

Homework 5: Suggested Solutions

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5.4

Show that $\hat{\theta} = \arg \max_{\theta \in \Theta} \ell_n(\theta) = \arg \max_{\theta \in \Theta} \mathcal{L}_n(\theta)$.

Proof. Since $\mathcal{L}_n(\theta)$ is merely a monotonic transformation of $\ell_n(\theta)$ by definition, maximizers will agree for this two objective functions \square

5.5

For the regression in-sample predicted values \hat{y}_i show that $\hat{y}_i | \mathbf{X} \sim \mathcal{N}(\mathbf{x}_i' \boldsymbol{\beta}, \sigma^2 h_{ii})$ where h_{ii} are leverage values (3.41)

Proof. From **Theorem 5.4**,

$$\hat{\boldsymbol{\beta}} \sim \mathcal{N}(\boldsymbol{\beta}, \sigma^2 (\mathbf{X}'\mathbf{X})^{-1}).$$

Then conditional on \mathbf{X} , \mathbf{x}_i' becomes constant for each i . By **Theorem 5.2** of Hansen's notes, it follows that $\hat{y}_i = \mathbf{x}_i' \hat{\boldsymbol{\beta}}$ follows a normal distribution with mean $\mathbf{x}_i' \boldsymbol{\beta}$ and variance $\sigma^2 \mathbf{x}_i' (\mathbf{X}'\mathbf{X})^{-1} \mathbf{x}_i$, i.e.

$$\hat{y}_i | \mathbf{X} \sim \mathcal{N}(\mathbf{x}_i' \boldsymbol{\beta}, \sigma^2 h_{ii})$$

where $h_{ii} = \mathbf{x}_i' (\mathbf{X}'\mathbf{X})^{-1} \mathbf{x}_i$ by (3.41). \square

5.6

In the normal regression model, show that the leave-one out prediction errors \tilde{e}_i and the standardized residuals \bar{e}_i are independent of $\hat{\boldsymbol{\beta}}$, conditional on \mathbf{X} . (Hint: Use (3.46) and (4.24)).

Proof. By **Theorem 5.6**, $\hat{\mathbf{e}}$ is independent of $\hat{\boldsymbol{\beta}}$. From (3.46) and (4.24), since \mathbf{M}^* and \mathbf{M} become constant conditional on \mathbf{X} , $\tilde{\mathbf{e}}$ and $\bar{\mathbf{e}}$ are also independent of $\hat{\boldsymbol{\beta}}$. \square

5.7

In the normal regression model show that the robust covariance matrix $\hat{V}_{\hat{\beta}}^{HC0}$, $\hat{V}_{\hat{\beta}}^{HC1}$, $\hat{V}_{\hat{\beta}}^{HC2}$ and $\hat{V}_{\hat{\beta}}^{HC3}$ are independent of the OLS estimator $\hat{\beta}$, conditional on \mathbf{X} .

Proof. Since the randomness of those four variance estimators is completely sourced from \hat{e}_i^2 for each i when \mathbf{X} is controlled, they must be also independent of $\hat{\beta}$ (because \hat{e} is independent of $\hat{\beta}$.) □