

$$1.1) \quad \mathbb{E}[X f(X)] = -\frac{\sigma^2}{\sqrt{2\pi\sigma^2}} \int_{-\infty}^{\infty} f(x) \left(-\frac{x}{\sigma^2}\right) \exp\left(-\frac{x^2}{2\sigma^2}\right) dx.$$

$$= -\sigma^2 \int_{-\infty}^{\infty} f(x) \frac{d}{dx} \left( \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{x^2}{2\sigma^2}\right) \right) dx.$$

integration by parts

$$= -\sigma^2 \left[ \left[ f(x) \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{x^2}{2\sigma^2}\right) \right]_{-\infty}^{\infty} - \int_{-\infty}^{\infty} f'(x) \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{x^2}{2\sigma^2}\right) dx \right]$$

$$= \sigma^2 \int_{-\infty}^{\infty} f'(x) \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{x^2}{2\sigma^2}\right) dx = \sigma^2 \mathbb{E}[f'(X)].$$

1.2) Let  $p = 2n$ ,

$$\mathbb{E}[X X^{2n-1}] = \sigma^2 \mathbb{E}\left[\frac{d}{dx} X^{2n-1}\right]$$

$$= \sigma^2 (2n-1) \mathbb{E}[X^{2n-2}]$$

$$\vdots$$

$$= \sigma^{2n} (2n-1)!! = \sigma^{2n} \frac{2n!}{2^n \cdot n!}$$

By Stirling approximation  $(2n)! \sim \sqrt{2\pi(2n)} \left(\frac{2n}{e}\right)^{2n}$  and  $n! \sim \sqrt{2\pi n} \left(\frac{n}{e}\right)^n$

we have

$$(\mathbb{E}[x^{2n}])^{\frac{1}{2n}} \leq C \sigma \sqrt{p}$$

$$\begin{aligned} 1.3) \quad (\mathbb{E} |x|^p)^{\frac{1}{p}} &\stackrel{\text{Monotonicity}}{\leq} (\mathbb{E} |x|^{p+1})^{\frac{1}{p+1}} \stackrel{(2)}{\leq} C \cdot \sqrt[p+1]{p} \\ &\leq C \cdot \sigma \sqrt{2p} \\ &\leq \sqrt{2} \cdot C \cdot \sigma \sqrt{p}. \end{aligned}$$

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$$\begin{aligned} 2.1 \quad \|\Sigma^+\|_F^2 &= \sqrt{\text{Tr}(\Sigma^+)^T \Sigma^+}^2 \\ &= \text{Tr}((\Sigma^+)^T \Sigma^+) \end{aligned}$$

Note that  $\Sigma$  has full rank with probability 1 and  $\Sigma \Sigma^T$  is invertible

with probability 1 (both facts can be proved by considering the

determinant.) Therefore

$$\text{Tr}((\Sigma^+)^T \Sigma^+) = \text{Tr}((\Sigma \Sigma^T)^+) = \text{Tr}((\Sigma \Sigma^T)^{-1})$$

w.p. 1

$$2.2. \quad \mathbb{E} \|\Omega^+\|_F^2 = \mathbb{E} \left( \sum x_i^{-1} \right) = \sum \mathbb{E}(x_i^{-1}) = \frac{m}{n-m-1}$$

$$2.3. \quad \left( \mathbb{E} \|\Omega^+\|_F^{2g} \right)^{1/g} \leq \sum \left( \mathbb{E}[x_i^{-g}] \right)^{1/g} \leq \frac{3m}{n-m+1}$$

$$2.4. \quad \Pr \left[ \|\Omega^+\|_F^2 \geq t \right] \leq \frac{\mathbb{E}(\|\Omega^+\|_F^{2g})}{t^g}$$

$$\Rightarrow \Pr \left[ \|\Omega^+\|_F^2 \geq \frac{3m}{n-m+1} u \right] \leq u^{-g}$$

$$3.1. \quad \Pr(|e_1^T v_0| \leq \varepsilon) = \Pr(-\varepsilon \leq e_1^T v_0 \leq \varepsilon) = 2 \Pr(0 \leq e_1^T v_0 \leq \varepsilon)$$

$$= 2 \int_0^\varepsilon C_{1,n} (1-y)^{n-3/2}$$

$$\leq 2\varepsilon C_{1,n} \leq \sqrt{\frac{2n}{\pi}} \varepsilon$$

$$\begin{aligned}
 4. \quad \Pr ( \|A\|_2 > \varepsilon \|Ax\|_2 ) &= \Pr ( \|A\|_2^2 > \varepsilon^2 \|Ax\|_2^2 ) \\
 &= \Pr ( \|A^T A\|_2 > \varepsilon^2 x^T A^T A x )
 \end{aligned}$$

Consider the spectral decomposition  $A^T A = U \Lambda U^T = U \begin{bmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{bmatrix} U^T$

with  $\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_n \geq 0$ .

$$\varepsilon^2 x^T A^T A x = \varepsilon^2 \sum_{j=1}^n \lambda_j (x^T u_j)^2 \geq \varepsilon^2 \lambda_1 (x^T u_1)^2.$$

$$\Rightarrow \Pr \{ \|A^T A\|_2 > \varepsilon^2 x^T A^T A x \} \leq \Pr \{ \lambda_1 > \varepsilon^2 \lambda_1 (x^T u_1)^2 \}$$

$$= \Pr \left\{ (x^T u_1)^2 < \frac{1}{\varepsilon^2} \right\}$$

by Lecture 1  
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$$\leq \sqrt{\frac{2}{n}} \leq \frac{1}{\varepsilon}.$$

$$\Rightarrow \Pr ( \|A\|_2 \leq \varepsilon \|Ax\|_2 ) \geq 1 - \sqrt{\frac{2}{n}} \leq \frac{1}{\varepsilon}$$