

For this quiz, use the following results. A) Suppose  $\mathbf{X} \sim N_k(\boldsymbol{\mu}, \boldsymbol{\Sigma})$ , then

- i)  $\mathbf{AX} \sim N_q(\mathbf{A}\boldsymbol{\mu}, \mathbf{A}\boldsymbol{\Sigma}\mathbf{A}^T)$ .
- ii)  $\mathbf{a} + b\mathbf{X} \sim N_k(\mathbf{a} + b\boldsymbol{\mu}, b^2\boldsymbol{\Sigma})$ .
- iii)  $\mathbf{AX} + \mathbf{d} \sim N_q(\mathbf{A}\boldsymbol{\mu} + \mathbf{d}, \mathbf{A}\boldsymbol{\Sigma}\mathbf{A}^T)$ .

(Find the mean and covariance matrix of the left hand side and plug in those values for the right hand side. **Be careful with the dimension  $k$  or  $q$ .**)

B) Suppose  $\mathbf{X}_n \xrightarrow{D} N_k(\boldsymbol{\mu}, \boldsymbol{\Sigma})$ . Then

- i)  $\mathbf{AX}_n \xrightarrow{D} N_q(\mathbf{A}\boldsymbol{\mu}, \mathbf{A}\boldsymbol{\Sigma}\mathbf{A}^T)$ .
- ii)  $\mathbf{a} + b\mathbf{X}_n \xrightarrow{D} N_k(\mathbf{a} + b\boldsymbol{\mu}, b^2\boldsymbol{\Sigma})$ .
- iii)  $\mathbf{AX}_n + \mathbf{d} \xrightarrow{D} N_q(\mathbf{A}\boldsymbol{\mu} + \mathbf{d}, \mathbf{A}\boldsymbol{\Sigma}\mathbf{A}^T)$ .

(The behavior of convergence in distribution to a MVN distribution is much like the behavior of the MVN distributions in A.)

1) By the OLS CLT,  $\sqrt{n}(\hat{\boldsymbol{\beta}} - \boldsymbol{\beta}) \xrightarrow{D} N_p(\mathbf{0}, \sigma^2 \mathbf{W})$ . Hence the limiting distribution of  $\sqrt{n}(\hat{\boldsymbol{\beta}} - \boldsymbol{\beta})$  is the  $N_p(\mathbf{0}, \sigma^2 \mathbf{W})$  distribution. Let  $\mathbf{A}$  be a constant  $r \times p$  matrix. Find the limiting distribution of  $\mathbf{A}\sqrt{n}(\hat{\boldsymbol{\beta}} - \boldsymbol{\beta})$ .

$$\xrightarrow{D} N_r(\mathbf{0}, \sigma^2 \mathbf{A} \mathbf{W} \mathbf{A}^T)$$

2) Suppose  $\mathbf{Z}_n \xrightarrow{D} N_k(\boldsymbol{\mu}, \mathbf{I})$ . Let  $\mathbf{A}$  be a constant  $r \times k$  matrix. Find the limiting distribution of  $\mathbf{A}(\mathbf{Z}_n - \boldsymbol{\mu})$ .

$$\xrightarrow{D} N_r(\mathbf{0}, \mathbf{A} \mathbf{A}^T)$$

3) For the parametric bootstrap,  $\mathbf{Y}^* = \mathbf{X}\hat{\boldsymbol{\beta}} + \mathbf{e}^*$  where  $\mathbf{e}^* = (e_1^*, \dots, e_m^*)^T$  and the  $e_i^*$  are iid  $N(0, \sigma_n^2)$  random variables. Assume this model satisfies the OLS CLT as  $m \rightarrow \infty$  where the OLS estimator is  $\hat{\boldsymbol{\beta}}^*$  and  $\boldsymbol{\beta}$  is replaced by  $\hat{\boldsymbol{\beta}}$ . Note that  $\sigma^2$  is replaced by  $\hat{\sigma}_n^2$ . Assume  $\hat{\boldsymbol{\beta}}$  is a  $p \times 1$  vector and  $\mathbf{Y}^*$  is an  $m \times 1$  vector. a) Then use the OLS CLT to find the limiting distribution of  $\sqrt{m}(\hat{\boldsymbol{\beta}}^* - \hat{\boldsymbol{\beta}})$  as  $m \rightarrow \infty$ .

$$\xrightarrow{D} N_p(\mathbf{0}, \sigma_n^2 \mathbf{W})$$

$$\hat{\Sigma}_n \rightarrow \sigma^2 \mathbf{W}$$

b) In a) you should get  $\sqrt{m}(\hat{\boldsymbol{\beta}}^* - \hat{\boldsymbol{\beta}}) \xrightarrow{D} N_p(\mathbf{0}, \boldsymbol{\Sigma}_n)$  where  $\boldsymbol{\Sigma}_n \xrightarrow{P} \boldsymbol{\Sigma}$ . The bootstrap proof technique says that suppose  $\sqrt{n}(T_n - \boldsymbol{\theta}) \xrightarrow{D} N_g(\mathbf{0}, \boldsymbol{\Sigma})$  and  $\boldsymbol{\Sigma}_n \xrightarrow{P} \boldsymbol{\Sigma}$  as  $n \rightarrow \infty$ , and for fixed  $n$ ,  $\sqrt{m}(T_{n,m}^* - T_n) \xrightarrow{D} N_g(\mathbf{0}, \boldsymbol{\Sigma}_n)$  as  $m \rightarrow \infty$ . Then  $\sqrt{n}(T_n^* - T_n) \xrightarrow{D} N_g(\mathbf{0}, \boldsymbol{\Sigma})$  as  $n \rightarrow \infty$ . Find the limiting distribution  $\sqrt{n}(\hat{\boldsymbol{\beta}}^* - \hat{\boldsymbol{\beta}})$  as  $n \rightarrow \infty$  where you may plug in  $\boldsymbol{\Sigma}$  in to the result (you do not need to compute  $\boldsymbol{\Sigma}$ ).

$$\sqrt{n}(\hat{\boldsymbol{\beta}}^* - \hat{\boldsymbol{\beta}}) \xrightarrow{D} N_p(\mathbf{0}, \hat{\Sigma})$$

$$\sim N_p(\mathbf{0}, \sigma^2 \mathbf{W})$$

either