

<b>Woodbury Matrix Identity</b>	$(A + UCV)^{-1} = A^{-1} - A^{-1}U(C^{-1} + VA^{-1}U)^{-1}VA^{-1}$
<b>Sherman Morrison Formula</b>	Given <i>invertible</i> $A \in R^{n \times n}$ and $u, v \in R^n$ vectors. Then $A + uv^T$ is invertible iff $1 + v^T A^{-1}u \neq 0$ . Then $(A + uv^T)^{-1} = A^{-1} - \frac{A^{-1}uv^T A^{-1}}{1 + v^T A^{-1}u}$
<b>Cauchy Schwarz Inequality</b>	$ \langle U, V \rangle  \leq \langle U, U \rangle \cdot \langle V, V \rangle$
<b>Markov Inequality</b>	$X$ is a non-negative rv and $a > 0$ then $P(X \geq a) \leq \frac{E(X)}{a}$
<b>Minkowski inequality</b>	$\ f + g\ _p \leq \ f\ _p + \ g\ _p$
<b>Holder Inequality</b>	Let $p, q \in [1, \infty], 1/p + 1/q = 1$ . Then for all real/complex functions $f, g, \ fg\ _1 \leq \ f\ _p \ g\ _q$
<b>Jensen Inequality</b>	Given $X$ is a rv and $\phi$ is convex. Then $\phi(E(X)) \leq E(\phi(X))$
<b>Johnson Lindenstrauss Property</b>	A distribution on matrices $S \in R^{k \times n}$ has $(\epsilon, \delta, l)$ -JL moment property if $\forall x \in R^n,  x _2 = 1, E_S( Sx _2^2 - 1 ^l) \leq \epsilon^l \delta$
<b>Hoeffding Inequality</b>	Let $X_1, \dots, X_n$ be independent rv in $[0, 1]$ . Let $\bar{X} = \frac{1}{n} \sum_i X_i$ . Then $P(\bar{X} - E(\bar{X}) \geq t) \leq \exp(-2nt^2)$
<b>Chebyshev Inequality</b>	$P( X - E(X)  \geq a) \leq \frac{Var(X)}{a^2}$
<b>Golden Thompson</b>	For Hermitian matrices $A, B$ $tr \exp(A + B) \leq tr(\exp(A) \exp(B))$
<b>Oppenheimer Inequality</b>	For PSD matrices $A, B$ , $\det(A \circ B) \geq (\prod a_{ii}) \det(B)$
<b>Ky Fan Matrix Inequality</b>	Let $A, B$ be $n \times n$ Hermitian matrices. Then $\lambda(A) + \lambda(B) \succ \lambda(A + B)$ , where $\succ$ denotes the majorization relation.
<b>McDiarmid Inequality</b>	Multidimensional Hoeffding. Let $X_1, \dots, X_m$ be independent rv taking values in $\chi$ . Further, let $f : \chi^m \mapsto \mathbb{R}$ be a function of $X_1, \dots, X_m$ that satisfies coordinate-wise bounded differences, $\forall i \forall x_1, \dots, x_m, x_i^* \in \chi,  f(x_1, \dots, x_i, \dots, x_m) - f(x_1, \dots, x_i^*, \dots, x_m)  \leq c_i$ . Then $\mathbb{P}( f - \mathbb{E}(f)  \geq \epsilon) \leq \exp\left(\frac{-2\epsilon^2}{\sum_{i=1}^m c_i^2}\right)$ .
<b>Cortes Sampling Concentration Inequality</b>	Let $X_i, i \in [m]$ be a set of drawn <i>without replacement</i> from an underlying finite set of $m + u$ elements. Let $f : \chi^m \mapsto \mathbb{R}$ be a function of $X_1, \dots, X_m$ be symmetric (up to permutation of parameters) and obey coordinate-wise bounded differences, then $\mathbb{P}( f - \mathbb{E}(f)  \geq \epsilon) \leq \exp\left(\frac{-2\epsilon^2}{\alpha(m, u)c^2}\right)$ , where $\alpha(m, u) = \frac{mu}{m+u-0.5} \times \frac{1}{1-1/(2\max(m, u))}$ .
<b>Simple Gaussian Concentration Inequality</b>	Let $X_1, \dots, X_n$ be iid $(0, 1)$ Gaussian rv. Let $f : \mathbb{R}^n \mapsto \mathbb{R}$ be 1-Lipschitz. Then for any $\lambda > 0$ , $\mathbb{P}( f(X) - \mathbb{E}(f(X))  \geq \lambda) \leq C \exp(-c\lambda^2)$ for constants $c, C$ .
<b>Bounds for central term in binomial coefficient</b>	$\frac{4^n}{2n+1} \leq \binom{2N}{N} \leq 4^n$

<b>Generating function of Catalan Numbers</b>	$G(z) = \sum_{i=0}^{\infty} \frac{1}{n+1} \binom{2n}{n} z^i = \frac{1-\sqrt{1-4z}}{2z} = \frac{2}{1+\sqrt{1-4z}}$
<b>Weyl Matrix Perturbation Inequality</b>	If $M = H + P$ are $n \times n$ matrices with eigenvalues $\mu_i, \nu_i, \rho_i$ (ordered in descending order) respectively, then $\nu_i + \rho_n \leq \mu_i \leq \nu_i + \rho_1$
<b>Walds Identity</b>	Let $X_1, \dots$ be a sequence of (potentially infinite) independent random variables with identical means. Let $N$ be any <i>stopping time</i> (or independent of $X$ s). Let $S_N = \sum_{i=1}^N X_i$ . Then $\mathbb{E}(S_N) = \mathbb{E}(N)\mathbb{E}(X)$ . Note $N$ can be weakly dependent on $X$ .
<b>Variance of Random Sum</b>	If $N$ and $X_i$ are independent and $X$ s have equal variance, $\text{Var}(\sum_{i=1}^N X) = \mathbb{E}(N)\text{Var}(X_i) + \mathbb{E}(X)^2\text{Var}(N)$
<b>Eckart-Young-Mirsky Theorem</b>	The k-truncated SVD of a matrix $A$ gives the best rank-k approximation for $A$ in the spectral and frobenius norm. $A_k = U\Sigma_k V^T = \text{argmin}_{B \text{ rank-k}} \ A - B\ _2 = \sigma_{k+1}(A)$
<b>Wolfe Conditions</b>	(i) (Armijo rule) $f(\mathbf{x}_k + \alpha_k \mathbf{p}_k) \leq f(\mathbf{x}_k) + c_1 \alpha_k \mathbf{p}_k^T \nabla f(\mathbf{x}_k)$ (ii) (Curvature) $-\mathbf{p}_k^T \nabla f(\mathbf{x}_k + \alpha_k \mathbf{p}_k) \leq -c_2 \mathbf{p}_k^T \nabla f(\mathbf{x}_k)$ , where $\alpha_k$ is step size, $\mathbf{p}_k$ is descent direction and $0 < c_1 < c_2 < 1$
<b>Euler-Lagrange Equation</b>	Functionals of form $S(f(t)) = \int_t L(f(t), \dot{f}(t), t) dt$ , stationary $f$ where $\frac{\partial L(f, \dot{f}, t)}{\partial f} = \frac{d}{dt} \frac{\partial L(f, \dot{f}, t)}{\partial \dot{f}}$