

Statements, rules, formulas and theorems:

L4P3

- **Multiplication Rule**

$$\begin{aligned} P(A_1 \cap A_2 \cap \dots \cap A_n) &= \\ &= P(A_1) * P(A_2 | A_1) * P(A_3 | A_1 \cap A_2) * \dots * P(A_n | A_1 \cap A_2 \cap \dots \cap A_{(n-1)}) \end{aligned}$$

L5P2

- **Multiplication Rule for a pair of independent events:**

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

- **Multiplication Rule for mutually independent events:**

$$P(A_1 \cap A_2 \cap \dots \cap A_n) = P(A_1) * P(A_2) * \dots * P(A_n)$$

L5P3

- **Total Probability Formula/формула полной вероятности**

Let a sample space be partitioned: $\Omega = \cup_{n \in [0..m]} H_n$;

then for every event $A \subseteq \Omega$

$$\begin{aligned} P(A) &= P(\cup_{n \in [0..m]} (A \cap H_n)) = \\ &= \sum_{n \in [0..m]} P(A \cap H_n) = \sum_{n \in [0..m]} P(A | H_n) * P(H_n). \end{aligned}$$

- **Bayes' Formula/формула Байеса**

Let a sample space be partitioned: $\Omega = \cup_{n \in [0..m]} H_n$;

then for every event A and hypothesis H_k the **posterior**

probability/апостериорная вероятность can be computed using **prior probabilities/априорные вероятности** as follows:

$$\begin{aligned} P(H_k | A) &= P(A | H_k) * P(H_k) / P(A) = \\ &= P(A | H_k) * P(H_k) / (\sum_{n \in [0..m]} P(A | H_n) * P(H_n)). \end{aligned}$$

L7P3

- **Total Expectation Formula/формула полного мат. ожидания**

If H_1, \dots, H_n is a partition then $P(A) = \sum_{k \in [1..n]} P(A | H_k) * P(H_k)$

and hence $P_X(x) = \sum_{k \in [1..n]} P_{X|H_k}(x) * P(H_k)$, then

$$M(X) = \sum_{k \in [1..n]} M(X | H_k) * P(H_k)$$

L9P1

- **Poisson Limit Theorem (law of rare events)/теорема Пуассона**

Assume we are given an infinite series of trials X_1, \dots, X_n, \dots where $X_n = B(n, p_n)$; $p_n = \mu/n$ (where μ is a constant).

For every $n > 0$ let $P_n(m)$ be the probability of m positive outcomes in the n -th trial; then this sequence converges to $e^{-\mu} * (\mu^m / m!)$

L9P2

- **Local Moivre-Laplace theorem/локальная теорема Муавра-Лапласа**

Assume we are given an infinite series of trials X_1, \dots, X_n , where $X_n = \text{binomial}(n, p)$ and $0 < p < 1$, $q = 1 - p$.

For every $n > 0$ let $P_n(m)$ be the probability of m positive outcomes in the n^{th} trial; then

$$P_n(m) = \frac{\phi_0(t_n)}{\sqrt{npq}} (1 + \alpha_n)$$

$$\text{where } \phi_0(t_n) = \frac{1}{\sqrt{2\pi}} e^{-t_n^2/2}, \quad t_n = \frac{m - np}{\sqrt{npq}}$$

$$\text{and } |\alpha_n| < \frac{c}{\sqrt{n}}$$

- **De Moivre-Laplace theorem/теорема Муавра-Лапласа**

Assume we are given an infinite series of trials X_1, \dots, X_n, \dots where $X_n = \text{binomial}(n, p)$ and $0 < p < 1$, $q = 1 - p$.

For every $n > 0$ let $P_n(m)$ be the probability of m positive outcomes in the n^{th} trial and $P_n(m_1, m_2) = P_n(m_1 \leq m \leq m_2)$; then

$$P_n(m_1, m_2) \approx \Phi_0(x_{m_2}) - \Phi_0(x_{m_1}),$$

$$x_{m_1} = \frac{m_1 - np}{\sqrt{npq}}, \quad x_{m_2} = \frac{m_2 - np}{\sqrt{npq}}$$

L10P1

- **Bernoulli's law/теорема Бернулли**

Assume we are given an infinite series of trials X_1, \dots, X_n, \dots , where $X_n = B(n, p)$, $0 < p < 1$. Then for every $\varepsilon > 0$:

$$\lim_{n \rightarrow \infty} P(|m/n - p| \leq \varepsilon) = 1$$

- **Chebyshev's inequality/неравенство Чебышёва**

Let X be a random variable with a finite expectation μ and finite non-zero deviation σ . Then for any real number $k > 0$

$$P(|X - \mu| \geq k \sigma) \leq 1/k^2.$$

- **Khinchin's (weak) law/слабый закон больших чисел**

Let X_1, X_2, \dots be an infinite sequence of *independent and identically distributed* (i.i.d. or IID) random variables with same sets of outcomes, finite expectation μ and finite non-zero deviation σ .

Let \underline{X}_n be $(X_1 + \dots + X_n)/n$

Then for every $\varepsilon > 0$ the probability

$$P(|\underline{X}_n - \mu| > \varepsilon) \rightarrow 0 \text{ (as } n \rightarrow \infty)$$

L11P3

- **Central Limit Theorem/центральная предельная теорема**

Let X_1, X_2, \dots be an infinite sequence of IID (i.e. independent and identically distributed) random variables with a finite expectation μ and finite non-zero deviation σ , $S_n = X_1 + \dots + X_n$.

Then $\frac{S_n - n\mu}{\sigma\sqrt{n}}$ converges in distribution to $N(0,1)$ (as $n \rightarrow \infty$).

- **68-95-99.7 rule/правило одного-двух-трёх сигм** is a shorthand to remember the percentage of values that lie within a band around the mean in a normal distribution:
–68.27%, 95.45% and 99.73% of the values lie within distance σ , 2σ and 3σ of the mean μ .