

# CS 574: Randomized Algorithms

## Lecture 11. Chernoff Bounds: An Application to Routing

September 29, 2015

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## Theorem

*(Chernoff Bound) Let  $X_i$  be independent bernoulli r.v.s (not necessarily identical), with  $E[X_i] = p_i$ . Let  $x = \sum_{i=1}^n X_i$  and  $\mu = \sum_{i=1}^n p_i$  then for every  $\beta \geq 1$  we have*

$$Pr[X \geq \beta\mu] \leq \left(\frac{e^{\beta-1}}{\beta^\beta}\right)^\mu$$

and

$$Pr[X \leq \frac{\mu}{\beta}] \leq \left(\frac{e^{1/\beta-1}}{\beta^\beta}\right)^\mu$$

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If we set  $\beta = 1 + \delta$  then we get slightly weaker but more simple bounds:

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$$Pr[X \geq (1 + \delta)\mu] \leq e^{-\frac{\delta^2\mu}{3}}$$

and

$$Pr[X \leq (1 - \delta)\mu] \leq e^{-\frac{\delta^2\mu}{2}}$$

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*Min-Congestion Disjoint Paths Problem*. Given a directed graph  $G(V, E)$  and a set of terminal pairs  $\mathcal{T} = \{(s_i, t_i)\}_{i=1}^k$ , where each  $s_i, t_i \in V$ , the goal is to choose for every pair  $i$ , a directed  $s_i - t_i$  path  $\gamma_i$  as to minimize the max congestion of any edge  $e$ .



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$$\text{opt}(\mathcal{J}) = \min \left\{ \max_{e \in E} \#\{i : e \in \gamma_i\} \right\}$$

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- Since NP-hard, what about approximation (find a set of paths such that the congestion is at most  $\alpha \cdot \text{OPT}$ )?

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## Theorem

*Let  $n = |V|$  and suppose that  $n \geq 4$ . If there is a multi-flow  $F$  that routes the demands  $(s_i, t_i)$  then there is an integral flow  $F'$  that routes the demands and*

$$\max_{e \in E} F'(e) \leq C \frac{\log n}{\log \log n} (1 + \max_{e \in E} F(e))$$

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**Class Assignment:** Write the (poly-sized) LP for the multi-flow above.