

Probability and Random Processes

ECS 315

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11 Multiple Random Variables



Office Hours:

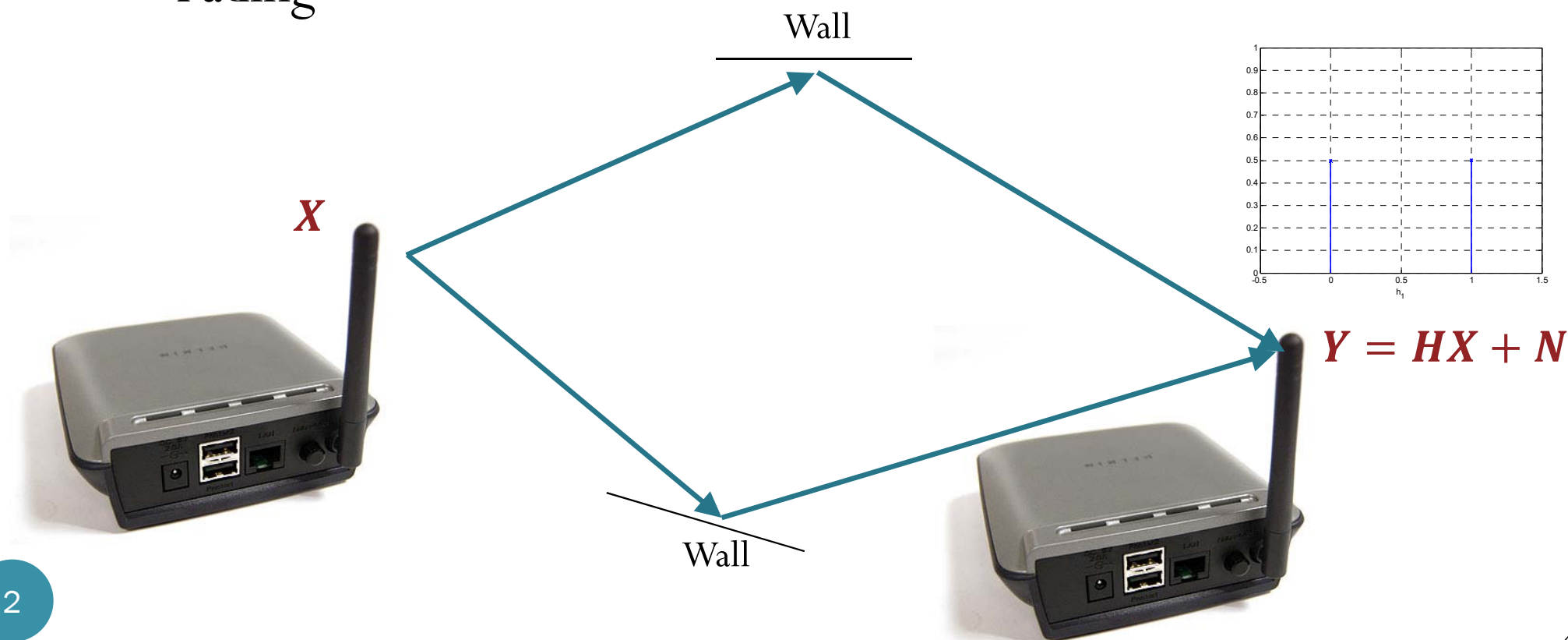
Check Google Calendar on the course website.

Dr.Prapun's Office:

6th floor of Sirindhralai building,
BKD

SISO Wireless Communications

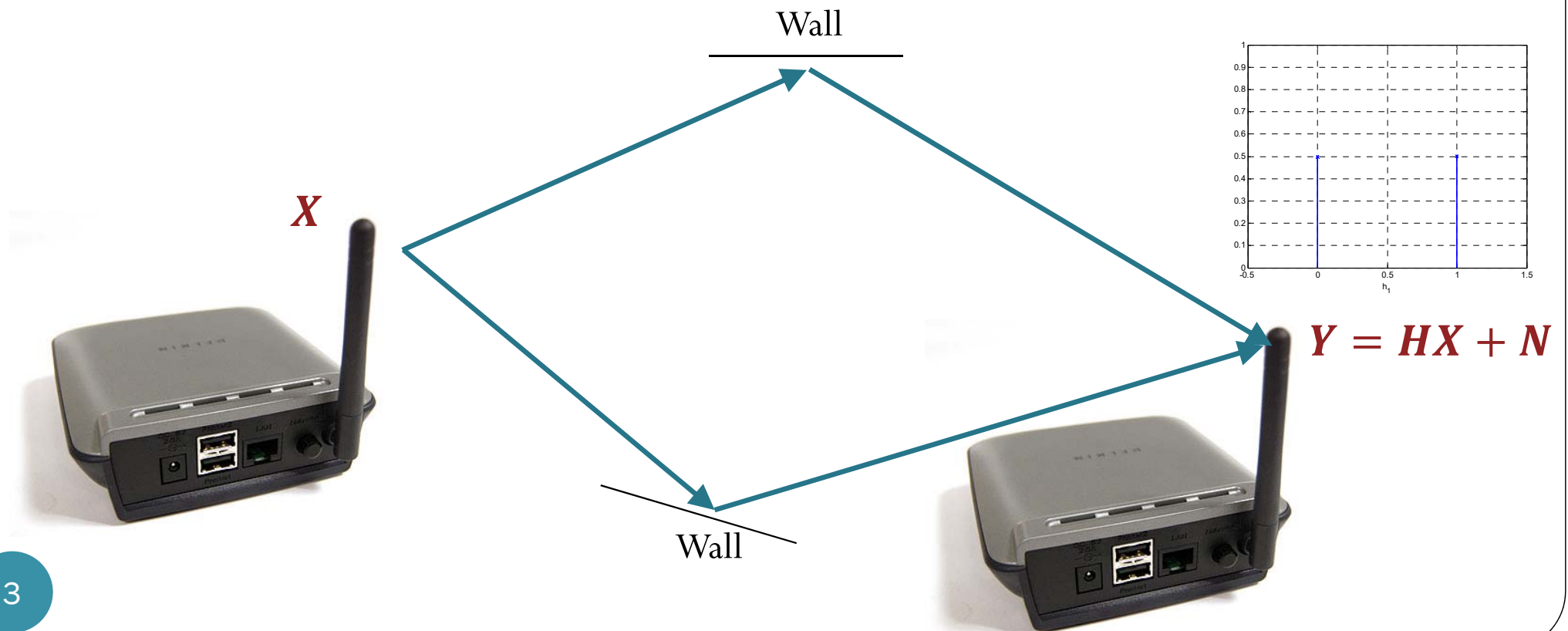
- Multipath propagation
- At the receiver, multiple copies of the signal may be combined constructively or destructively.
- Fading



SISO Wireless Communications

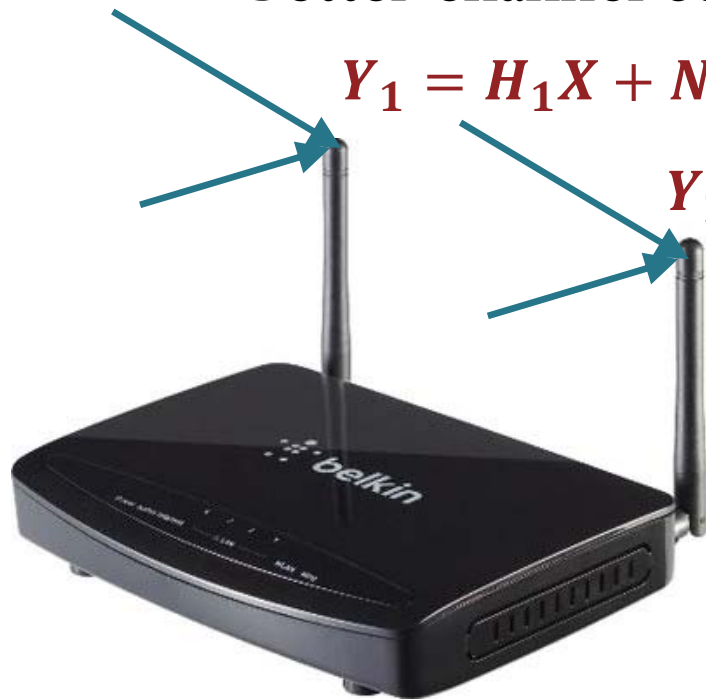
- H = channel coefficient (quality)
- For simplicity, let's assume two possible values for H : good (1) or bad (0).

Its value is random. $H: 0 \ 1 \ 0 \ 0 \ 1 \ 1 \ 1 \ 1 \ 0 \ 1 \dots$



MIMO Wireless Communications

- Here, there are **two** antennas to receive the signals
- Use the antenna that receive stronger signal (less fading; better channel condition)

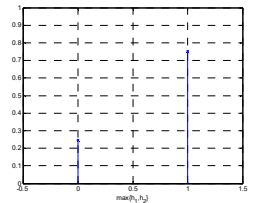
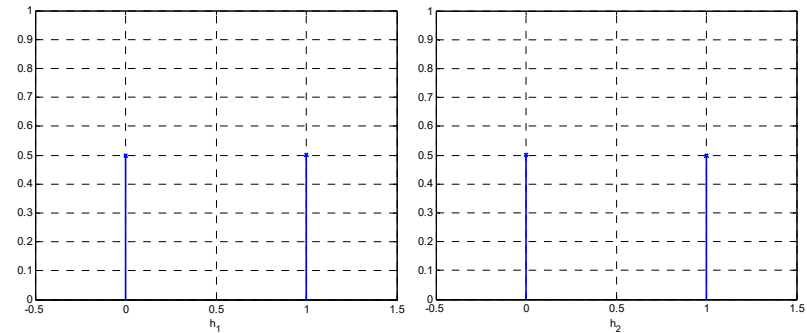


$$Y_1 = H_1X + N_1$$

$$Y_2 = H_2X + N_2$$

H_1 :	0	1	0	0	1	1	1	1	0	1	...
H_2 :	0	1	1	0	0	0	0	1	0	1	...

H_{used} :	0	1	1	0	1	1	1	1	0	1	...
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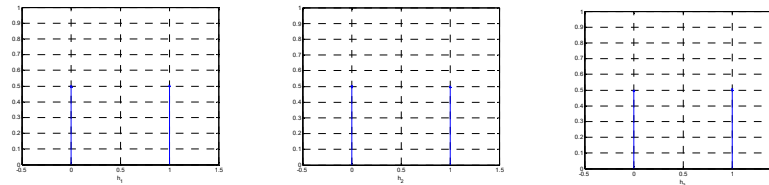
MIMO Wireless Communications

- Here, there are **three** antennas to receive the signals
- Use the antenna that receive the strongest signal (least fading; best channel condition)

$$Y_1 = H_1X + N_1$$

$$Y_2 = H_2X + N_2$$

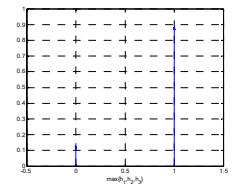
$$Y_3 = H_3X + N_3$$



$H_1:$	0	1	0	0	1	1	1	1	0	1 ...
$H_2:$	0	1	1	0	0	0	0	1	0	1 ...
$H_3:$	0	1	1	1	1	0	1	0	1	1 ...

↓

$H_{\text{used}}:$	0	1	1	1	1	1	1	1	1	1 ...
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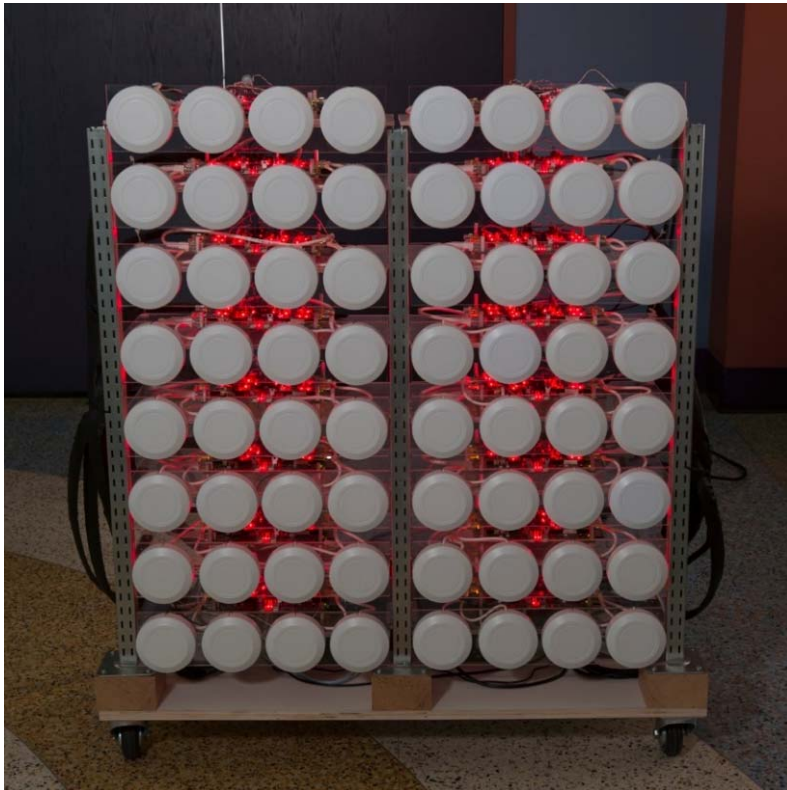
MIMO Communications

- Of course, even more antennas is also possible.

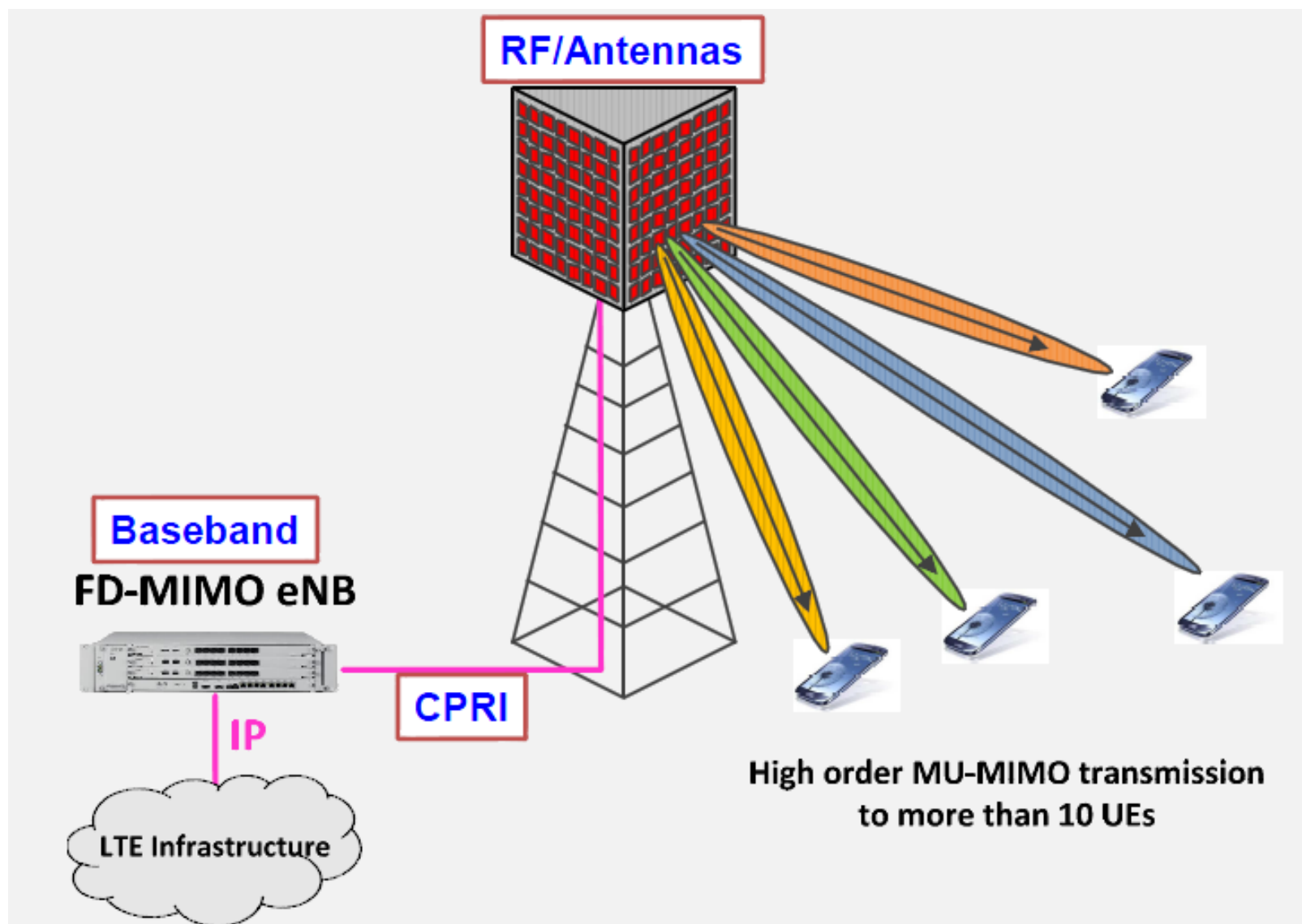


Very Large MIMO Systems

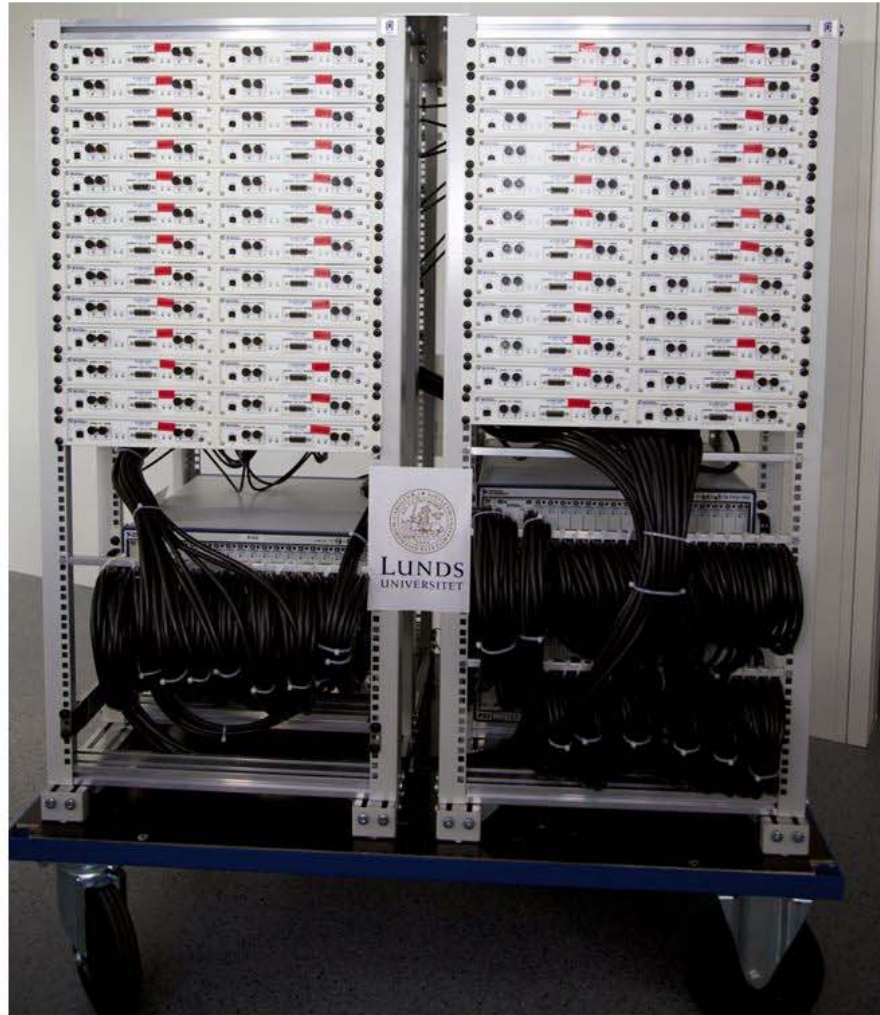
- “antenna array”



Very Large MIMO Systems



Very Large MIMO Systems



Very Large MIMO Systems



Chapter 6 vs. Chapter 11

Joint probability

$$P(A \cap B)$$

Joint event

Conditional probability

$$P(A|B) = \frac{P(A \cap B)}{P(B)}$$
$$= \frac{P(B|A)P(A)}{P(B)}$$

Joint pmf

$$p_{X,Y}(x, y) = P[X = x, Y = y]$$

$$A = [X = x]$$

$$B = [Y = y]$$

Conditional pmf

$$p_{X|Y}(x|y) = \frac{p_{X,Y}(x, y)}{p_Y(y)}$$
$$= \frac{p_{Y|X}(y|x)p_X(x)}{p_Y(y)}$$

Events A and B are **independent**:

$$P(A \cap B) = P(A)P(B)$$

RVs X and Y are **independent**:

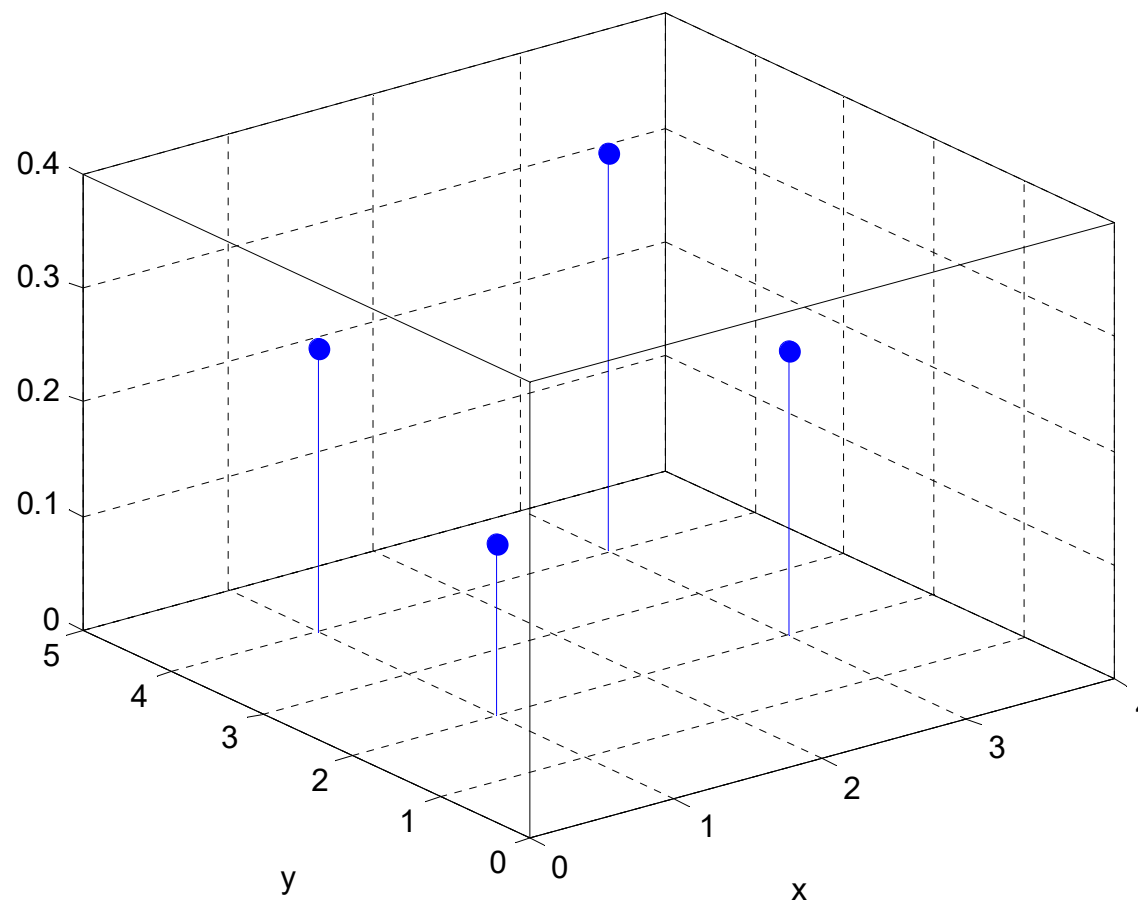
$$p_{X,Y}(x, y) = p_X(x)p_Y(y) \text{ for any } x \text{ and } y$$

Example: small joint pmf matrix Ex. 11.7

```
close all; clear all;  
x = [1 3];  
y = [2 4];  
PXY = [3/20 5/20; 5/20 7/20];
```

```
[X Y] = meshgrid(x,y);  
X = X. '; Y = Y. ';
```

```
stem3(X,Y,PXY,'filled')  
xlim([0,4])  
ylim([0,5])  
xlabel('x')  
ylabel('y')
```



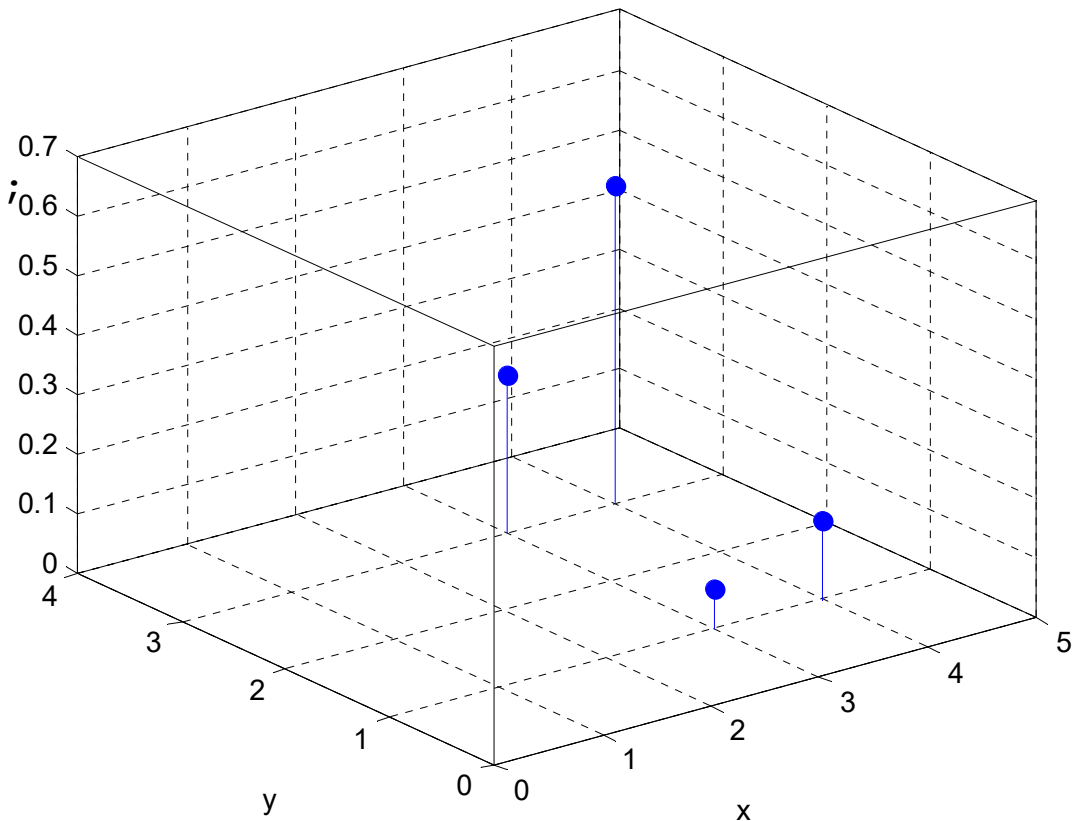
(More)

Example: small joint pmf matrix Ex. 11.26

```
close all; clear all;  
x = [3 4];  
y = [1 3];  
PXY = [1/15 4/15; 2/15 8/15];
```

```
[X Y] = meshgrid(x,y);  
X = X. '; Y = Y. ';
```

```
stem3(X,Y,PXY,'filled')  
xlim([0,5])  
ylim([0,4])  
xlabel('x')  
ylabel('y')
```



Example: large joint pmf matrix

```
close all; clear all;  
n = 10; p = 3/5;  
x = 0:n;  
y = 0:n;
```

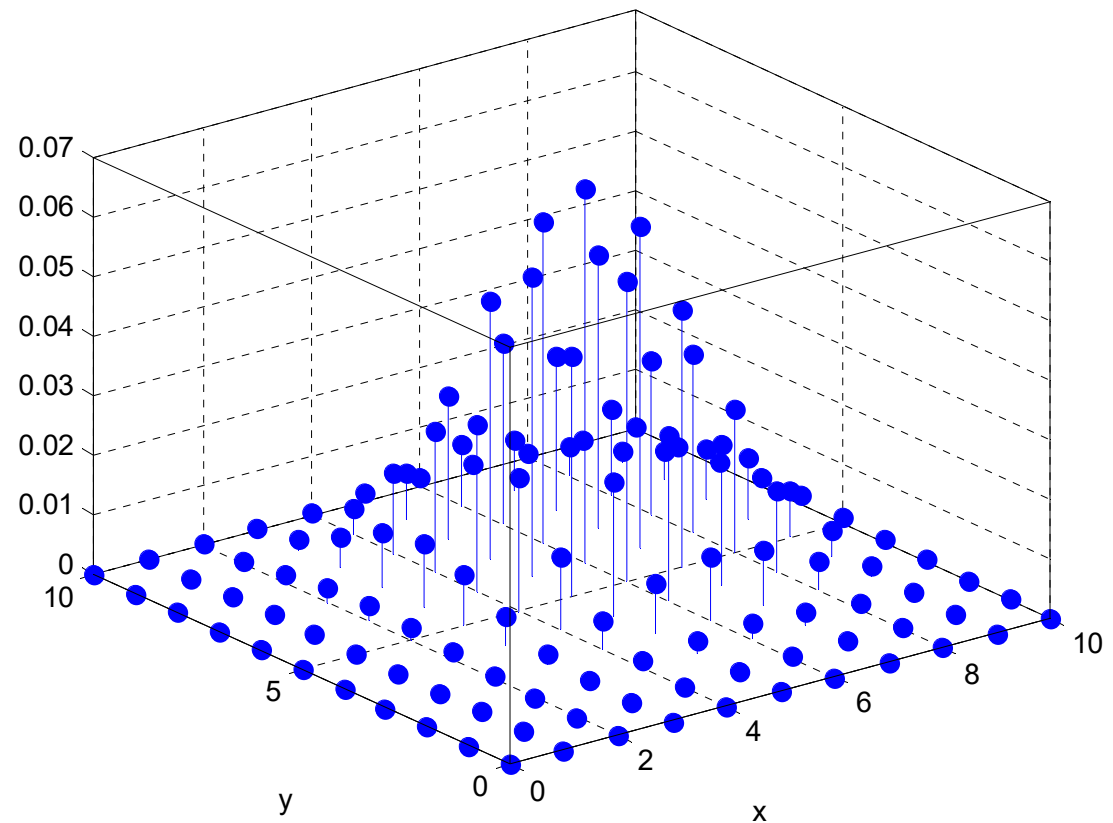
```
pX = binopdf(x,n,p);  
pY = binopdf(y,n,p);
```

```
PXY = pX.'*pY;
```

```
[X Y] = meshgrid(x,y);  
X = X. '; Y = Y. ';
```

```
stem3(X,Y,PXY, 'filled')  
%mesh(X,Y,PXY)  
%surf(X,Y,PXY)
```

```
xlabel('x')  
ylabel('y')
```



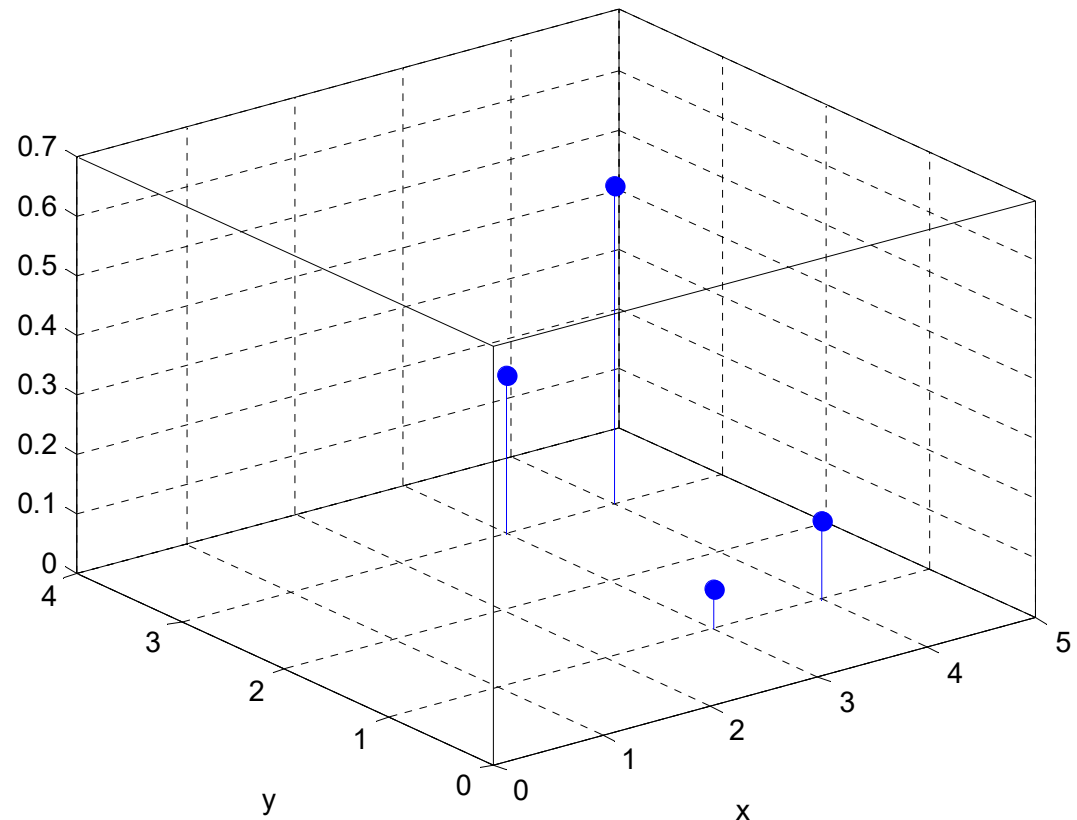
Example: small joint pmf matrix Ex. 11.29

$$P_{X,Y} = \begin{array}{c|cc} x \backslash y & 1 & 3 \\ \hline 3 & \frac{1}{15} & \frac{4}{15} \\ 4 & \frac{2}{15} & \frac{8}{15} \end{array}$$

```
close all; clear all;  
x = [3 4];  
y = [1 3];  
PXY = [1/15 4/15; 2/15 8/15];
```

```
[X Y] = meshgrid(x,y);  
X = X. '; Y = Y. ';
```

```
stem3(X,Y,PXY,'filled')  
xlim([0,5])  
ylim([0,4])  
xlabel('x')  
ylabel('y')
```



Evaluation of Probability

- Consider two random variables X and Y .
- Suppose their **joint pmf matrix** is

	y	2	3	4	5	6
x	1	0.1	0.1	0	0	0
$P_{X,Y}$	3	0.1	0	0	0.1	0
	4	0	0.1	0.2	0	0
	6	0	0	0	0	0.3

$$p_{X,Y}(4,3) = P[X = 4 \text{ and } Y = 3] = 0.1$$

and



Evaluation of Probability

- Consider two random variables X and Y .
- Suppose their **joint pmf matrix** is

$$P_{X,Y} = \begin{array}{c|ccccc} & \begin{array}{c} y \\ \hline 2 \quad 3 \quad 4 \quad 5 \quad 6 \end{array} \\ \begin{array}{c} x \\ \hline 1 \\ 3 \\ 4 \\ 6 \end{array} & \begin{bmatrix} 0.1 & 0.1 & 0 & 0 & 0 \\ 0.1 & 0 & 0 & 0.1 & 0 \\ 0 & 0.1 & 0.2 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0.3 \end{bmatrix} \end{array}$$

- Find $P[X + Y < 7]$



Evaluation of Probability

- Consider two random variables X and Y .
- Suppose their **joint pmf matrix** is

$$P_{X,Y} = \begin{array}{c|ccccc} & \begin{array}{c} y \\ \hline 2 \quad 3 \quad 4 \quad 5 \quad 6 \end{array} \\ \begin{array}{c} x \\ \hline 1 \\ 3 \\ 4 \\ 6 \end{array} & \begin{bmatrix} 0.1 & 0.1 & 0 & 0 & 0 \\ 0.1 & 0 & 0 & 0.1 & 0 \\ 0 & 0.1 & 0.2 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0.3 \end{bmatrix} \end{array}$$

- Find $P[X + Y < 7]$

Step 1: Find the pairs (x,y) that satisfy the condition “ $x+y < 7$ ”

One way to do this is to first construct the matrix of $x+y$.

$$x + y = \begin{array}{c|ccccc} & \begin{array}{c} y \\ \hline 2 \quad 3 \quad 4 \quad 5 \quad 6 \end{array} \\ \begin{array}{c} x \\ \hline 1 \\ 3 \\ 4 \\ 6 \end{array} & \begin{bmatrix} 3 & 4 & 5 & 6 & 7 \\ 5 & 6 & 7 & 8 & 9 \\ 6 & 7 & 8 & 9 & 10 \\ 8 & 9 & 10 & 11 & 12 \end{bmatrix} \end{array}$$


Evaluation of Probability

- Consider two random variables X and Y .
- Suppose their **joint pmf matrix** is

$$P_{X,Y} = \begin{array}{c|ccccc} & y & 2 & 3 & 4 & 5 & 6 \\ \hline x & 1 & 0.1 & 0.1 & 0 & 0 & 0 \\ & 3 & 0.1 & 0 & 0 & 0.1 & 0 \\ & 4 & 0 & 0.1 & 0.2 & 0 & 0 \\ & 6 & 0 & 0 & 0 & 0 & 0.3 \end{array}$$

- Find $P[X + Y < 7]$

Step 2: Add the corresponding probabilities from the joint pmf (matrix)

$$P[X + Y < 7] = 0.1 + 0.1 + 0.1 = 0.3$$

$$x + y = \begin{array}{c|ccccc} & y & 2 & 3 & 4 & 5 & 6 \\ \hline x & 1 & 3 & 4 & 5 & 6 & 7 \\ & 3 & 5 & 6 & 7 & 8 & 9 \\ & 4 & 6 & 7 & 8 & 9 & 10 \\ & 6 & 8 & 9 & 10 & 11 & 12 \end{array}$$


Joint pmf matrix for independent RVs

Command Window

```
>> pX = [1/3 2/3]
pX =
    0.3333    0.6667
>> pY = [1/5 4/5]
pY =
    0.2000    0.8000
>> sym(pX' * pY)
ans =
 [ 1/15, 4/15]
 [ 2/15, 8/15]
>>
```



Joint pmf for two i.i.d. RVs

```
close all; clear all;  
n = 10; p = 3/5;  
x = 0:n;  
y = 0:n;
```

```
pX = binopdf(x,n,p);  
pY = binopdf(y,n,p);
```

```
PXY = pX.'*pY;
```

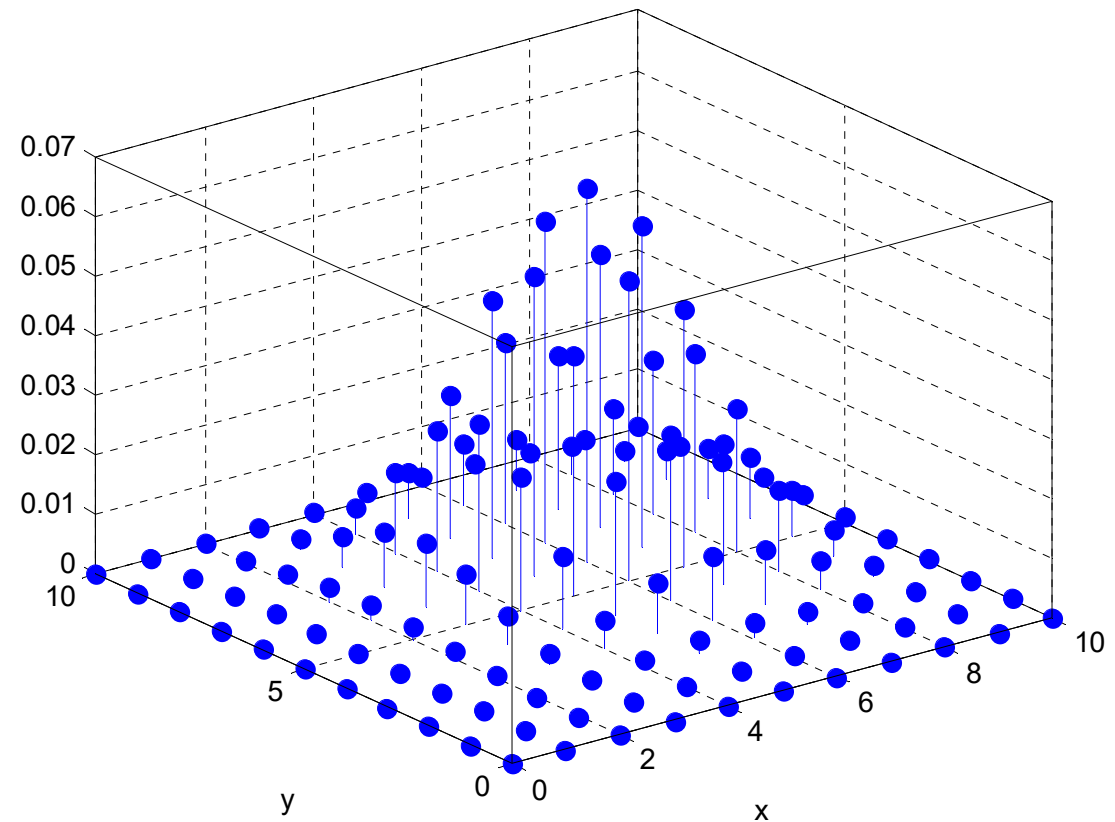
Note how the pmfs
are multiplied because
of the independence.

```
[X Y] = meshgrid(x,y);  
X = X. '; Y = Y. ';
```

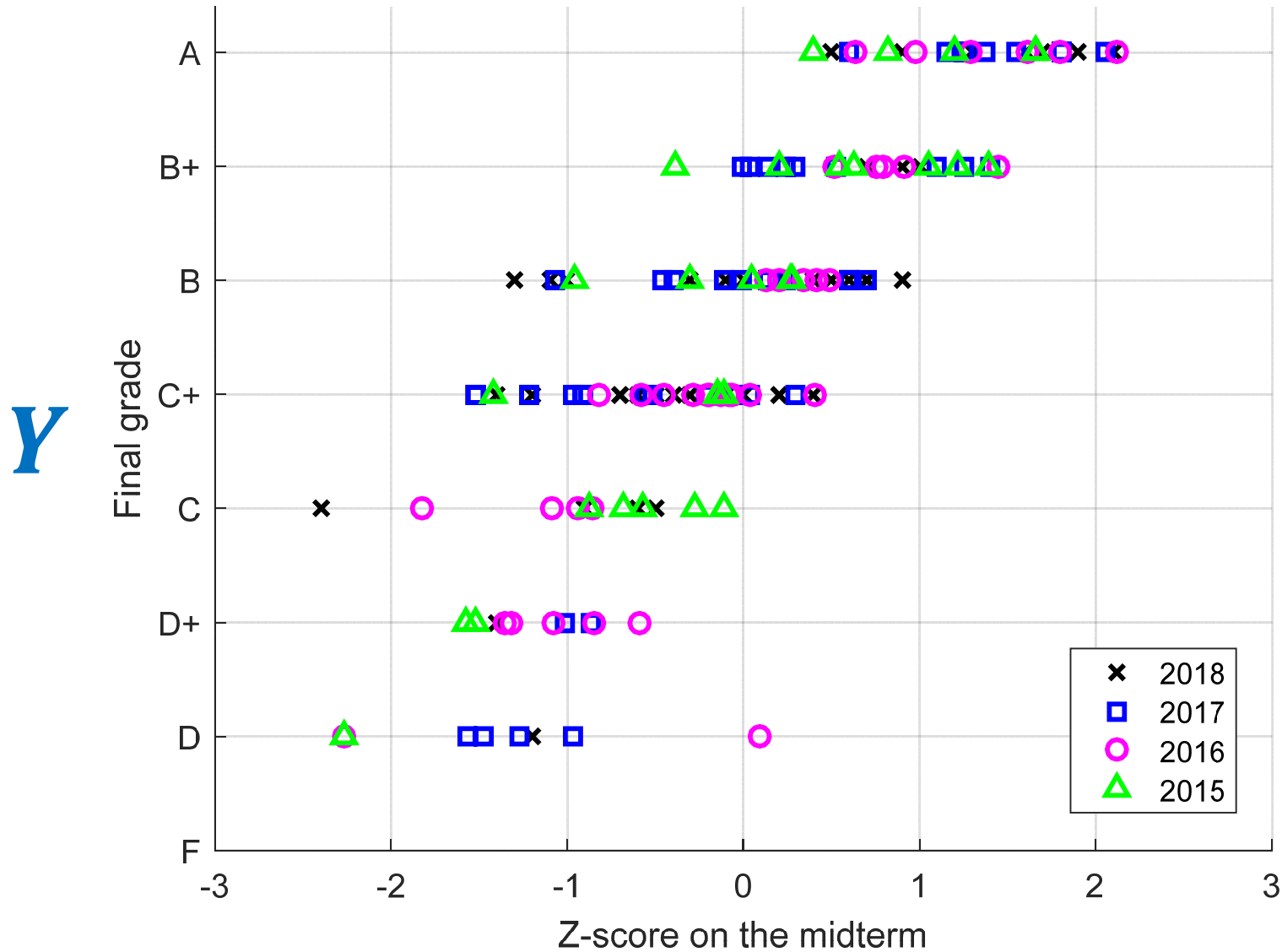
```
stem3(X,Y,PXY, 'filled')  
%mesh(X,Y,PXY)  
%surf(X,Y,PXY)
```

```
xlabel('x')  
ylabel('y')
```

$$X, Y \stackrel{\text{i.i.d.}}{\sim} \mathcal{B}\left(10, \frac{3}{5}\right)$$



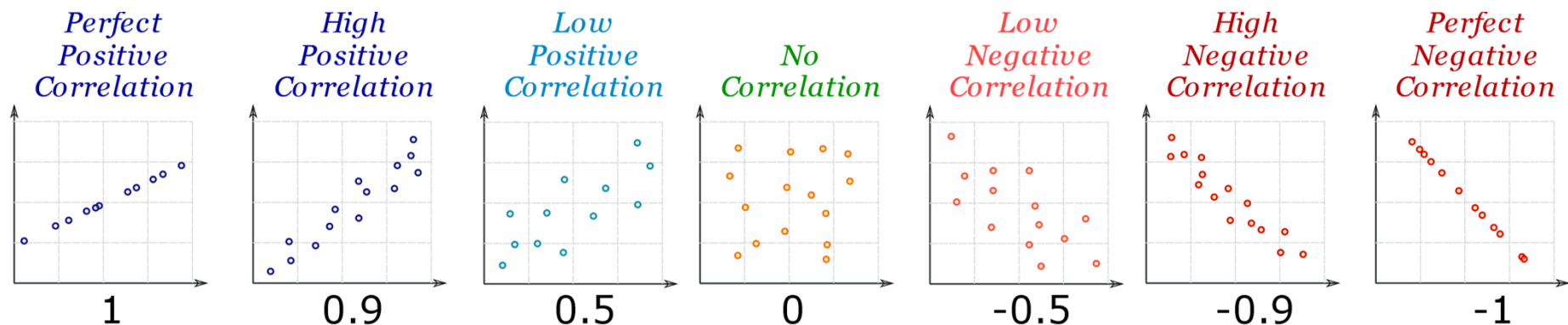
Dependency



“Correlation”

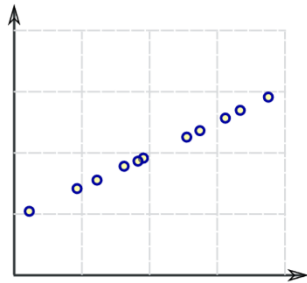
Actually, this is “correlation coefficient”

- Correlation **measures a specific kind of dependency**.
 - Dependence = statistical relationship between two random variables (or two sets of data).
 - Correlation measures “**linear**” relationship between two random variables.



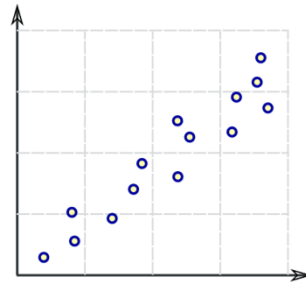
Correlation Coefficients

*Perfect
Positive
Correlation*



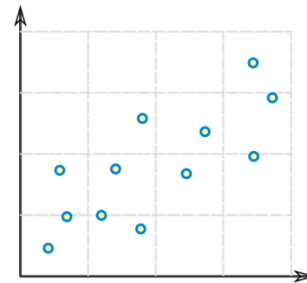
1

*High
Positive
Correlation*



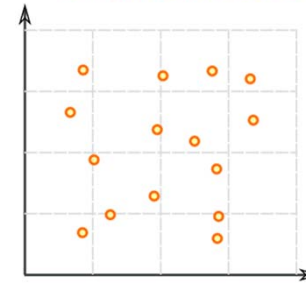
0.9

*Low
Positive
Correlation*



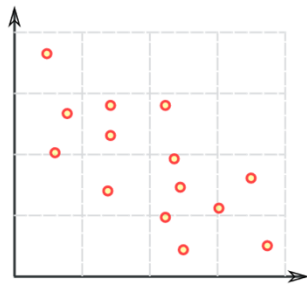
0.5

*No
Correlation*



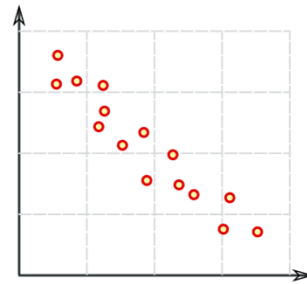
0

*Low
Negative
Correlation*



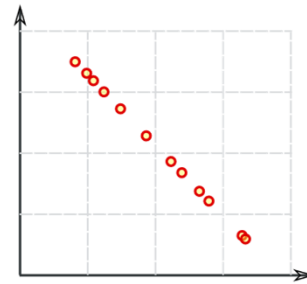
-0.5

*High
Negative
Correlation*



-0.9

*Perfect
Negative
Correlation*



-1



Correlation

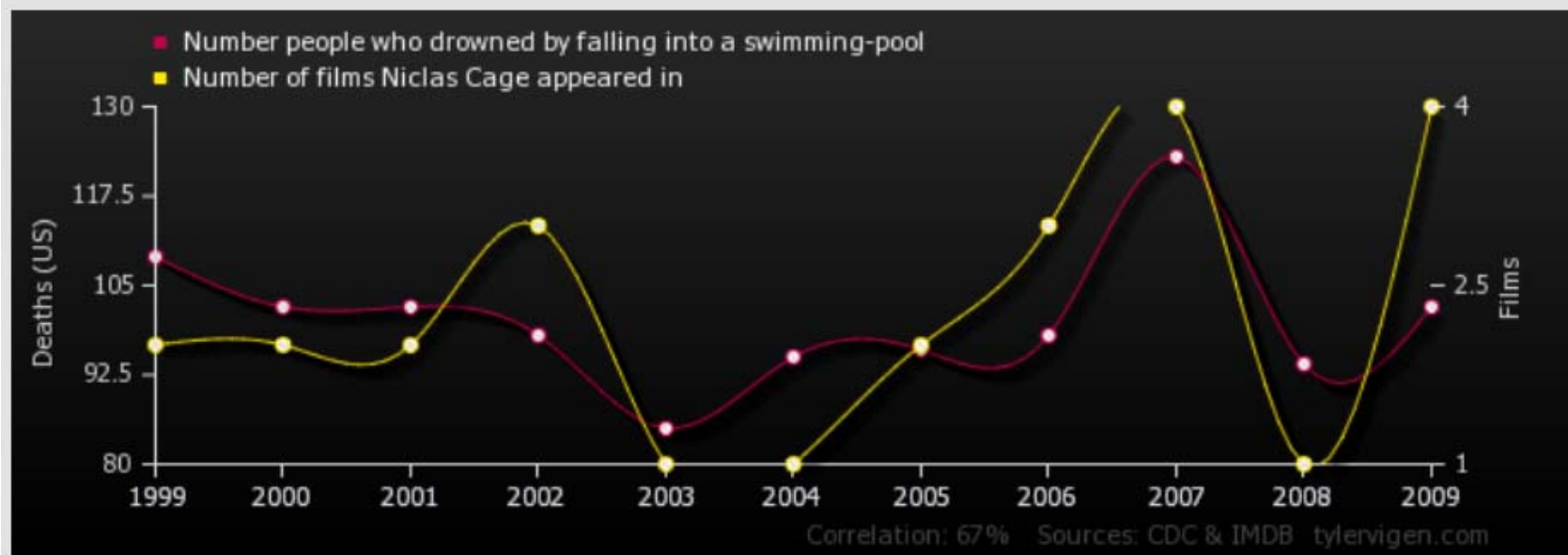
Actually, this is “correlation coefficient”

- Correlation measures a specific kind of dependency.
 - Dependence = statistical relationship between two random variables (or two sets of data).
 - Correlation measures “linear” relationship between two random variables.
- Correlation and causality.
 - “Correlation does not imply causation”
 - Correlation cannot be used to infer a causal relationship between the variables.

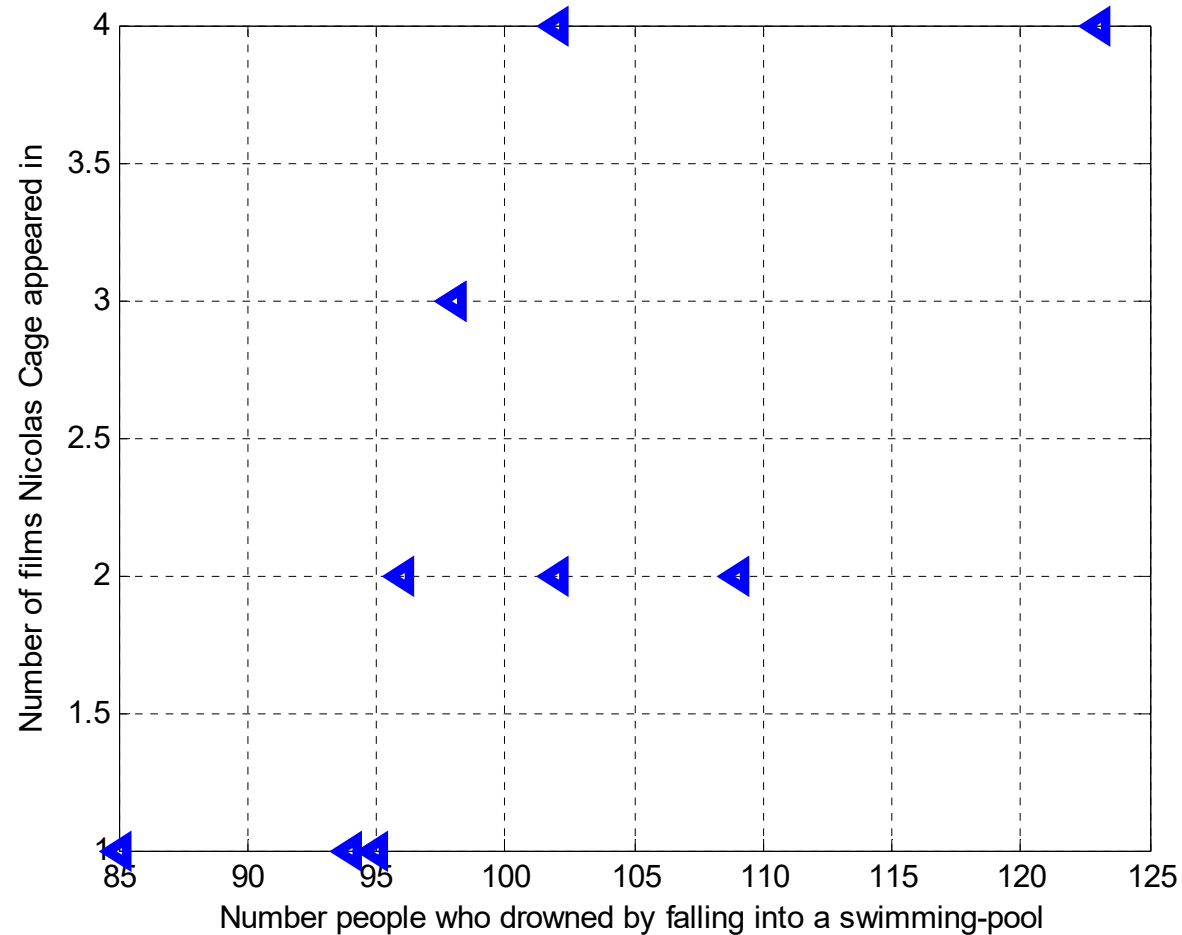


Two “Unrelated” Events

Number people who drowned by falling into a swimming-pool
correlates with
Number of films Nicolas Cage appeared in



Two “Unrelated” Events

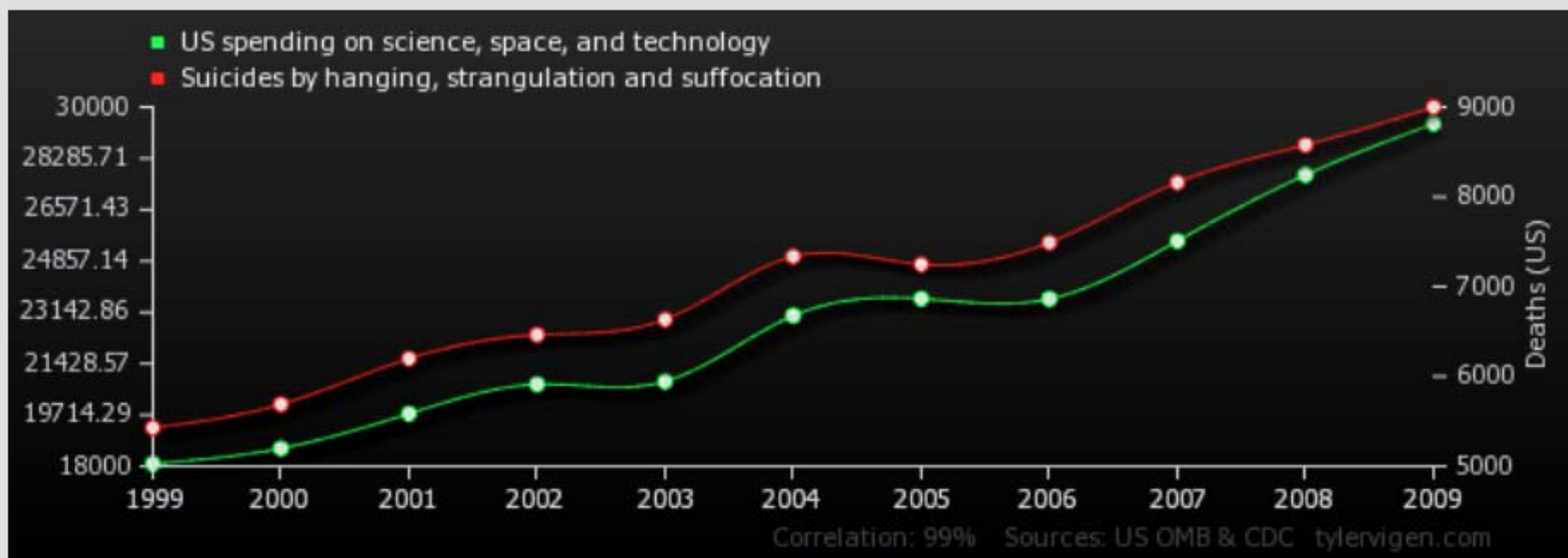


Correlation: 0.666004

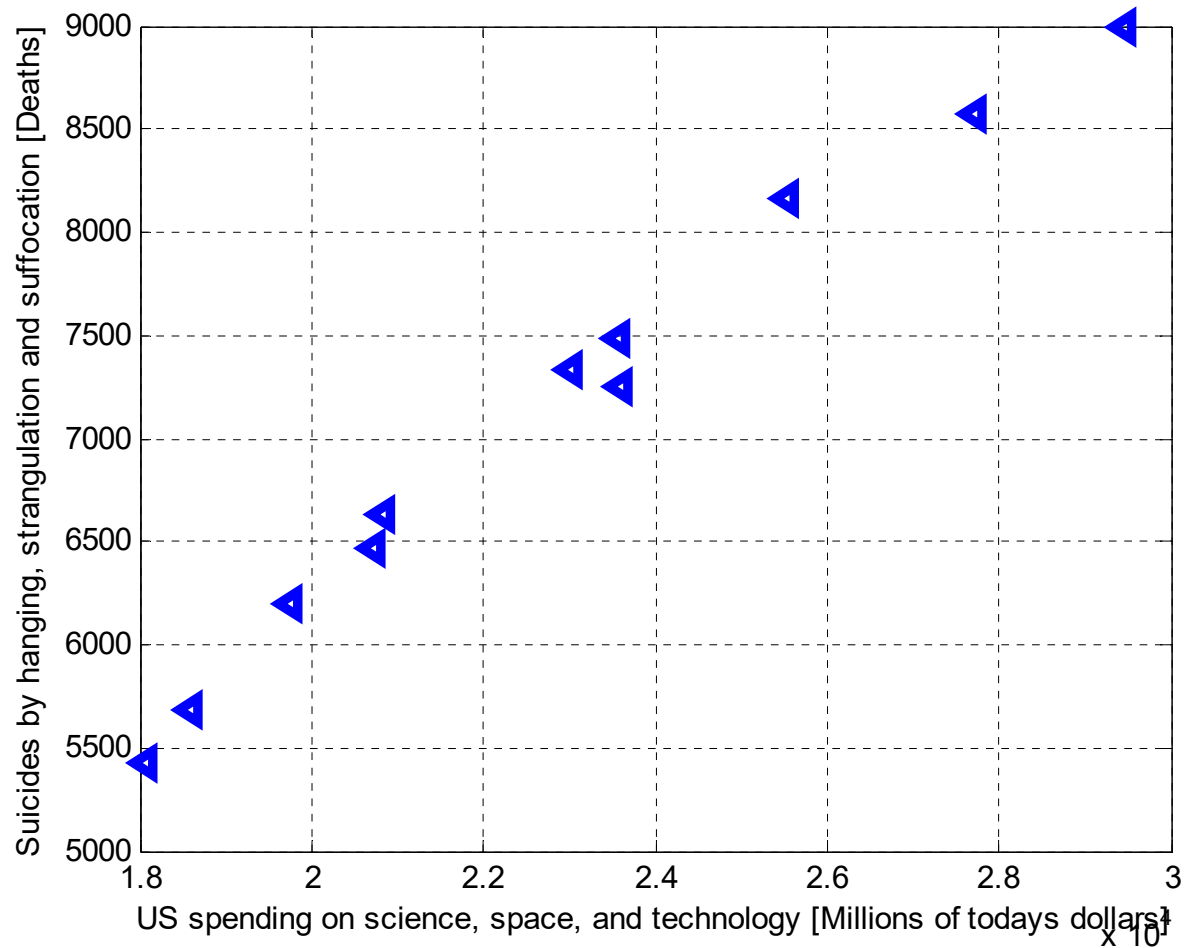
<http://www.tylervigen.com/>

Spurious Correlation

US spending on science, space, and technology
correlates with
Suicides by hanging, strangulation and suffocation



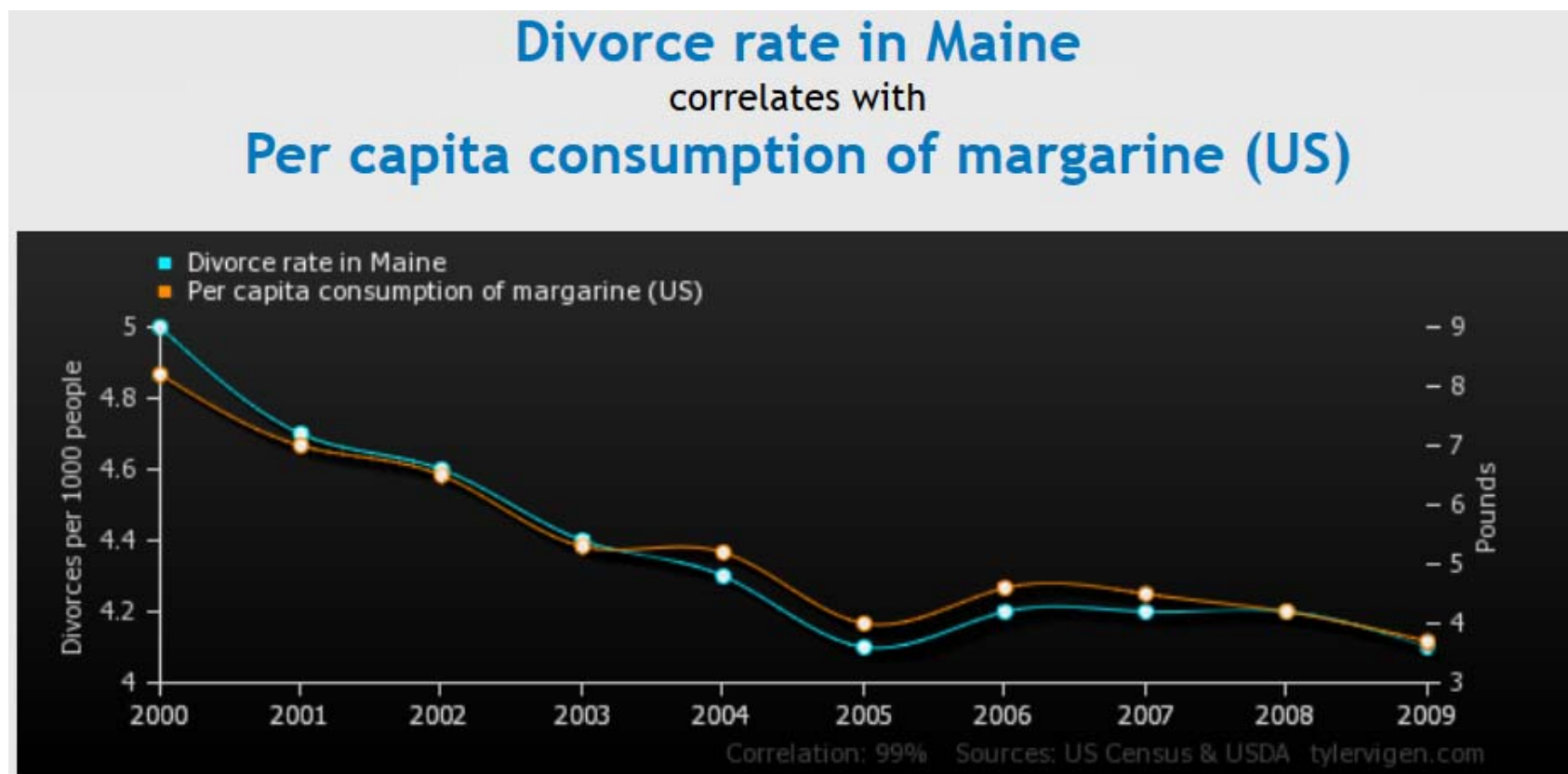
Spurious Correlation



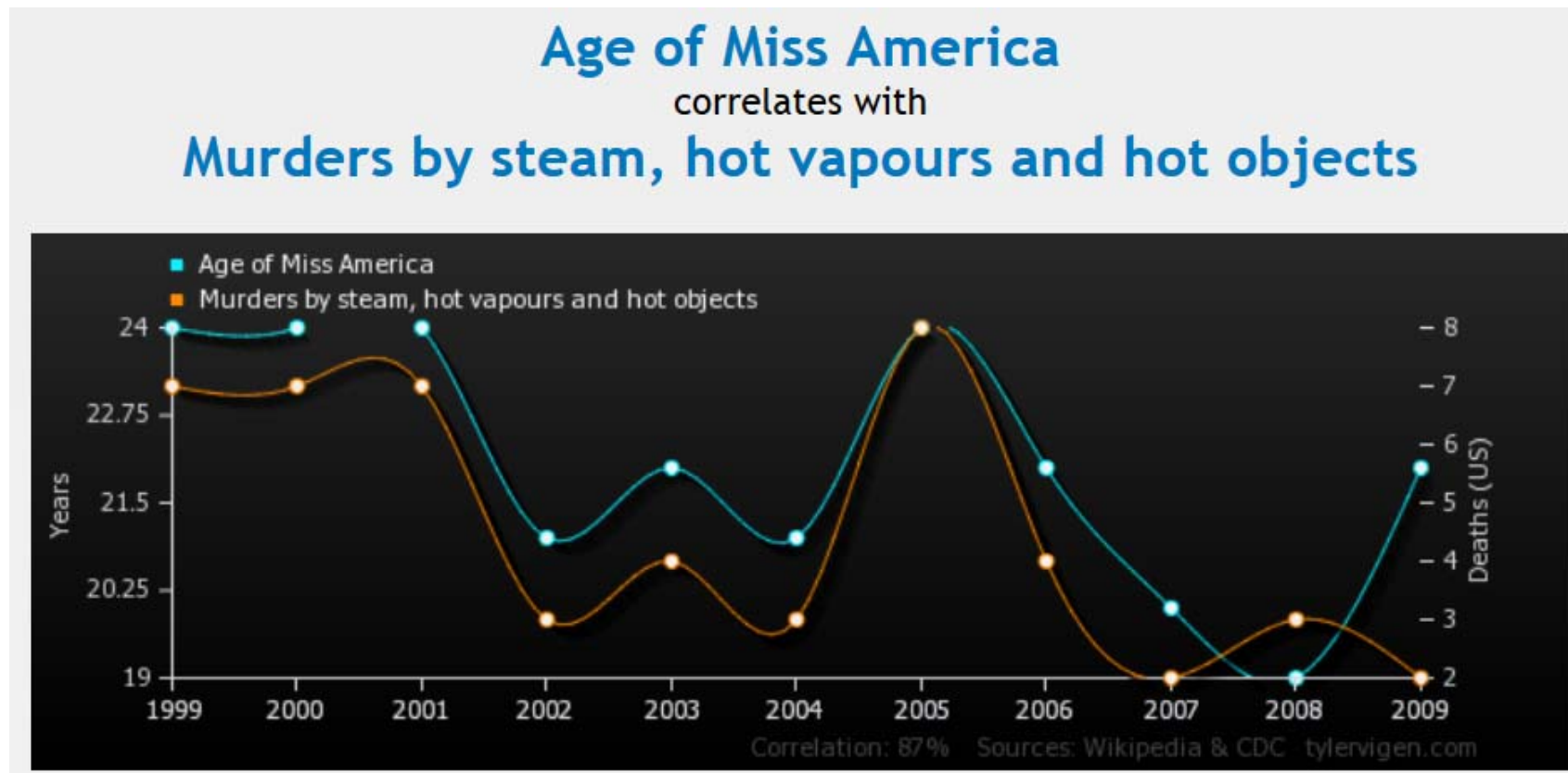
Correlation: 0.992082

<http://www.tylervigen.com/>

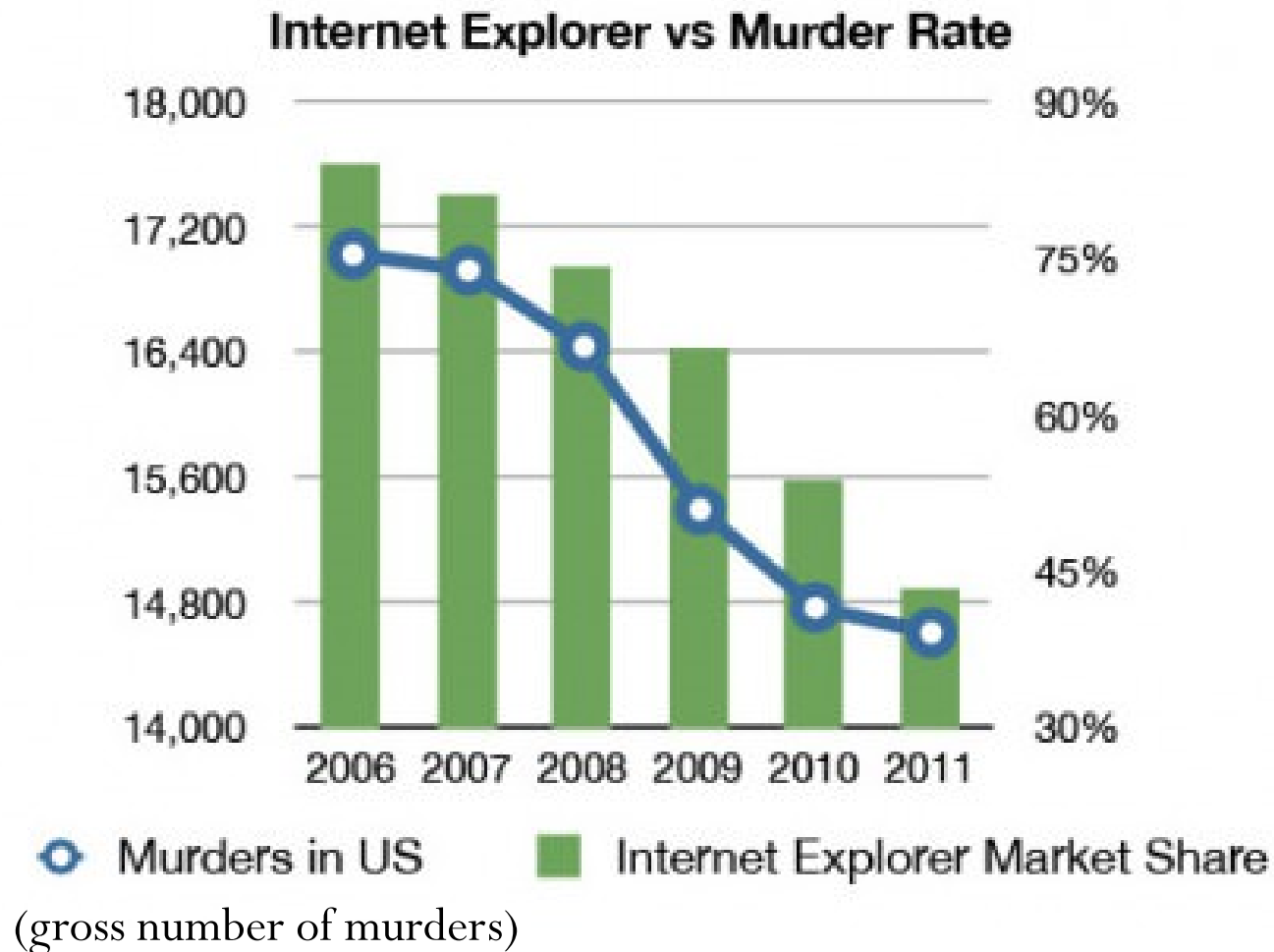
Spurious Correlation



Spurious Correlation



Spurious Correlation



Spurious Correlation

