ECE531 Spring 2013 Quiz 12

Your Name: _ SOLUTION

Instructions: This quiz is worth a total of 100 points. The quiz is open book and open notes. You may also use a calculator. You may not use a computer, phone, or tablet. Please show your work on each problem and box/circle your final answers. Points may be deducted for a disorderly presentation of your solution.

Suppose you wish to detect the presence of a temperature trend in Worcester, e.g. an indicator of global warming, by analyzing average monthly temperature data from Worcester over the last 30 years. A plot of the actual average monthly temperatures in Worcester is shown in Figure 1 below.

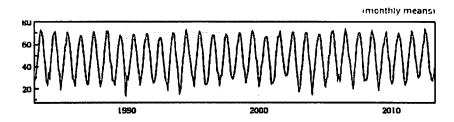


Figure 1: Average monthly temperature in Worcester since 1983 (from Wolfram Alpha).

Each data sample is modeled as

$$Y_k = a + bk + c\sin(\omega_0 k + \phi) + W_k$$

for k = -n, ..., n corresponding to a month index with ω_0 and ϕ known and $W \sim \mathcal{N}(0, \sigma^2 I)$ with σ^2 known. The parameters a, b, and c are all assumed unknown.

- 1. 20 points. Briefly explain the physical significance of the parameters a, b, and c.
- 2. 80 points. Suppose the hypotheses are

$$\mathcal{H}_0$$
: $b=0$
 \mathcal{H}_1 : $b \neq 0$

$$\mathcal{H}_1 : b \neq 0$$

Find the GLRT decision statistic (using approximations as appropriate) and describe how you would find the decision threshold so that the false positive probability constraint $P_{\mathsf{fp}} \leq \alpha$ is satisfied. Hint:

If
$$G = \begin{bmatrix} d & 0 & 0 \\ 0 & e & f \\ 0 & f & g \end{bmatrix}$$
 then $G^{-1} = \begin{bmatrix} 1/d & 0 & 0 \\ 0 & g/r & -f/r \\ 0 & -f/r & e/r \end{bmatrix}$ with $r = eg - f^2$

(1

II a: models the average temperature over the 30 year data record.

b: models a linear temperature trend over the data

c: Models the effect of seasonality [Since we the temperatures should be periodic over one year, we would expect $12 \, \omega_0 = 2 \, \mathrm{T}$, so $\omega_0 = T/6$]

Need to apply theorem 7.1 with $A = [0 \mid 0]$ and b = 0. Let $0 = \begin{bmatrix} a \\ b \\ c \end{bmatrix}$ then $H = \begin{bmatrix} 1 - n & sin(-nw_0 + \phi) \end{bmatrix}$ $\begin{bmatrix} 1 & n & sin(nw_0 + \phi) \end{bmatrix}$

Then $H^TH = \begin{bmatrix} 2n+1 & 0 & \approx 0 \\ 0 & 2\sum_{k=1}^{n} k^2 & \beta \\ \approx 0 & \beta & n \end{bmatrix}$

Since $\sum_{k=-n}^{n} \sin^{2}(k\omega_{0} + \phi) = \frac{1}{2} \sum_{k=-n}^{n} (1 - \cos(k2\omega_{0} + 2\phi))$ $\approx \frac{2n+1}{2} \approx n \quad \text{for } n \text{ su Afferently large}$

and with $\beta = \sum_{k=-n}^{\infty} K \sin(kw_0 + \phi)$ (this is not small, in general)

Hence $(H^TH)^{\frac{1}{2}}$ $\left[\begin{array}{c} \frac{1}{2n-1} & 0 & 0 \\ 0 & \frac{n}{r} & -\frac{\beta}{r} \\ 0 & -\frac{\beta}{r} & \frac{2\hat{\Sigma}k^2}{r} \end{array}\right]$ with $r = 2n\hat{\Sigma}k^2 - \beta^2$

continued ...

so now apply the theorem ...

$$A\hat{\theta}_{1}-b \quad A(H^{T}H)^{-1}H^{T}y$$

$$= \left[0 \quad \hat{\gamma} \quad -\frac{\beta}{\gamma}\right]H^{T}y^{-1} = S^{T}H^{T}y^{-1}$$

$$A(H^{T}H)^{-1}A^{T} = \hat{\gamma}$$

$$S_{0}$$

$$T(y) = \frac{\left(s^{T}H^{T}y\right)^{T} \cdot \Gamma\left(s^{T}H^{T}y\right)}{f^{2}} = \frac{\left(\Gamma_{n}\right)y^{T}H \cdot ss^{T}H^{T}y}{f^{2}}$$

$$SS^{T} = \begin{bmatrix} 0 & 0 & 0 \\ 0 & (\frac{n}{r})^{2} & -\frac{\beta n}{r^{2}} \\ 0 & -\frac{\beta n}{r^{2}} & \frac{\beta^{2}}{r^{2}} \end{bmatrix}$$

$$\frac{\Gamma}{n} s s^{T} = \begin{bmatrix} 0 & 0 & 0 \\ 0 & \frac{1}{r} & -\frac{\beta}{r} \\ 0 & -\frac{\beta}{r} & \frac{\beta^{2}}{nr} \end{bmatrix} = \Lambda$$
with $\beta = \sum_{k=-n}^{n} K sin(kw_{0} + \phi)$
and $r = 2n \sum_{k=1}^{n} K^{2} - \beta^{2}$

so
$$T(y) = \frac{y^T H \Lambda H^T y}{\sigma^2}$$
 is the decision statistic.

Note: the zeros in A effectively cause T(y) to ignore the constant component in the observation

To determine the decision threshold, we use the tail probability of a x2 random variable with 1 degree of freedom.

But a X2 random variable with 1 degree of freedom is equivalent to a squared standard Normal random variable

Hence
$$P_{fp} = 2Q(V) = \lambda \Rightarrow V = (Q^{-1}(\frac{\omega}{2}))^2$$