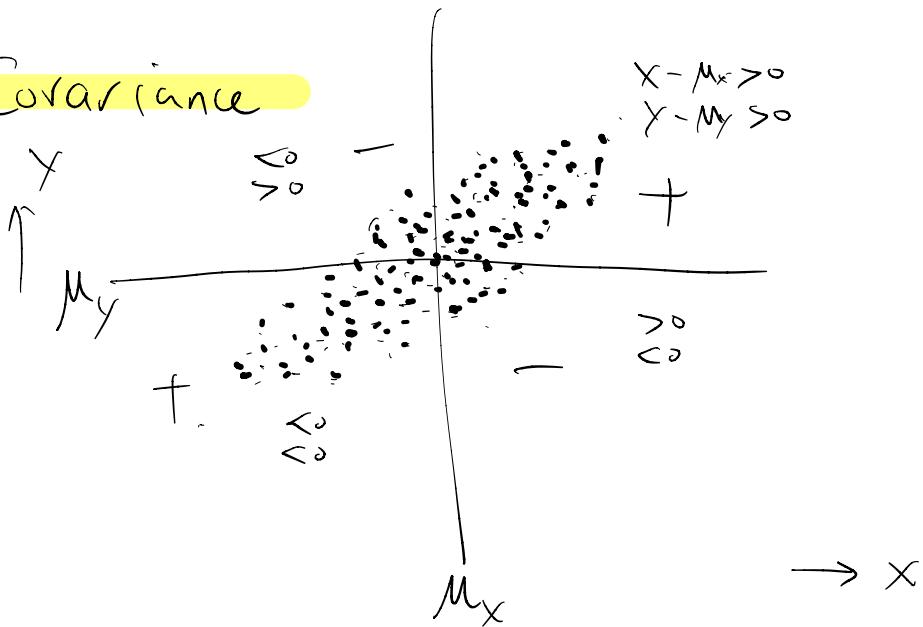


Covariance



$$\mu_x = E(x) \quad \mu_y = E(y)$$

$$\text{Cov}(x, y) = E((x - E(x))(y - E(y)))$$

$$\text{Cov}(x, x) = \text{Var}(x) = \sigma_x^2$$

$$\text{Cov}(y, y) = \text{Var}(y) = \sigma_y^2$$

$$\begin{aligned} \text{Corr}(x, y) &= \text{Cov}\left(\frac{x - \mu_x}{\sigma_x}, \frac{y - \mu_y}{\sigma_y}\right) \\ &= \frac{\text{Cov}(x, y)}{\sqrt{\text{Var}(x)} \sqrt{\text{Var}(y)}} = \rho \text{ standardize} \end{aligned}$$

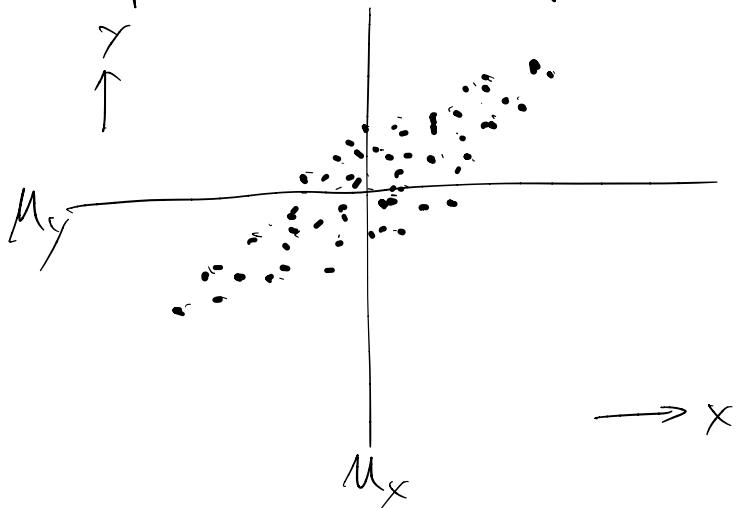
Covariance & Correlation

	X	Y	\tilde{X}	\tilde{Y}
1	x_1	y_1		
2	x_2	y_2		
:	:	:		
i	x_i	y_i	$\tilde{x}_i = x_i - \bar{M}_x$	$\tilde{y}_i = y_i - \bar{M}_y$
:	:	:		
n	x_n	y_n		

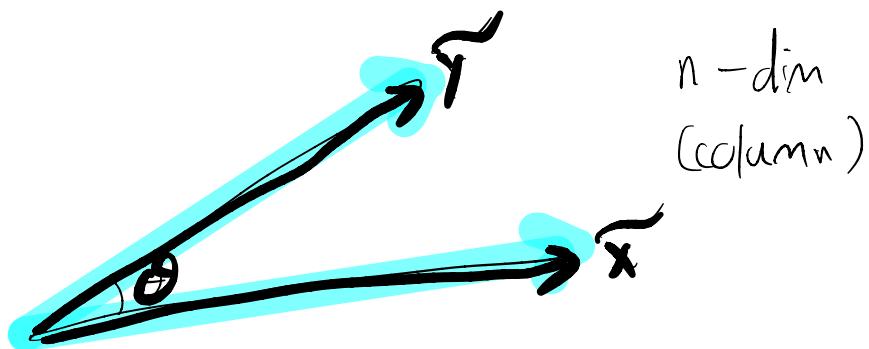
$$\begin{aligned} \text{Cov}(X, Y) &\approx \frac{1}{n} \sum_{i=1}^n (x_i - \bar{M}_x)(y_i - \bar{M}_y) \\ &= \frac{1}{n} \sum_{i=1}^n \tilde{x}_i \tilde{y}_i \\ &= \frac{1}{n} \langle \tilde{X}, \tilde{Y} \rangle \end{aligned}$$

$$\begin{aligned} \text{Var}(X) &= \text{Cov}(X, X) = \frac{1}{n} \langle \tilde{X}, \tilde{X} \rangle = \frac{1}{n} |\tilde{X}|^2 \\ \text{Var}(Y) &= \dots \quad \dots \quad = \frac{1}{n} |\tilde{Y}|^2 \end{aligned}$$

Scatterplot & Vector plot



2-dimensional
(row)

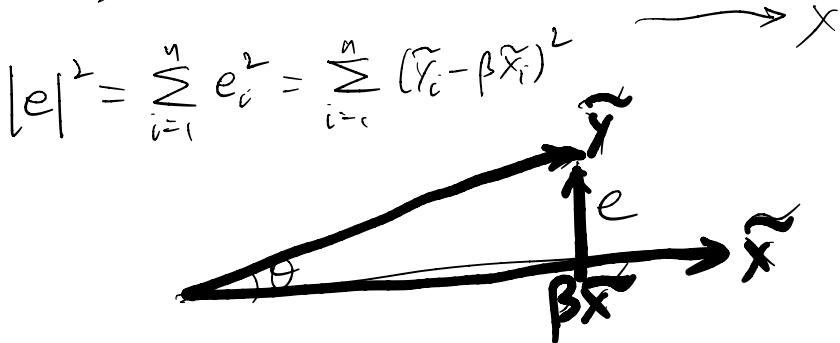
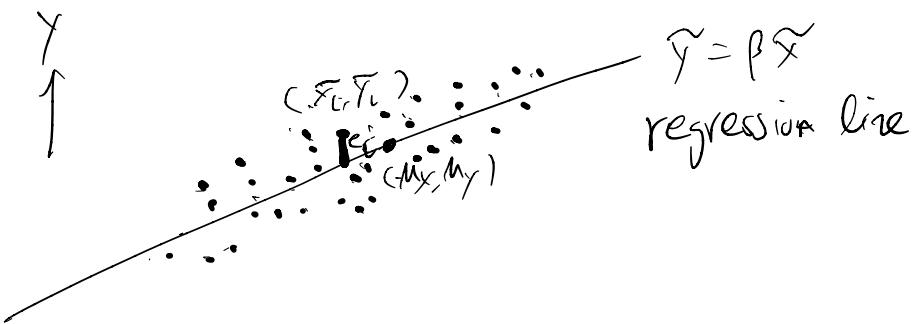


n-dim
(column_n)

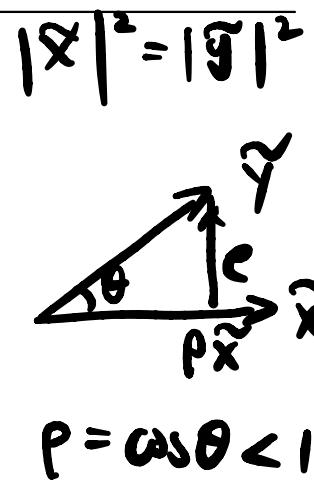
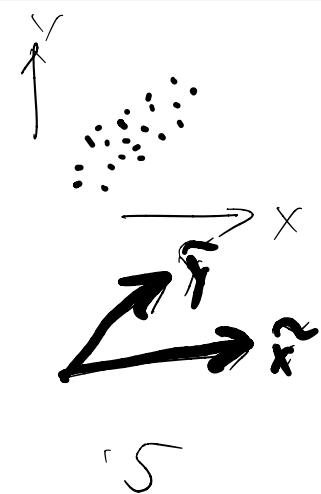
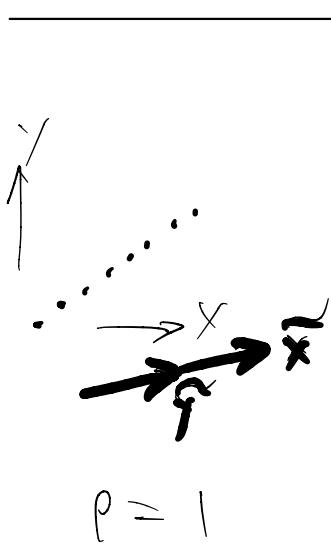
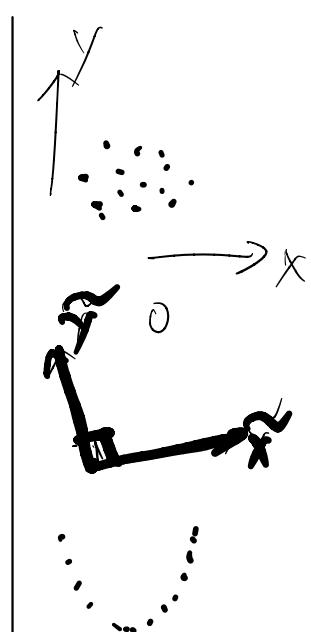
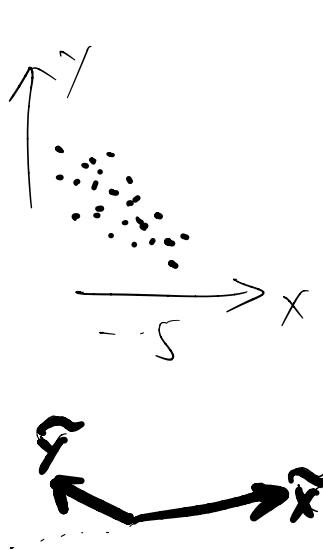
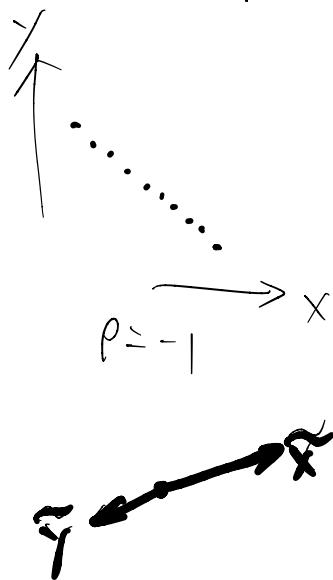
$$\text{Corr}(x, y) = \frac{\text{Cov}(x, y)}{\sqrt{\text{Var}(x)} \sqrt{\text{Var}(y)}} = \frac{\langle \tilde{x}, \tilde{y} \rangle}{|\tilde{x}| |\tilde{y}|} = \cos \theta$$

Correlation & Regression

	\tilde{x}	\tilde{y}	$\beta \tilde{x}$	$e = \tilde{y} - \beta \tilde{x}$
1				
2				
:				
i	$\tilde{x}_i = x_i - M_x$	$\tilde{y}_i = y_i - M_y$	$\beta \tilde{x}_i$	$e_i = \tilde{y}_i - \beta \tilde{x}_i$
:				
n				

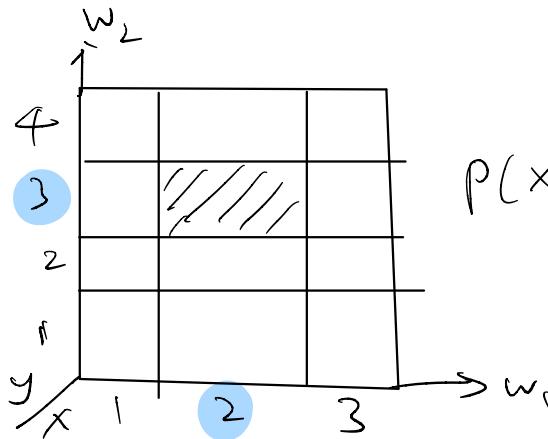


Scatter plot & Vector plot



Independence & Covariance

discrete

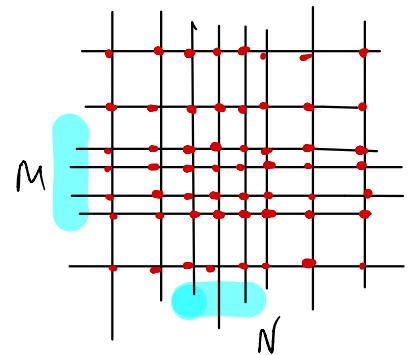


$$P(X, Y) = P(X) P(Y)$$

$M \times N$ points

Continuous

$$f(x, y) = f(x) f(y)$$



$$\text{Cov}(X, Y) = E((X - \mu_X)(Y - \mu_Y))$$

$$= \iint (x - \mu_x)(y - \mu_y) f(x) f(y) dx dy$$

$$= \int (x - \mu_x) f(x) dx \int (y - \mu_y) f(y) dy$$

$$= E(X - \mu_X) E(Y - \mu_Y) = 0$$

Conversely



$$X \sim \text{Unif} [-1, 1]$$

$$Y = X^2$$

$$E((X - \mu_X)(Y - \mu_Y))$$

$$= E(X(Y - \mu_Y))$$

$$= E(X(X^2 - \mu_Y))$$

$$\geq E(X^3) - E(X)\mu_Y = 0$$

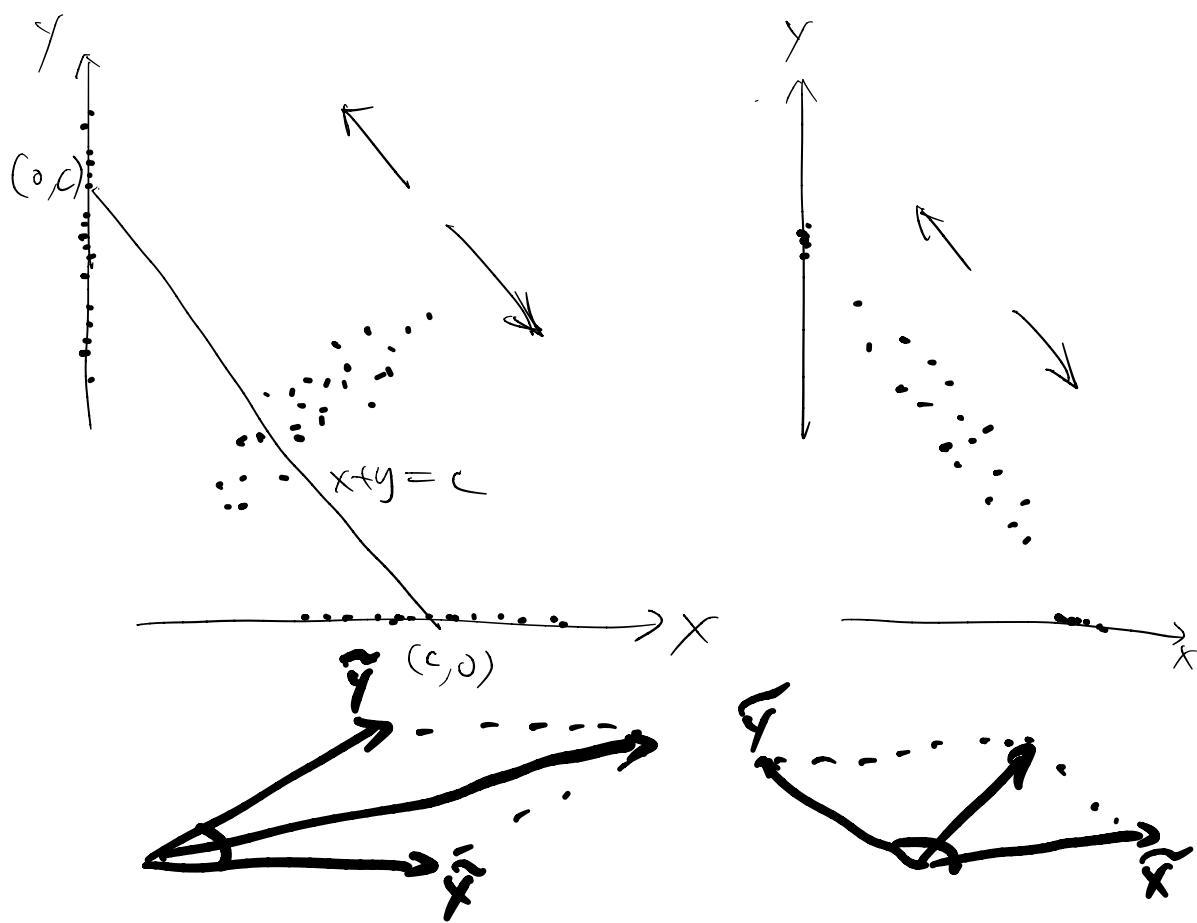
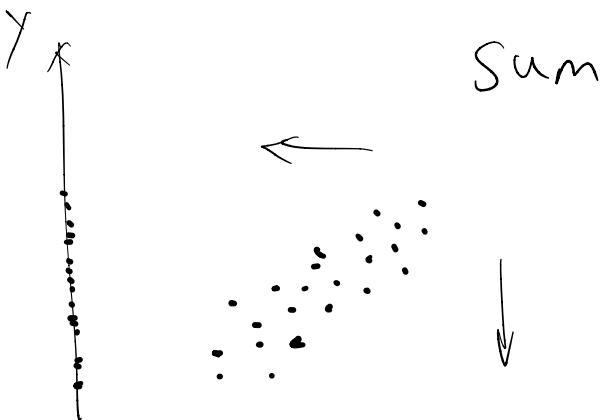
Independence $\longrightarrow \text{Cov} = 0$

$$\text{Cov}(X, Y) = E(XY) - E(X)E(Y)$$

Sum & average

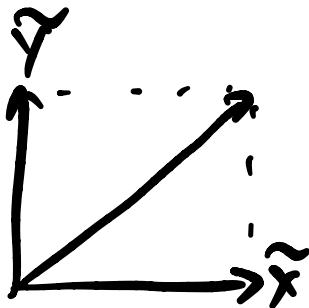
$$\begin{aligned} E(X+Y) &= \int (x+y) f(x,y) dx dy \\ &= \int x f(x,y) dx dy + \int y f(x,y) dx dy \\ &= E(X) + E(Y) \end{aligned}$$

$$\begin{aligned} \text{Var}(X+Y) &= E((X+Y) - E(X+Y))^2 \\ &= E((X-E(X)) + (Y-E(Y)))^2 \\ &= E((X-E(X))^2 + E(Y-E(Y))^2 \\ &\quad + 2E((X-E(X))(Y-E(Y))) \\ &= \text{Var}(X) + \text{Var}(Y) + 2\text{Cov}(X,Y) \end{aligned}$$



Independent $\rightarrow \text{Cov} = 0$

$$\text{Var}(X+Y) = \text{Var}(X) + \text{Var}(Y)$$



$$X_1 \quad X_2 \quad \dots \quad X_i \quad \dots \quad X_n \stackrel{\text{iid}}{\sim} f(x)$$

iid : independent & identically distributed

$$\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i \quad E(X_i) = \mu \\ \text{Var}(X_i) = \sigma^2$$

$$E(\bar{X}) = \frac{1}{n} \sum_{i=1}^n E(X_i) = \mu$$

$$\text{Var}(\bar{X}) = \frac{1}{n^2} \sum_{i=1}^n \text{Var}(X_i) = \frac{n\sigma^2}{n^2} = \frac{\sigma^2}{n}$$

Law of large number

$$P(|\bar{X} - \mu| > \varepsilon) \rightarrow 0$$

Special case

prob = proportion

(1) $X_i \sim \text{Bernoulli}(\frac{1}{2})$ "wholesale"

$$\mu = E(X_i) = \frac{1}{2} \quad \sigma^2 = \text{Var}(X_i) = \frac{1}{4}$$

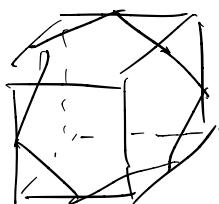
$$\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i = \frac{\text{# of heads}}{n} = \text{freq of heads}$$

Among all 2^n seqs., proportion of "bad" ones

(2) $X_i \sim \text{Unif}[0, 1] \rightarrow 0$

$$\mu = E(X_i) = \frac{1}{2} \quad \sigma^2 = \text{Var}(X_i) = \frac{1}{12}$$

$$(X_1, X_2, \dots, X_n) \sim \text{Unif}[0, 1]^n$$

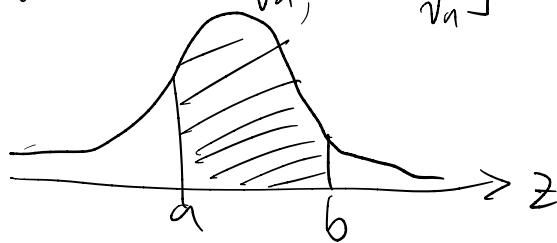


Volume of off-diagonal pieces $\rightarrow 0$

Central limit theorem

$$Z = \frac{\bar{X} - E(\bar{X})}{\sqrt{\text{Var}(\bar{X})}} = \frac{\bar{X} - \mu}{\sigma/\sqrt{n}} \xrightarrow{D} N(0, 1)$$

$$P(Z \in [a, b]) = \int_a^b \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}} dz$$
$$X \in [\mu + a \frac{\sigma}{\sqrt{n}}, \mu + b \frac{\sigma}{\sqrt{n}}]$$



or

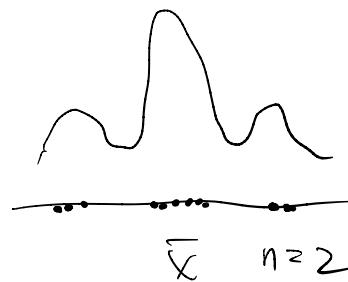
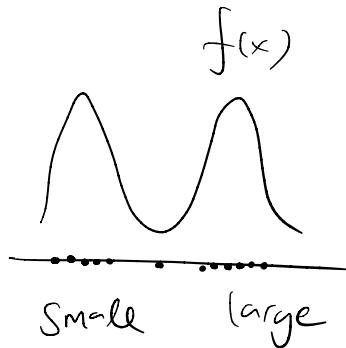
$$\bar{X} \rightarrow N\left(\mu, \frac{\sigma^2}{n}\right)$$

↓

CLT

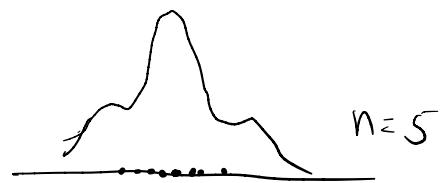
LLN

Example

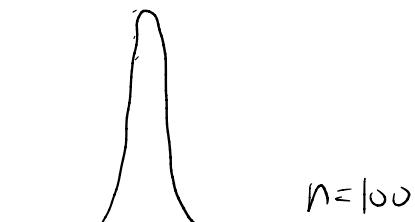


$$x_1 \quad x_2 \quad \stackrel{iid}{\sim} \quad f(x)$$

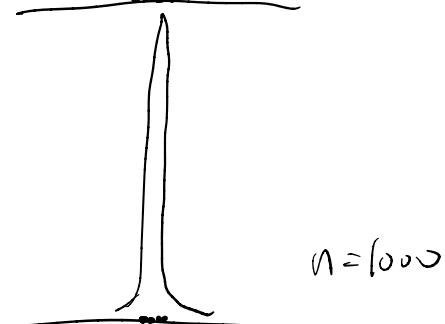
$$\bar{x} = \frac{x_1 + x_2}{2}$$



	x_2	small	large
small		small	medium
large		medium	large



$$\bar{x} \sim N(\mu, \frac{\sigma^2}{n})$$



Summary

Basic concepts

1 random variable

2 random variables

Limiting theorems

Stochastic processes