

# MS-E1603 - Useful formulas

## Inequalities

Markov	$\mathbb{P}( X  \geq a) \leq \frac{\mathbb{E}( X )}{a}, a \geq 0$
Chebyshev	$\mathbb{P}( X - \mathbb{E}(X)  \geq a) \leq \frac{\text{Var}(X)}{a^2}$
Cauchy-Schwarz	$\mathbb{E}(XY) \leq \sqrt{\mathbb{E}(X^2)\mathbb{E}(Y^2)}$ $(\sum_{i=1}^n x_i y_i)^2 \leq (\sum_{i=1}^n x_i^2) (\sum_{i=1}^n y_i^2)$
Hoeffding	$\mathbb{P}( S_n - \mathbb{E}(S_n)  \geq a) \leq 2e^{-\frac{2a^2}{\sum_{i=1}^n (b_i - a_i)^2}},$ where $S_n = \sum_{i=1}^n X_i$ with $a_i \leq X_i \leq b_i, \perp$
Jensen	$\phi(\mathbb{E}(X)) \leq \mathbb{E}(\phi(X))$ for convex $\phi$
Chernoff bound	$\mathbb{P}(X \geq a) \leq \inf_{t>0} e^{-ta} \mathbb{E}(e^{tX}),$
Union bound	$\mathbb{P}(\bigcup_{i=1}^n A_i) \leq \sum_{i=1}^n \mathbb{P}(A_i),$ where $n \in \mathbb{N}$ or $n = \infty$

## Statistics and combinatorics

Binomial coefficient	$\binom{n}{k} = \frac{n!}{(n-k)!k!}$ $\frac{n^k}{k^k} \leq \binom{n}{k} \leq \frac{n^k}{k!}$
Binomial theorem	$(x + y)^n = \sum_{k=0}^n \binom{n}{k} x^{n-k} y^k$
Inclusion-exclusion	$\mathbb{P}(\bigcup_i^n A_i) = \sum_i^n \mathbb{P}(A_i) - \sum_{i<j} \mathbb{P}(A_i \cap A_j) + \dots$
Order statistics	$\mathbb{P}(X_{(k)} \leq a) = \sum_{i=0}^{n-k} \binom{n}{i} F(a)^{n-i} (1 - F(a))^i$ where $X_1, \dots, X_n$ i.i.d. with cdf $F(x)$
Law of total probability	$\mathbb{P}(A) = \mathbb{E}[\mathbb{P}(A X)]$ esp. $\mathbb{P}(A) = \sum_i \mathbb{P}(A X = x_i) \mathbb{P}(X = x_i)$
Law of total expectation	$\mathbb{E}(X) = \mathbb{E}[\mathbb{E}[X Y]]$
Law of total variance	$\text{Var}(X) = \text{Var}(\mathbb{E}[X Y]) + \mathbb{E}[\text{Var}(X Y)]$

## Convergence of random variables

In probability	$X_n \xrightarrow{p} X \iff \mathbb{P}( X_n - X  \geq a) \rightarrow 0,$ for all $a > 0$
Weakly/in distribution	$X_n \xrightarrow{d} X \iff \mathbb{E}(f(X_n)) \rightarrow \mathbb{E}(f(X))$ for all bounded and continuous $f$
Almost surely*	$X_n \xrightarrow[n \rightarrow \infty]{a.s.} X \iff \mathbb{P}(X_n \rightarrow X) = 1$
Weak/strong laws of large numbers	(weak) $\frac{1}{n} \sum_{i=1}^n X_i \xrightarrow{p} \mathbb{E}(X)$ (strong) $\frac{1}{n} \sum_{i=1}^n X_i \xrightarrow[n \rightarrow \infty]{a.s.} \mathbb{E}(X)$ where $X, X_1, X_2, \dots$ i.i.d.
Central limit theorem	$\frac{1}{\sigma\sqrt{n}} \sum_{i=1}^n (X_i - \mathbb{E}(X)) \xrightarrow{d} \mathcal{N}(0, 1)$ where $X, X_1, X_2, \dots$ i.i.d. and $\sigma^2 = \text{Var}(X) < \infty$

## Miscellaneous

"Little-o" notation	$f(x) = o(g(x)) \iff f(x)/g(x) \rightarrow 0$
"Big-o" notation	$f(x) = O(g(x)) \iff \limsup  f(x)/g(x)  < \infty$
"Big-theta" notation	$f(x) = \Theta(g(x)) \iff f(x) = O(g(x))$ and $g(x) = O(f(x))$
"A with high probability/w.h.p"	$\Pr(A) \rightarrow 1$
Taylor's approximation	$f(x) = f(a) + f'(a)(x-a) + \dots + \frac{f^{(k)}(a)}{k!}(x-a)^k$ $+ o( x-a ^k)$
Stirling's approximation	$n! = \Gamma(n+1) = \sqrt{2\pi n} e^{-n} n^n (1 + O(n^{-1}))$ $\ln n! = n \ln n - n + O(\ln n)$
Dominated convergence*	$X_n \rightarrow X,  X_n  \leq Y$ and $\mathbb{E}(Y) < \infty$ $\implies \mathbb{E}(X_n) \rightarrow \mathbb{E}(X)$
Monotone convergence*	$X_n \rightarrow X$ and $0 \leq X_1 \leq X_2 \leq \dots$ $\implies \mathbb{E}(X_n) \rightarrow \mathbb{E}(X)$

\* see e.g. MS-E1600 Probability theory for details