$$\hat{P}(y) \quad \begin{array}{cc} \text{Rte-113} & \text{Rte-75} \\ \left[\begin{array}{cc} 0.6 & 0.4 \end{array} \right] \end{array}$$

$$\hat{P}(x_1|y) \quad \begin{array}{cc} \text{Rte-113} & \text{Rte-75} \\ \text{no} & \left[\begin{array}{cc} 0.83 & 0.50 \end{array} \right] \\ \text{yes} & \left[\begin{array}{cc} 0.17 & 0.50 \end{array} \right] \end{array}$$

$$\hat{P}(x_2|y) \quad \begin{array}{cc} \text{Rte-113} & \text{Rte-75} \\ \text{cloudy} & \left[\begin{array}{cc} 0.17 & 0.50 \end{array} \right] \\ \text{raining} & \left[\begin{array}{cc} 0.17 & 0.25 \end{array} \right] \\ \text{sunny} & \left[\begin{array}{cc} 0.67 & 0.25 \end{array} \right] \end{array}$$

$$\hat{P}(x_3|y) \quad \begin{array}{cc} \text{Rte-113} & \text{Rte-75} \\ \text{no} & \left[\begin{array}{cc} 0.83 & 0.25 \end{array} \right] \\ \text{yes} & \left[\begin{array}{cc} 0.17 & 0.75 \end{array} \right] \end{array}$$

Naïve Bayes Example III

Input: $x = (\text{yes}, \text{sunny}, \text{yes})$

Possibility: $y = \text{Rte-113}$

$$\hat{P}(y=\text{Rte-113})\hat{P}(x_1=\text{yes} \mid y=\text{Rte-113})\hat{P}(x_2=\text{sunny} \mid y=\text{Rte-113})\hat{P}(x_3=\text{yes} \mid y=\text{Rte-113}) \\ = 0.6 \times 0.17 \times 0.67 \times 0.17 = 0.0111$$

Possibility: $y = \text{Rte-75}$

$$\hat{P}(y=\text{Rte-75})\hat{P}(x_1=\text{yes} \mid y=\text{Rte-75})\hat{P}(x_2=\text{sunny} \mid y=\text{Rte-75})\hat{P}(x_3=\text{yes} \mid y=\text{Rte-75}) \\ = 0.4 \times 0.50 \times 0.25 \times 0.75 = 0.0375$$

Therefore, the rule reports Rte-75

$$\hat{P}(y) \quad \begin{array}{cc} \text{Rte-113} & \text{Rte-75} \\ \left[\begin{array}{cc} 0.6 & 0.4 \end{array} \right] \end{array}$$

$$\hat{P}(x_1|y) \quad \begin{array}{cc} \text{Rte-113} & \text{Rte-75} \\ \text{no} & \left[\begin{array}{cc} 0.83 & 0.50 \end{array} \right] \\ \text{yes} & \left[\begin{array}{cc} 0.17 & 0.50 \end{array} \right] \end{array}$$

$$\hat{P}(x_2|y) \quad \begin{array}{cc} \text{Rte-113} & \text{Rte-75} \\ \text{cloudy} & \left[\begin{array}{cc} 0.17 & 0.50 \end{array} \right] \\ \text{raining} & \left[\begin{array}{cc} 0.17 & 0.25 \end{array} \right] \\ \text{sunny} & \left[\begin{array}{cc} 0.67 & 0.25 \end{array} \right] \end{array}$$

$$\hat{P}(x_3|y) \quad \begin{array}{cc} \text{Rte-113} & \text{Rte-75} \\ \text{no} & \left[\begin{array}{cc} 0.83 & 0.25 \end{array} \right] \\ \text{yes} & \left[\begin{array}{cc} 0.17 & 0.75 \end{array} \right] \end{array}$$