## Fourmlas Sheet.

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## Probability.

Multiplicative Chernoff bound. Suppose  $X_1,...,X_n$  are independence random variables taking values in  $\{0,1\}$  Let X denote their sum and let  $\mu = \mathbf{E}\left[\sum_i^n X_i\right]$  denote the sum's expected value. Then for any  $\delta > 0$ :

$$\Pr[X \ge (1+\delta)\,\mu] \le e^{-2\frac{\delta^2\mu^2}{n}} \Pr[|X-\mu| \ge \delta\mu] \le 2e^{-\delta^2\mu/3}, \qquad 0 \le \delta \le 1$$

**Bernstein inequalities.**  $X_1, ..., X_n$  are independence random variables with zero mean  $(\mu = 0)$ . Suppose that  $|X_i| \leq M$  almost surely, for all *i*. Then, for all positive t:

$$\mathbf{Pr}\left[\sum_{i}^{n}X_{i} \geq t\right] \leq \exp\left(-\frac{\frac{1}{2}t^{2}}{\sum_{i}\mathbf{E}\left[X_{i}^{2}\right] + \frac{1}{3}M}t\right)$$

For example, consider coins taking values  $\pm 1$  with probability  $\frac{1}{2}$ , then for every positive  $\varepsilon$ .

$$\mathbf{Pr}\left[\left|\frac{1}{n}\sum_{i}^{n}X_{i}\right| \geq \varepsilon\right] \leq 2\exp\left(-\frac{n\varepsilon^{2}}{2\left(1+\frac{\varepsilon}{3}\right)}\right)$$

There is also a weakly dependent generalization version, that go as follow. Let  $X_0, X_1, X_2, \ldots X_n$  random variables. Suppose that for all integers i it holds:

$$\begin{split} &\mathbf{E}\left[X_{i}|X_{0},X_{1},X_{2},\ldots X_{i-1}\right]=0 \\ &\mathbf{E}\left[X_{i}^{2}|X_{0},X_{1},X_{2},\ldots X_{i-1}\right]=R_{i}\mathbf{E}\left[X_{i}^{2}\right] \\ &\mathbf{E}\left[X_{i}^{k}|X_{0},X_{1},X_{2},\ldots X_{i-1}\right] \\ &\leq \frac{1}{2}\mathbf{E}\left[X_{i}|X_{0},X_{1},X_{2},\ldots X_{i-1}\right]L^{k-2}k! \end{split}$$

Then:

$$\mathbf{Pr}\left[\sum_{i}^{n} X_{i} \geq 2t \sqrt{\sum_{i=1}^{n} R_{i} \mathbf{E}\left[X_{i}^{2}\right]}\right] \leq \exp\left(-t^{2}\right)$$

**Jensen's inequality.** If X is a random variable and  $\phi$  is a convex function, then:

$$\phi\left(\mathbf{E}\left[X\right]\right) \leq \mathbf{E}\left[\phi\left(X\right)\right] \Rightarrow \mathbf{E}\left[X\right] \leq \phi^{-1}\left(\mathbf{E}\left[\phi\left(X\right)\right]\right)$$
$$\mathbf{E}\left[X\right] \leq \ln\left(\mathbf{E}\left[e^{X}\right]\right)$$
$$\mathbf{E}\left[X\right] \geq e^{\mathbf{E}\left[\ln\left(X\right)\right]}$$

**Paley–Zygmund inequality.** bounds the probability that a positive random variable is small, in terms of its first two moments. Could be thought as the lower bound Markov version. If a r.v X is always positive and has a finate variance, then for  $0 \le \tau \ge 1$ :

$$\mathbf{Pr}\left[X > \tau \mathbf{E}\left[X\right]\right] \ge \left(1 - \tau\right)^2 \frac{\mathbf{E}\left[X\right]^2}{\mathbf{E}\left[X^2\right]}$$
$$\mathbf{Pr}\left[X > \mathbf{E}\left[X\right] - \tau\sigma\right] \ge \frac{\tau^2}{1 + \tau^2}$$

Marcinkiewicz–Zygmund inequality.  $X_1, ..., X_n$  are independence random variables with zero mean  $(\mu = 0)$  and  $\mathbf{E}[|X_i|^p] < \infty$ , then there exist constants  $A_p, B_p$  which depend only on p such:

$$A_p \mathbf{E}\left[\left(\sum_{i=1}^n |X_i|^2\right)^{p/2}\right] \leq \mathbf{E}\left[|\sum_{i=1}^n X_i|^p\right] \leq B_p \mathbf{E}\left[\left(\sum_{i=1}^n |X_i|^2\right)^{p/2}\right]$$

Cauchy–Schwarz Expectation Inequality. Let X, Y be random variables then the inequality becomes:

$$|\mathbf{E}\left[XY\right]|^2 \leq \mathbf{E}\left[X^2\right]\mathbf{E}\left[Y^2\right]$$

**Union Of Pairwise Independent.** Denote by  $\{A_i, i \in \{1, 2, ..., n\}\}$  a set of n bernoulli events with probability of occurrence  $\mathbb{P}(A_i) = p_i$  for each i. Suppose the Joint probability distribution probabilities are given by  $\mathbb{P}(A_i \cap A_j) = p_{ij}$  for every pair of indices (i, j). Kounias Bounds for the probability of a union by:

$$\mathbb{P}(\cup_{i} A_{i}) \leq \sum_{i=1}^{n} p_{i} - \max_{j \in \{1, 2, \dots, n\}} \sum_{i \neq j} p_{ij},$$

which subtracts the maximum weight of a star spanning tree on a complete graph. Hunter-Worsley prove that is sufficient to consider only the wight of the minium spanning tree.

However, when the variables are **pairwise independent** Ramachandra-Natarajan showed that the Kounias-Hunter-Worsley is tight.

## Inequalitys.

**Sedrakyan's inequality.** For any reals $a_0, a_1, a_2, \dots a_n$  and positive eals  $b_0, b_1, b_2, \dots b_n$  we have:

$$\frac{a_1^2}{b_1} + \frac{a_2^2}{b_2} + \dots + \frac{a_n^2}{b_n} \ge \frac{(a_1 + a_2 + \dots + a_n)^2}{b_1 + b_2 + \dots + b_n}$$

## Expanders.

**Second Eigenvalue.** Let A be the adjacency matrix of  $\Delta$  regular graph, then the seconed eigenvalue is:

$$\lambda = \max_{f \perp \mathbf{1}} \frac{f^\top A f}{f^\top f}$$

An exapmle for use case, consider the Cayley Graph defined by the union of two generators et and a homriphisem of it, namly S and gS for some  $g \in$  the group. Then we have that the new spacrtial gap is at most two times the orignal one:

$$\lambda' = \max_{f \perp 1} \frac{f^{\top} (A_S + A_{gS}) f}{f^{\top} f}$$
$$\leq \max_{f \perp 1} \frac{f^{\top} A_S f}{f^{\top} f} + \max_{f \perp 1} \frac{f^{\top} A_{gS} f}{f^{\top} f}$$
$$\leq \lambda + \lambda = 2\lambda$$