

# Notes of [Paul, 2007]

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## Abstract

*Key words:* High dimension, Huber loss.

## 1 Weak concentration inequalities for quadratic forms

**Lemma 1.** *Suppose  $X$  and  $Y$  are iid  $N_n(0, I_n)$ ,  $C$  is a  $n \times n$  matrix,  $\|C\|$  is the largest singular value of  $C$ . Then for  $0 < \delta < 1$ ,*

$$\Pr\left(\frac{1}{n}|X^T CY| > t\right) \leq 2 \exp\left(-\frac{(1-\delta^2)nt^2}{2\|C\|^2}\right), \quad \text{for } 0 \leq t \leq \frac{\delta}{1-\delta^2}\|C\|.$$

*Proof.* For  $\lambda \leq \frac{1}{\|C\|}$ , we have

$$\begin{aligned} \Pr(X^T CY > \lambda nt) &\leq \exp(-nt) \mathbb{E} \exp(\lambda X^T CY) \\ &= \exp(-\lambda nt) \mathbb{E} \exp\left(\frac{\lambda^2}{2}\|CY\|^2\right) \\ &= \exp(-\lambda nt) |I_n - \lambda^2 C^T C|^{-1/2}. \end{aligned} \tag{1}$$

Now, use the fact that  $\log |I_n - \lambda^2 C^T C| = \sum_{i=1}^n \log(1 - \lambda^2 \lambda_i(C^T C))$ . We have

$$\begin{aligned} -\log |I_n - \lambda^2 C^T C| &= \sum_{i=1}^n -\log(1 - \lambda^2 \lambda_i(C^T C)) \\ &\leq -n \log(1 - \lambda^2 \|C\|^2) \\ &\leq n \frac{\lambda^2 \|C\|^2}{1 - \lambda^2 \|C\|^2}. \end{aligned}$$

Hence with (1), for  $0 < \delta < 1$  and  $0 \leq \lambda \leq \frac{\delta}{\|C\|}$ , we have,

$$\begin{aligned} \Pr(X^T CY > \lambda nt) &\leq \exp(-n\lambda t + \frac{n}{2} \frac{\lambda^2 \|C\|^2}{1 - \lambda^2 \|C\|^2}) \\ &= \exp \left\{ n(-\lambda \|C\| \frac{t}{\|C\|} + \frac{1}{2} \frac{(\lambda \|C\|)^2}{1 - (\lambda \|C\|)^2}) \right\} \quad (2) \\ &= \exp \left\{ n(-\lambda \|C\| \frac{t}{\|C\|} + \frac{1}{2} \frac{(\lambda \|C\|)^2}{1 - \delta^2}) \right\}. \end{aligned}$$

The last expression is a quadratic form of  $\lambda \|C\|$ , it takes it's minimum when  $\lambda \|C\| = \frac{1-\delta^2}{\|C\|} t$ . To make this value fall in the interval  $(0, \delta)$ , we require  $t \leq \frac{\delta}{1-\delta^2} \|C\|$ . Thus, for  $0 \leq t \leq \frac{\delta}{1-\delta^2} \|C\|$ , we have

$$\begin{aligned} \Pr(X^T CY > \lambda nt) &\leq \exp \left\{ n(-\frac{1-\delta^2}{\|C\|} t \frac{t}{\|C\|} + \frac{1}{2} \frac{(\frac{1-\delta^2}{\|C\|} t)^2}{1 - \delta^2}) \right\} \\ &= \exp \left\{ n(-\frac{1-\delta^2}{\|C\|} t \frac{t}{\|C\|} + \frac{1}{2} \frac{(\frac{1-\delta^2}{\|C\|} t)^2}{1 - \delta^2}) \right\} \\ &= \exp \left\{ -\frac{n}{2} \frac{1-\delta^2}{\|C\|^2} t^2 \right\}. \end{aligned}$$

□

**Lemma 2.** Suppose  $X$  is distributed as  $N_n(0, I_n)$ . Then for  $0 < \delta < 1$ ,

$$\Pr(\frac{1}{n}(X^T CX - \text{tr } C) > t) \leq 2 \exp(-\frac{\delta nt^2}{4\|C\|^2}) \quad \text{for } t \leq \frac{\|C\|(1-\delta)}{\delta}.$$

*Proof.*

$$\begin{aligned} \Pr(\frac{1}{n}(X^T CX - \text{tr } C) > t) &= \Pr(\lambda X^T CX > \lambda(nt + \text{tr } C)) \\ &\leq \exp(-\lambda(nt + \text{tr } C)) \mathbb{E}(\exp(\lambda X^T CX)) \\ &= \exp(-\lambda(nt + \text{tr } C)) |I_n - 2\lambda C|^{-1/2} \\ &= \exp \left( -\lambda(nt + \text{tr } C) - \frac{1}{2} \log |I_n - 2\lambda C| \right) \\ &\leq \exp \left( -\lambda nt - \lambda \text{tr } C - \frac{1}{2} \log |I_n - 2\lambda C| \right). \end{aligned}$$

We have the inequality

$$-\log(1-u) - u \leq \frac{u^2}{2(1-|u|)} \quad \text{for } u \in (-1, 1).$$

Hence

$$\Pr\left(\frac{1}{n}(X^T C X - \text{tr } C) > t\right) \leq \exp\left(-\lambda n t + \frac{1}{2} \sum_{i=1}^n \frac{(2\lambda\lambda_i(C))^2}{2(1 - 2|\lambda\lambda_i(C)|)}\right).$$

For  $\lambda \leq \frac{1-\delta}{2\|C\|}$

$$\begin{aligned} -\lambda n t + \frac{1}{2} \sum_{i=1}^n \frac{(2\lambda\lambda_i(C))^2}{2(1 - 2|\lambda\lambda_i(C)|)} &\leq -\lambda n t + \frac{1}{2} \sum_{i=1}^n \frac{(2\lambda\lambda_i(C))^2}{2\delta} \\ &\leq -\lambda n t + \frac{n\lambda^2\|C\|^2}{\delta} \end{aligned}$$

For  $t \leq \frac{\|C\|(1-\delta)}{\delta}$ , we let  $\lambda = \frac{\delta t}{2\|C\|^2}$ , then

$$-\lambda n t + \frac{n\lambda^2\|C\|^2}{\delta} \leq -\frac{\delta n t^2}{4\|C\|^2}$$

□

## 2 Concentration bound for the eigenvalues of wishart matrix

**Proposition 1.** *Let  $\mathbf{Z}$  be  $N \times n$  random matrix with all elements iid distributed as  $N(0,1)$ . Assume  $N/n \rightarrow \gamma \in (0,1)$ . Let  $\kappa_\gamma = (1 + \sqrt{\gamma})^2$ . Then for any  $0 < \delta < \kappa_\gamma/2$ ,*

$$\Pr(|\lambda_1(\frac{1}{n}\mathbf{Z}\mathbf{Z}^T) - \kappa_\gamma| \geq \delta) \leq 2 \exp\left(-\frac{n\delta^2}{32\kappa_\gamma}\right), \quad \text{for } n \geq n_0(\gamma, \delta)$$

where  $n_0(\gamma, \delta)$  is an integer large enough such that  $|\text{Med}(\lambda_1(\frac{1}{n}\mathbf{Z}\mathbf{Z}^T)) - \kappa_\gamma| \leq \delta/4$  for  $n \geq n_0(\gamma, \delta)$ .

## References

## References

[Paul, 2007] Paul, D. (2007). Asymptotics of sample eigenstruture for a large dimensional spiked covariance model. *Statistica Sinica*, 17:1617.