

CSCI 347: Introduction to Data Mining

Week 4b - Categorical Data

BUT FIRST.... RETRO ON HOMEWORK 1

Estimated variance of X_j : $\hat{\sigma}_j^2 = \frac{1}{n-1} \sum_{i=1}^n (x_{ij} - \hat{\mu}_j)^2$

Estimated standard deviation of X_j : $\hat{\sigma}_j = \sqrt{\hat{\sigma}_j^2}$

Covariance of X_i and X_j : $\hat{\sigma}_{ij} = \frac{1}{n-1} \sum_{k=1}^n (x_{ki} - \hat{\mu}_i)(x_{kj} - \hat{\mu}_j)$

Person's correlation coefficient of X_i and X_j : $\hat{\rho}_{ij} = \frac{\hat{\sigma}_{ij}}{\hat{\sigma}_i \hat{\sigma}_j}$

COMMON DATA TYPES

- Data is most often either *numerical* or *categorical*

	temperature	length	type	weight
$D =$ specimen 1	0.2	23	A	5.7
specimen 2	0.4	1	B	5.4
specimen 3	1.8	0.5	C	5.2
specimen 4	5.6	50	A	5.1
specimen 5	-0.5	34	A	5.3
specimen 6	0.4	19	B	5.4
specimen 7	1.1	11	A	5.5

RECALL: EUCLIDEAN DISTANCE

L_2 norm:

$$\|x_i - x_j\|_2 = \sqrt{\sum_{k=1}^m (x_{ik} - x_{jk})^2} \quad \text{where } x_i \text{ and } x_j \text{ are vectors, and there are } m \text{ dimensions}$$

	X_1	X_2	X_3
x_1	0.2	23	5.7
x_2	0.4	1	5.4
x_3	1.8	0.5	5.2
x_4	5.6	50	5.1
x_5	-0.5	34	5.3
x_6	0.4	19	5.4
x_7	1.1	11	5.5

$$\begin{aligned} \|x_1 - x_2\|_2 &= \sqrt{\sum_{k=1}^3 (x_{1k} - x_{2k})^2} \\ &= \sqrt{(x_{11} - x_{21})^2 + (x_{12} - x_{22})^2 + (x_{13} - x_{23})^2} \\ &= \sqrt{(0.2 - 0.4)^2 + (23 - 1)^2 + (5.7 - 5.4)^2} \\ &= \sqrt{(-0.2)^2 + (22)^2 + (0.3)^2} \\ &= 22.0 \end{aligned}$$

WHAT IF WE ALSO HAVE CATEGORICAL VARIABLES

L_2 norm:

$$\|x_i - x_j\|_2 = \sqrt{\sum_{k=1}^m (x_{ik} - x_{jk})^2} \quad \text{where } x_i \text{ and } x_j \text{ are vectors, and there are } m \text{ dimensions}$$

	X_1	X_2	X_3	X_4	$\ x_1 - x_2\ _2 = \sqrt{\sum_{k=1}^4 (x_{1k} - x_{2k})^2}$ $= \sqrt{(x_{11} - x_{21})^2 + (x_{12} - x_{22})^2 + (x_{13} - x_{23})^2 + (x_{14} - x_{24})^2}$ $= \sqrt{(0.2 - 0.4)^2 + (23 - 1)^2 + (5.7 - 5.4)^2 + (A - B)^2}$
x_1	0.2	23	5.7	A	
x_2	0.4	1.	5.4	B	
x_3	1.8	0.5	5.2	C	
x_4	5.6	50	5.1	A	
x_5	-0.5	34	5.3	B	
x_6	0.4	19	5.4	C	

$D =$

x_7	1.1	11	5.5	C
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LABEL ENCODING

L_2 norm:

$$\|x_i - x_j\|_2 = \sqrt{\sum_{k=1}^m (x_{ik} - x_{jk})^2}$$
 where x_i and x_j are vectors, and there are m dimensions

$A => 0$

$B => 1$

$C => 2$

$D =$		X_1	X_2	X_3	X_4
	x_1	0.2	23	5.7	A
	x_2	0.4	1.	5.4	B
	x_3	1.8	0.5	5.2	C
	x_4	5.6	50	5.1	A
	x_5	-0.5	34	5.3	B
	x_6	0.4	19	5.4	C
	x_7	1.1	11	5.5	C

$$\|x_1 - x_2\|_2 = \sqrt{\sum_{k=1}^4 (x_{1k} - x_{2k})^2}$$

$$= \sqrt{(x_{11} - x_{21})^2 + (x_{12} - x_{22})^2 + (x_{13} - x_{23})^2 + (x_{14} - x_{24})^2}$$

$$= \sqrt{(0.2 - 0.4)^2 + (23 - 1)^2 + (5.7 - 5.4)^2 + (A - B)^2}$$

LABEL ENCODING

L_2 norm:

$\|x_i - x_j\|_2 = \sqrt{\sum_{k=1}^m (x_{ik} - x_{jk})^2}$ where x_i and x_j are vectors, and there are m dimensions

$A => 0$

$B => 1$

$C => 2$

$D =$

	X_1	X_2	X_3	X_4
x_1	0.2	23	5.7	0
x_2	0.4	1.	5.4	1
x_3	1.8	0.5	5.2	2
x_4	5.6	50	5.1	0
x_5	-0.5	34	5.3	1
x_6	0.4	19	5.4	2
x_7	1.1	11	5.5	2

$\|x_1 - x_2\|_2 = \sqrt{\sum_{k=1}^4 (x_{1k} - x_{2k})^2}$

$= \sqrt{(x_{11} - x_{21})^2 + (x_{12} - x_{22})^2 + (x_{13} - x_{23})^2 + (x_{14} - x_{24})^2}$

$= \sqrt{(0.2 - 0.4)^2 + (23 - 1)^2 + (5.7 - 5.4)^2 + (A - B)^2}$

LABEL ENCODING

L_2 norm:

$\|x_i - x_j\|_2 = \sqrt{\sum_{k=1}^m (x_{ik} - x_{jk})^2}$ where x_i and x_j are vectors, and there are m dimensions

$A => 0$

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$C => 2$

	X_1	X_2	X_3	X_4
x_1	0.2	23	5.7	0
x_2	0.4	1.	5.4	1
x_3	1.8	0.5	5.2	2
x_4	5.6	50	5.1	0
x_5	-0.5	34	5.3	1
x_6	0.4	19	5.4	2
x_7	1.1	11	5.5	2

$\|x_1 - x_2\|_2 = \sqrt{\sum_{k=1}^4 (x_{1k} - x_{2k})^2}$

$= \sqrt{(x_{11} - x_{21})^2 + (x_{12} - x_{22})^2 + (x_{13} - x_{23})^2 + (x_{14} - x_{24})^2}$

$= \sqrt{(0.2 - 0.4)^2 + (23 - 1)^2 + (5.7 - 5.4)^2 + (0 - 2)^2}$

$= \sqrt{(0.2)^2 + (22)^2 + (0.3)^2 + (-2)^2}$

$= 22.09$

PROBLEM WITH LABEL ENCODING AND DIST

L_2 norm:

$$\|x_i - x_j\|_2 = \sqrt{\sum_{k=1}^m (x_{ik} - x_{jk})^2} \quad \text{where } x_i \text{ and } x_j \text{ are vectors, and there are } m \text{ dimensions}$$

$A \Rightarrow 0$

$B \Rightarrow 1$

$C \Rightarrow 2$

	X_1	X_2	X_3	X_4
x_1	0.2	23	5.7	0
x_2	0.4	1.	5.4	1
x_3	1.8	0.5	5.2	2
x_4	5.6	50	5.1	0
x_5	-0.5	34	5.3	1
x_6	0.4	19	5.4	2
x_7	1.1	11	5.5	2

$D =$

$$\begin{aligned} \|x_1 - x_2\|_2 &= \sqrt{\sum_{k=1}^4 (x_{1k} - x_{2k})^2} \\ &= \sqrt{(x_{11} - x_{21})^2 + (x_{12} - x_{22})^2 + (x_{13} - x_{23})^2 + (x_{14} - x_{24})^2} \\ &= \sqrt{(0.2 - 0.4)^2 + (23 - 1)^2 + (5.7 - 5.4)^2 + (0 - 2)^2} \\ &= \sqrt{(0.2)^2 + (22)^2 + (0.3)^2 + (-2)^2} \\ &= 22.09 \end{aligned}$$

ONE-HOT ENCODING

$\|x_i - x_j\|_2 = \sqrt{\sum_{k=1}^m (x_{ik} - x_{jk})^2}$ where x_i and x_j are vectors, and there are m dimensions

$D =$

x_1

x_2

x_3

x_4

x_5

x_6

x_7

X_1

X_2

X_3

X_4

0.2

23

5.7

A

0.4

1.

5.4

B

1.8

0.5

5.2

C

5.6

50

5.1

A

-0.5

34

5.3

B

0.4

19

5.4

C

1.1

11

5.5

C

$D =$

x_1

x_2

x_3

x_4

x_5

x_6

x_7

X_1

X_2

X_3

X_{4A}

X_{4B}

X_{4C}

0.2

23

5.7

1

0

0

0.4

1.

5.4

0

1

0

1.8

0.5

5.2

0

0

1

5.6

50

5.1

1

0

0

-0.5

34

5.3

0

1

0

0.4

19

5.4

0

0

1

1.1

11

5.5

0

0

1

ONE-HOT ENCODING

$$\|x_i - x_j\|_2 = \sqrt{\sum_{k=1}^m (x_{ik} - x_{jk})^2} \quad \text{where } x_i \text{ and } x_j \text{ are vectors, and there are } m \text{ dimensions}$$

	X_1	X_2	X_3	X_4
x_1	0.2	23	5.7	A
x_2	0.4	1.	5.4	B
x_3	1.8	0.5	5.2	C
x_4	5.6	50	5.1	A
x_5	-0.5	34	5.3	B
x_6	0.4	19	5.4	C
x_7	1.1	11	5.5	C



	X_1	X_2	X_3	X_{4A}	X_{4B}	X_{4C}
x_1	0.2	23	5.7	1	0	0
x_2	0.4	1.	5.4	0	1	0
x_3	1.8	0.5	5.2	0	0	1
x_4	5.6	50	5.1	1	0	0
x_5	-0.5	34	5.3	0	1	0
x_6	0.4	19	5.4	0	0	1
x_7	1.1	11	5.5	0	0	1

ONE-HOT ENCODING

$$\|x_i - x_j\|_2 = \sqrt{\sum_{k=1}^m (x_{ik} - x_{jk})^2} \quad \text{where } x_i \text{ and } x_j \text{ are vectors, and there are } m \text{ dimensions}$$

	X_1	X_2	X_3	X_4		X_1	X_2	X_3	X_{4A}	X_{4B}	X_{4C}
x_1	0.2	23	5.7	A	x_1	0.2	23	5.7	1	0	0
x_2	0.4	1.	5.4	B	x_2	0.4	1.	5.4	0	1	0
x_3	1.8	0.5	5.2	C	x_3	1.8	0.5	5.2	0	0	1
x_4	5.6	50	5.1	A	x_4	5.6	50	5.1	1	0	0
x_5	-0.5	34	5.3	B	x_5	-0.5	34	5.3	0	1	0
x_6	0.4	19	5.4	C	x_6	0.4	19	5.4	0	0	1
x_7	1.1	11	5.5	C	x_7	1.1	11	5.5	0	0	1

ONE-HOT ENCODING

$$\|x_i - x_j\|_2 = \sqrt{\sum_{k=1}^m (x_{ik} - x_{jk})^2} \quad \text{where } x_i \text{ and } x_j \text{ are vectors, and there are } m \text{ dimensions}$$

	X_1	X_2	X_3	X_4	
x_1	0.2	23	5.7	A	\rightarrow
x_2	0.4	1.	5.4	B	
x_3	1.8	0.5	5.2	C	
x_4	5.6	50	5.1	A	
x_5	-0.5	34	5.3	B	
x_6	0.4	19	5.4	C	
x_7	1.1	11	5.5	C	

	X_1	X_2	X_3	X_{4A}	X_{4B}	X_{4C}
x_1	0.2	23	5.7	1	0	0
x_2	0.4	1.	5.4	0	1	0
x_3	1.8	0.5	5.2	0	0	1
x_4	5.6	50	5.1	1	0	0
x_5	-0.5	34	5.3	0	1	0
x_6	0.4	19	5.4	0	0	1
x_7	1.1	11	5.5	0	0	1

$$\begin{aligned}
 \|x_1 - x_2\|_2 &= \sqrt{\sum_{k=1}^6 (x_{1k} - x_{2k})^2} = \sqrt{(x_{11} - x_{21})^2 + (x_{12} - x_{22})^2 + (x_{13} - x_{23})^2 + (x_{14} - x_{24})^2 + (x_{15} - x_{25})^2 + (x_{16} - x_{26})^2} \\
 &= \sqrt{(0.2 - 0.4)^2 + (23 - 1)^2 + (5.7 - 5.4)^2 + (1 - 0)^2 + (0 - 1)^2 + (0 - 0)^2} \\
 &= \sqrt{(0.2)^2 + (22)^2 + (0.3)^2 + (1)^2 + (-1)^2 + (0)^2} = 22.05
 \end{aligned}$$

ONE-HOT ENCODING: DOT PRODUCT

*For one-hot encoded data,
the number of matching categorical values is the dot product of their vectors*

X_4			X_{4A}	X_{4B}	X_{4C}		
$D =$	x_1	A	\rightarrow	x_1	1	0	0
	x_2	B		x_2	0	1	0
	x_3	C		x_3	0	0	1
	x_4	A		x_4	1	0	0
	x_5	B		x_5	0	1	0
	x_6	C		x_6	0	0	1
	x_7	C		x_7	0	0	1

$$x_1 \cdot x_2 = 1 * 0 + 0 * 1 + 0 * 0 = 0$$

ONE-HOT ENCODING: DOT PRODUCT

*For one-hot encoded data,
the number of matching categorical values is the dot product of their vectors*

	X_4		X_{4A}	X_{4B}	X_{4C}
x_1	A		1	0	0
x_2	B		0	1	0
x_3	C		0	0	1
x_4	A		1	0	0
x_5	B		0	1	0
x_6	C		0	0	1
x_7	C		0	0	1

$$x_1 \cdot x_2 = 1 * 0 + 0 * 1 + 0 * 0 = 0$$

$$x_6 \cdot x_7 = 0 * 0 + 0 * 0 + 1 * 1 = 1$$

ONE-HOT ENCODING: DOT PRODUCT

.....

*For one-hot encoded data,
the number of matching categorical values is the dot product of their vectors*

	X_1	X_2
x_1	A	H
x_2	B	L
x_3	C	L
x_4	A	L
x_5	B	H
x_6	C	L
x_7	C	H



	X_{1A}	X_{1B}	X_{1C}	X_{2A}	X_{2B}
x_1	1	0	0	1	0
x_2	0	1	0	0	1
x_3	0	0	1	0	1
x_4	1	0	0	0	1
x_5	0	1	0	1	0
x_6	0	0	1	0	1
x_7	0	0	1	1	0

$$x_1 \cdot x_2 = 1 * 0 + 0 * 1 + 0 * 0 + 1 * 0 + 0 * 1 = 0$$

ONE-HOT ENCODING: DOT PRODUCT

.....
*For one-hot encoded data,
the number of matching categorical values is the dot product of their vectors*

	X_1	X_2		X_{1A}	X_{1B}	X_{1C}	X_{2A}	X_{2B}
x_1	A	H		1	0	0	1	0
x_2	B	L		0	1	0	0	1
x_3	C	L		0	0	1	0	1
$D = x_4$	A	L	→	1	0	0	0	1
x_5	B	H		0	1	0	1	0
x_6	C	L		0	0	1	0	1
x_7	C	H		0	0	1	1	0

$$x_1 \cdot x_2 = 1 * 0 + 0 * 1 + 0 * 0 + 1 * 0 + 0 * 1 = 0$$

$$x_2 \cdot x_3 = 0 * 0 + 1 * 0 + 0 * 1 + 0 * 0 + 1 * 1 = 1$$

ONE-HOT ENCODING: DOT PRODUCT

.....

*For one-hot encoded data,
the number of matching categorical values is the dot product of their vectors*

	X_1	X_2		X_{1A}	X_{1B}	X_{1C}	X_{2A}	X_{2B}
x_1	A	H		1	0	0	1	0
x_2	B	L		0	1	0	0	1
x_3	C	L		0	0	1	0	1
x_4	A	L		1	0	0	0	1
x_5	B	H		0	1	0	1	0
x_6	C	L		0	0	1	0	1
x_7	C	H		0	0	1	1	0

$$x_1 \cdot x_2 = 1 * 0 + 0 * 1 + 0 * 0 + 1 * 0 + 0 * 1 = 0$$

$$x_2 \cdot x_3 = 0 * 0 + 1 * 0 + 0 * 1 + 0 * 0 + 1 * 1 = 1$$

$$x_3 \cdot x_6 = 0 * 0 + 0 * 0 + 1 * 1 + 0 * 0 + 1 * 1 = 2$$

ONE-HOT ENCODING: HAMMING DISTANCE

Hamming Distance,
the number of mismatches between two vectors

$$\delta_H(x_i, x_j) = \text{sum}(x_i \oplus x_j)$$

ONE-HOT ENCODING: HAMMING DISTANCE

Hamming Distance,
the number of mismatches between two vectors

$$\delta_H(x_i, x_j) = \text{sum}(x_i \oplus x_j)$$

Recall XOR \oplus

<i>a</i>	<i>b</i>	<i>a</i> \oplus <i>b</i>
0	0	0
0	1	1
1	0	1
1	1	0

ONE-HOT ENCODING: HAMMING DISTANCE

.....

Hamming Distance,
the number of mismatches between two vectors

$$\delta_H(x_i, x_j) = \text{sum}(x_i \oplus x_j)$$

Recall XOR \oplus

<i>a</i>	<i>b</i>	<i>a</i> \oplus <i>b</i>
0	0	0
0	1	1
1	0	1
1	1	0

D =

	<i>X</i> ₁	<i>X</i> ₂
<i>x</i> ₁	A	H
<i>x</i> ₂	B	L
<i>x</i> ₃	C	L
<i>x</i> ₄	A	L
<i>x</i> ₅	B	H
<i>x</i> ₆	C	L
<i>x</i> ₇	C	H



D =

	<i>X</i> _{1A}	<i>X</i> _{1B}	<i>X</i> _{1C}	<i>X</i> _{2A}	<i>X</i> _{2B}
<i>x</i> ₁	1	0	0	1	0
<i>x</i> ₂	0	1	0	0	1
<i>x</i> ₃	0	0	1	0	1
<i>x</i> ₄	1	0	0	0	1
<i>x</i> ₅	0	1	0	1	0
<i>x</i> ₆	0	0	1	0	1
<i>x</i> ₇	0	0	1	1	0

$$\delta_H(x_1, x_2)$$

ONE-HOT ENCODING: HAMMING DISTANCE

Hamming Distance,
the number of mismatches between two vectors

$$\delta_H(x_i, x_j) = \text{sum}(x_i \oplus x_j)$$

Recall XOR \oplus

a	b	$a \oplus b$
0	0	0
0	1	1
1	0	1
1	1	0

	X_1	X_2
x_1	A	H
x_2	B	L
x_3	C	L
x_4	A	L
x_5	B	H
x_6	C	L
x_7	C	H



	X_{1A}	X_{1B}	X_{1C}	X_{2A}	X_{2B}
x_1	1	0	0	1	0
x_2	0	1	0	0	1
x_3	0	0	1	0	1
x_4	1	0	0	0	1
x_5	0	1	0	1	0
x_6	0	0	1	0	1
x_7	0	0	1	1	0

$$\delta_H(x_1, x_2) = \text{sum}(x_1 \oplus x_2)$$

ONE-HOT ENCODING: HAMMING DISTANCE

Hamming Distance,
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$$\delta_H(x_i, x_j) = \text{sum}(x_i \oplus x_j)$$

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$D =$

	X_1	X_2
x_1	A	H
x_2	B	L
x_3	C	L
x_4	A	L
x_5	B	H
x_6	C	L
x_7	C	H



$D =$

	X_{1A}	X_{1B}	X_{1C}	X_{2A}	X_{2B}
x_1	1	0	0	1	0
x_2	0	1	0	0	1
x_3	0	0	1	0	1
x_4	1	0	0	0	1
x_5	0	1	0	1	0
x_6	0	0	1	0	1
x_7	0	0	1	1	0

$$\delta_H(x_1, x_2) = \text{sum}(x_1 \oplus x_2) = (1 \oplus 0) + (0 \oplus 1) + (0 \oplus 0) + (1 \oplus 0) + (0 \oplus 1)$$

ONE-HOT ENCODING: HAMMING DISTANCE

Hamming Distance,
the number of mismatches between two vectors

$$\delta_H(x_i, x_j) = \text{sum}(x_i \oplus x_j)$$

Recall XOR \oplus

a	b	$a \oplus b$
0	0	0
0	1	1
1	0	1
1	1	0

$D =$

	X_1	X_2
x_1	A	H
x_2	B	L
x_3	C	L
x_4	A	L
x_5	B	H
x_6	C	L
x_7	C	H



$D =$

	X_{1A}	X_{1B}	X_{1C}	X_{2A}	X_{2B}
x_1	1	0	0	1	0
x_2	0	1	0	0	1
x_3	0	0	1	0	1
x_4	1	0	0	0	1
x_5	0	1	0	1	0
x_6	0	0	1	0	1
x_7	0	0	1	1	0

$$\delta_H(x_1, x_2) = \text{sum}(x_1 \oplus x_2) = (1 \oplus 0) + (0 \oplus 1) + (0 \oplus 0) + (1 \oplus 0) + (0 \oplus 1) = 1 + 1 + 0 + 1 + 1 = 4$$

ONE-HOT ENCODING: HAMMING DISTANCE

Hamming Distance,
the number of mismatches between two vectors

$$\delta_H(x_i, x_j) = \text{sum}(x_i \oplus x_j)$$

Recall XOR \oplus

a	b	$a \oplus b$
0	0	0
0	1	1
1	0	1
1	1	0

$D =$

	X_1	X_2
x_1	A	H
x_2	B	L
x_3	C	L
x_4	A	L
x_5	B	H
x_6	C	L
x_7	C	H



$D =$

	X_{1A}	X_{1B}	X_{1C}	X_{2A}	X_{2B}
x_1	1	0	0	1	0
x_2	0	1	0	0	1
x_3	0	0	1	0	1
x_4	1	0	0	0	1
x_5	0	1	0	1	0
x_6	0	0	1	0	1
x_7	0	0	1	1	0

$$\delta_H(x_1, x_2) = \text{sum}(x_1 \oplus x_2) = (1 \oplus 0) + (0 \oplus 1) + (0 \oplus 0) + (1 \oplus 0) + (0 \oplus 1) = 1 + 1 + 0 + 1 + 1 = 4$$

$$\delta_H(x_2, x_3) = ??$$

$$\delta_H(x_3, x_6) = ??$$

ONE-HOT ENCODING: HAMMING DISTANCE

Hamming Distance,
the number of mismatches between two vectors

$$\delta_H(x_i, x_j) = \text{sum}(x_i \oplus x_j)$$

Recall XOR \oplus

a	b	$a \oplus b$
0	0	0
0	1	1
1	0	1
1	1	0

$D =$

	X_1	X_2
x_1	A	H
x_2	B	L
x_3	C	L
x_4	A	L
x_5	B	H
x_6	C	L
x_7	C	H



$D =$

	X_{1A}	X_{1B}	X_{1C}	X_{2A}	X_{2B}
x_1	1	0	0	1	0
x_2	0	1	0	0	1
x_3	0	0	1	0	1
x_4	1	0	0	0	1
x_5	0	1	0	1	0
x_6	0	0	1	0	1
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$$\delta_H(x_2, x_3) = 0 \oplus 0 + 1 \oplus 0 + 0 \oplus 1 + 0 \oplus 0 + 1 \oplus 1 = 2$$

$$\delta_H(x_3, x_6) = ??$$

ONE-HOT ENCODING: HAMMING DISTANCE

Hamming Distance,
the number of mismatches between two vectors

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x_6	C	L
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$D =$

	X_{1A}	X_{1B}	X_{1C}	X_{2A}	X_{2B}
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x_6	0	0	1	0	1
x_7	0	0	1	1	0

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$$\delta_H(x_2, x_3) = 0 \oplus 0 + 1 \oplus 0 + 0 \oplus 1 + 0 \oplus 0 + 1 \oplus 1 = 2$$

$$\delta_H(x_3, x_6) = ??$$

ONE-HOT ENCODING: HAMMING DISTANCE

Hamming Distance,
the number of mismatches between two vectors

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$$\delta_H(x_3, x_6) = 0 \oplus 0 + 0 \oplus 0 + 1 \oplus 1 + 0 \oplus 0 + 1 \oplus 1 = 0$$

SET COMPARISON: JACCARD SIMILARITY

Jaccard Similarity,

the size of the intersection over the size of the union

$$J(x_i, x_j) = \frac{|x_i \cap x_j|}{|x_i \cup x_j|} = \frac{\text{sum}(x_i \wedge x_j)}{\text{sum}(x_i \vee x_j)}$$

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	X_1	X_2	X_3	X_4	X_5
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	X_1	X_2	X_3	X_4	X_5
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$$J(x_2, x_3) = ??$$

$$J(x_3, x_6) = ??$$

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$$J(x_i, x_j) = \frac{|x_i \cap x_j|}{|x_i \cup x_j|} = \frac{\text{sum}(x_i \wedge x_j)}{\text{sum}(x_i \vee x_j)}$$

$D =$

	X_1	X_2	X_3	X_4	X_5
x_1	1	0	0	1	0
x_2	0	1	0	0	1
x_3	0	0	1	0	1
x_4	1	0	0	0	1
x_5	0	1	0	1	0
x_6	0	0	1	0	1
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	X_1	X_2	X_3	X_4	X_5
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