

# Concentration Inequalities (cont.)

$X_1, \dots, X_n$  - indep. r.v.'s

$f(X_1, \dots, X_n)$  : function that is "not too sensitive" to changes in each  $X_i$

$$\mathbb{P} \left\{ \left| f(X_1, \dots, X_n) - \mathbb{E}f(X_1, \dots, X_n) \right| \geq \varepsilon \right\} \leq K_1 e^{-K_2 \varepsilon}$$

where  $K_1 > 0$  is a small "absolute" constant

$$K_2 = K_2(f, \varepsilon) > 0$$

Recap: exponential Markov trick

$$\mathbb{P} \{ X - \mathbb{E}X \geq t \} \leq \inf_{s \geq 0} \left\{ e^{-st} \mathbb{E}[e^{s(X - \mathbb{E}X)}] \right\} \quad t > 0$$

$$\mathbb{P} \{ X - \mathbb{E}X \leq -t \} \leq \inf_{s \geq 0} \left\{ e^{-st} \mathbb{E}[e^{s(X - \mathbb{E}X)}] \right\}$$

- need good bounds on  $\psi(s) := \log \mathbb{E}[e^{s(X - \mathbb{E}X)}]$

Hoeffding's Lemma. If  $X \in [a, b]$  a.s., then

$$\log \mathbb{E}[e^{s(X - \mathbb{E}X)}] \leq \frac{s^2 (b-a)^2}{8}$$

Corollary: 1)  $\mathbb{P} \{ X \geq \mathbb{E}X + t \} \leq \exp\left(-\frac{2t^2}{(b-a)^2}\right)$   
 $\mathbb{P} \{ X \leq \mathbb{E}X - t \} \leq \exp\left(-\frac{2t^2}{(b-a)^2}\right)$

2) If  $X_i \in [a_i, b_i]$  a.s.,  $1 \leq i \leq n$ , and are independent, then

$$\mathbb{P} \left\{ \sum_{i=1}^n X_i - \sum_{i=1}^n \mathbb{E}X_i \geq t \right\} \leq \exp\left(-\frac{2t^2}{\sum_{i=1}^n (b_i - a_i)^2}\right)$$

# Proof of Hoeffding's Lemma

w.l.o.g.  $\mathbb{E}X = 0$ ,  $\psi(s) = \log \mathbb{E}(e^{sX})$

1) Properties of  $\psi(s)$ :  $\psi'(s) = \frac{\mathbb{E}(Xe^{sX})}{\mathbb{E}(e^{sX})}$   
 $\psi''(s) = \frac{\mathbb{E}(X^2 e^{sX})}{\mathbb{E}(e^{sX})} - \left( \frac{\mathbb{E}(Xe^{sX})}{\mathbb{E}(e^{sX})} \right)^2$

2) Twisting:  $X \sim P \rightarrow U \sim \tilde{P}$

$$\int_{\mathbb{R}} f(u) \tilde{P}(du) = \frac{\int_{\mathbb{R}} f(x) e^{sx} P(dx)}{\int_{\mathbb{R}} e^{sx} P(dx)} \quad \forall f$$

$$P[U \in A] = \frac{\mathbb{E}[\mathbb{1}_{\{X \in A\}} e^{sX}]}{\mathbb{E}[e^{sX}]}$$

3)  $X \in (a, b)$  a.s.  $\Rightarrow U \in (a, b)$  a.s.

$$\mathbb{E}U = \frac{\mathbb{E}[Xe^{sX}]}{\mathbb{E}[e^{sX}]} = \psi'(s)$$

$$\text{Var}(U) = \mathbb{E}U^2 - (\mathbb{E}U)^2 = \psi''(s)$$

4) Claim: if  $U$  is any r.v. that takes values in  $(a, b)$  a.s., then

$$\text{Var}(U) \leq \frac{(b-a)^2}{4}$$

5) For any  $s$ ,  $\psi''(s) \leq \frac{(b-a)^2}{4}$ .

6)  $\psi(s) = \int_0^s \int_0^t \psi''(r) dr dt$

$$\int_0^t \psi''(r) dr = \psi'(t) - \underbrace{\psi'(0)}_{=0} = \psi'(t)$$

$$\int_0^s \psi'(t) dt = \psi(s) - \underbrace{\psi(0)}_{=0} = \psi(s)$$

$$\Rightarrow \psi(s) = \int_0^s \int_0^t \underbrace{\psi''(r)}_{\leq \frac{(b-a)^2}{4}} dr dt \leq \frac{s^2 (b-a)^2}{8}$$



McDiarmid's inequality: Concentration beyond sums

$X_1, \dots, X_n$  - indep. r.v.'s taking values in  $\mathcal{X}$

$f: \mathcal{X}^n \rightarrow \mathbb{R}$

$\mathbb{P} \{ |f(X_1, \dots, X_n) - \mathbb{E} f(X_1, \dots, X_n)| \geq t \} \leq ?$

-  $f$  is not too sensitive to each  $x_i$

$$x^n = (x_1, \dots, x_n) = (\underbrace{x_1, \dots, x_{i-1}}_{x^{i-1}}, x_i, \underbrace{x_{i+1}, \dots, x_n}_{x_{i+1}^n})$$

Discrete gradient:

$$\Delta_i f(x_1, \dots, x_n) = \sup_{x_i} f(x^{i-1}, x_i, x_{i+1}^n) - \inf_{x_i} f(x^{i-1}, x_i, x_{i+1}^n)$$

Then  $f$  has bounded differences when

$$\underbrace{\sup_{x^n \in \mathcal{X}^n} \Delta_i f(x_1, \dots, x_n)}_{\|\Delta_i f\|_\infty} \leq c_i < \infty \quad \text{for all } i.$$

Thm (McDiarmid's inequality) If  $f: \mathcal{X}^n \rightarrow \mathbb{R}$  has bounded differences, then, for  $X_1, \dots, X_n$  independent,

$$\mathbb{P} \left\{ f(X^n) - \mathbb{E} f(X^n) \geq \epsilon \right\} \leq \exp \left( - \frac{2\epsilon^2}{\sum_{i=1}^n \|\Delta_i f\|_\infty^2} \right)$$

$$\mathbb{P} \left\{ f(X^n) - \mathbb{E} f(X^n) \leq -\epsilon \right\} \leq \exp \left( - \frac{2\epsilon^2}{\sum_{i=1}^n \|\Delta_i f\|_\infty^2} \right)$$

Note:  $\|\Delta_i f\|_\infty$  measures the sensitivity of  $f$  to  $i$ th input.

## Examples

1)  $f(x_1, \dots, x_n) = x_1 + \dots + x_n \quad a_i \leq x_i \leq b_i$

$$\Delta_i f(x^n) = \sup_{x_i \in (a_i, b_i)} (x_1 + \dots + x_i + \dots + x_n) - \inf_{x_i' \in (a_i, b_i)} (x_1 + \dots + x_i' + \dots + x_n)$$

$$= \sup_{x_i \in (a_i, b_i)} \sup_{x_i' \in (a_i, b_i)} (x_i - x_i') = b_i - a_i$$

$$\Rightarrow \mathbb{P} \left\{ \left| \sum_{i=1}^n X_i - \sum_{i=1}^n \mathbb{E} X_i \right| \geq \epsilon \right\} \leq 2 \exp \left( - \frac{2\epsilon^2}{\sum_{i=1}^n (b_i - a_i)^2} \right)$$

2) Uniform deviations of empirical means

$$X_1, \dots, X_n \stackrel{\text{iid}}{\sim} P$$

$$P_n(A) := \frac{1}{n} \sum_{i=1}^n \mathbb{1}_{\{X_i \in A\}} \quad - \text{empirical frequency of } A$$

$$\mathbb{E} P_n(A) = P(A)$$

$0 \leq P_n(A) \leq 1$ , function of  $X_1, \dots, X_n$

$$P \left\{ |P_n(A) - P(A)| \geq \epsilon \right\} \leq 2e^{-2n\epsilon^2} \text{ (Hoeffding)}$$

$$f(X_1, \dots, X_n) = \sup_{A \in \mathcal{A}} |P_n(A) - P(A)|$$

(see this later in binary classification)

$$= \sup_{A \in \mathcal{A}} \left| \frac{1}{n} \sum_{i=1}^n \mathbb{1}_{\{X_i \in A\}} - P(A) \right|$$

Later: if  $\mathcal{A}$  is not too "rich,"  $\mathbb{E} f(X^n) \sim \frac{C}{\sqrt{n}}$

$$P \left\{ \sup_{A \in \mathcal{A}} |P_n(A) - P(A)| \geq \mathbb{E} \sup_{A \in \mathcal{A}} |P_n(A) - P(A)| + \epsilon \right\} \leq e^{-2n\epsilon^2}$$

Claim:  $f(x^n) = \sup_{A \in \mathcal{A}} \left| \frac{1}{n} \sum_{i=1}^n \mathbb{1}_{\{X_i \in A\}} - P(A) \right|$

$$\|\Delta_i f\|_\infty \leq 1/n \quad \forall i$$

$$\sum_{i=1}^n \|\Delta_i f\|_\infty^2 \leq \frac{1}{n^2} \cdot n = \frac{1}{n}$$

$$\Rightarrow \exp\left(-\frac{2\epsilon^2}{\sum_{i=1}^n \|\Delta_i f\|_\infty^2}\right) \leq \exp(-2n\epsilon^2)$$

$$\Delta_i f(x^n) = \sup_{x_i} \sup_{A \in \mathcal{A}} \left| \frac{1}{n} \sum_{j=1}^n \mathbb{1}_{\{x_j \in A\}} - P(A) \right|$$

$$- \inf_{x_i} \sup_{A \in \mathcal{A}} \left| \frac{1}{n} \sum_{j=1}^n \mathbb{1}_{\{x_j \in A\}} - P(A) \right|$$

$$\left| \frac{1}{n} \sum_{j=1}^n \mathbb{1}_{\{x_j \in A\}} - P(A) \right|$$

$$= \left| \frac{1}{n} \sum_{j \neq i} \mathbb{1}_{\{x_j \in A\}} + \frac{1}{n} \mathbb{1}_{\{x_i \in A\}} - P(A) \right|$$

$$\left| \frac{1}{n} \sum_{j \neq i} \mathbb{1}_{\{x_j \in A\}} + \frac{1}{n} \mathbb{1}_{\{x_i \in A\}} - P(A) \right|$$

$$- \left| \frac{1}{n} \sum_{j \neq i} \mathbb{1}_{\{x_j \in A\}} + \frac{1}{n} \mathbb{1}_{\{x_i' \in A\}} - P(A) \right|$$

$$|a-b| \leq |a-c| + |c-b|$$

$$\leq \frac{1}{n} \left| \mathbb{1}_{\{x_i \in A\}} - \mathbb{1}_{\{x_i' \in A\}} \right| \leq \frac{1}{n}.$$