## Problem Set \# 2

(1) The following data were generated by a Weibull distribution: ${ }^{1}$

| 1.3043 | 0.4895 | 1.2742 | 1.4019 | 0.3255 | 0.2996 | 0.2642 |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- |
| 1.0878 | 1.9461 | 0.4761 | 3.6454 | 0.1534 | 1.2357 | 0.9638 |
| 0.3345 | 1.1227 | 2.0296 | 1.2797 | 0.9608 | 2.0070 |  |

(a) Obtain the maximum likelihood estimates of $\alpha$ and $\beta$, and give the asymptotic covariance matrix for the estimates.
(b) Carry out the Wald test of the hypothesis that $\beta=1$.
(c) Obtain the maximum likelihood estimates of $\alpha$ under the hypothesis that $\beta=1$.
(d) Using the results of parts (a) and (b), carry out a likelihood ratio test of the hypothesis that $\beta=1$.
(e) Carry out a Lagrange multiplier test of the hypothesis that $\beta=1$.
(2) For the classical regression model $\boldsymbol{y}=\boldsymbol{X} \boldsymbol{\beta}+\boldsymbol{\varepsilon}$ with no constant term and $K$ regressors, what is

$$
\operatorname{plim} \frac{R^{2} / K}{\left(1-R^{2}\right) /(n-K)},
$$

assuming that the true value of $\beta$ is zero?
(3) Prove that, under the hypothesis that $\boldsymbol{R} \boldsymbol{\beta}=\boldsymbol{r}$, the estimator

$$
s_{R}^{2}=\frac{\left(\boldsymbol{y}-\boldsymbol{X} \widehat{\boldsymbol{\beta}}_{R}\right)^{\prime}\left(\boldsymbol{y}-\boldsymbol{X} \widehat{\boldsymbol{\beta}}_{R}\right)}{n-K+J},
$$

where $J$ is the number of restrictions, is unbiased for $\sigma^{2}$.
(4) What is the covariance matrix of the GLS estimator and the difference between it and the OLS estimator?

$$
\operatorname{Cov}\left(\widehat{\boldsymbol{\beta}}_{G L S}, \widehat{\boldsymbol{\beta}}_{G L S}-\widehat{\boldsymbol{\beta}}_{O L S}\right),
$$

where $\widehat{\boldsymbol{\beta}}_{G L S}=\left(\mathbf{X}^{\prime} \boldsymbol{\Omega}^{-1} \mathbf{X}\right)^{-1} \mathbf{X}^{\prime} \boldsymbol{\Omega}^{-1} \boldsymbol{y}$, and $\widehat{\boldsymbol{\beta}}_{O L S}=\left(\mathbf{X}^{\prime} \mathbf{X}\right)^{-1} \mathbf{X}^{\prime} \boldsymbol{y}$.
(5) Suppose that $y$ has the following density function:

$$
f(y \mid \boldsymbol{x})=\frac{e^{-y / \boldsymbol{\beta}^{\prime} \boldsymbol{x}}}{\boldsymbol{\beta}^{\prime} \boldsymbol{x}}, y>0
$$

Then $\mathrm{E}[y \mid \boldsymbol{x}]=\boldsymbol{\beta}^{\prime} \boldsymbol{x}$ and $\operatorname{Var}[y \mid \boldsymbol{x}]=\left(\boldsymbol{\beta}^{\prime} \boldsymbol{x}\right)^{2}$. For this model, prove that GLS and MLE are the same.

[^0](6) Assume that the model
$$
y^{*}=\beta x^{*}+\varepsilon
$$
conform to all of the assumptions of the classical normal regression model. If data on $y^{*}$ and $x^{*}$ were available, all the apparatus of the classical model would apply. ${ }^{2}$ We assume that
$$
y=y^{*}+v \quad \text { with } v \sim \mathcal{N}\left(0, \sigma_{v}^{2}\right)
$$
and
$$
x=x^{*}+u \quad \text { with } u \sim \mathcal{N}\left(0, \sigma_{u}^{2}\right) .
$$
(a) Prove that when only $x^{*}$ is measured with error, the squared correlation between $y$ and $x$ is less than that between $y^{*}$ and $x^{*}$.
(b) Does the same hold true if $y^{*}$ is also measured with error?

[^1]
[^0]:    ${ }^{1}$ The density function of a Weibull distribution is $f(x)=\alpha \beta x^{\beta-1} e^{-\alpha x^{\beta}}$ for $x>0$ and $\alpha, \beta>0$.

[^1]:    ${ }^{2}$ Notice that here $*$ denotes unobserved variables.

