EXAMINATIONS OF THE ROYAL STATISTICAL SOCIETY

GRADUATE DIPLOMA, 2015

MODULE 3 : Stochastic processes and time series

Time allowed: Three hours

Candidates should answer FIVE questions.

All questions carry equal marks.
The number of marks allotted for each part-question is shown in brackets.

Graph paper and Official tables are provided.

Candidates may use calculators in accordance with the regulations published in the Society's "Guide to Examinations" (document Ex1).

The notation log denotes logarithm to base e.
Logarithms to any other base are explicitly identified, e.g. log_{10}.

Note also that \( \binom{n}{r} \) is the same as \( ^nC_r \).
1. (a) In the context of a time-homogeneous Markov chain with states $i = 0, 1, 2, \ldots$, define the $n$-step transition probability $p_{ij}^{(n)}$. Derive the Chapman-Kolmogorov equations for the $p_{ij}^{(n)}$ indicating clearly at which step(s) you make use of (i) time-homogeneity, (ii) the Markov property.

(b) A software development company is conducting competency challenges, comprising a sequence of independent debugging exercises. Each programmer must attempt the exercises successively until awarded a Certificate of Programming Competency, for which either a perfect score on one exercise or a score of at least 90% on two consecutive exercises is required.

This process can be modelled as a Markov chain with three states which depend on the most recent results as follows.

- State 0: score under 90% on most recent attempt
- State 1: score between 90% and 99% on most recent attempt, but under 90% on the previous attempt (if one exists)
- State 2: obtain Certificate

The one-step transition matrix, $P$, can be written in terms of parameters $\theta$ and $\phi$ as

\[
\begin{pmatrix}
1 - \theta & \phi \\
1 - \theta & 0 & \theta \\
0 & 0 & 1
\end{pmatrix}
\]

where $0 < \phi < \theta < 1$.

(i) Write down the interpretations of the parameters $\theta$ and $\phi$ in $P$. Find the missing entry for $p_{01}$ indicated by * and explain the values of $p_{11}$ and $p_{22}$.

(ii) For $i = 0$ or 1, let $\tau_i$ be the expected number of transitions needed to reach state 2 given that the chain is currently in state $i$. By conditioning on the outcome of the next observation, write down a pair of backward equations for $\tau_0$ and $\tau_1$.

(iii) Show that

\[
\tau_0 = \frac{1 + \theta - \phi}{\theta^2 + \phi(1-\theta)}.
\]

(iv) In each exercise, only 5% of programmers obtain 100% and 75% obtain under 90%.

Two programmers score respectively 83% and 96% on their first attempt. Calculate the difference, if any, between the expected number of further attempts each requires to gain the Certificate of Programming Competency.
Two players, Amy and Betsy, compete in a simple game which consists of a sequence of rounds whose results are independent of each other and which continues until one or other of the players loses all her capital, i.e. is ruined. At each play, £1 is transferred from the loser to the winner; Amy has probability $\gamma$ of winning, where $0 < \gamma < 1$, while Betsy has probability $1 - \gamma$ of winning. Suppose that Amy starts with capital £$a$ and Betsy starts with capital £$b$, and let the total capital in pounds of both players combined be $T = a + b$.

(i) Let $X_A$ denote Amy's current capital and $p_i$ denote the probability that Amy is eventually ruined, given that $X_A = i$, where $0 \leq i \leq T$.

(a) Show that the $p_i$ satisfy the difference equation

$$p_i = \gamma p_{i+1} + (1 - \gamma) p_{i-1} \quad \text{for} \quad i = 1, \ldots, T - 1.$$ 

State the values of $p_0$ and $p_T$, justifying your answers.

(b) Assuming that $\gamma \neq \frac{1}{2}$, verify that the difference equation in part (a) is satisfied by $p_i = A + B\phi^i$, where $\phi = \frac{1 - \gamma}{\gamma}$.

(c) Hence solve the difference equation of part (a), assuming that $\gamma \neq \frac{1}{2}$, and deduce that at the start of the game the probability that Amy is eventually ruined is given by

$$\frac{\phi^T - \phi^a}{\phi^T - 1}.$$ 

(ii) Suppose instead that Amy starts with capital £$a$ but Betsy has unlimited resources. Find the probability that Amy is eventually ruined, distinguishing between the two cases $0 < \gamma < 0.5$ and $0.5 < \gamma < 1$. 

(6)
3. (i) Write down the distribution of \( N(t) \), the number of events in time \( t \) in a Poisson process of rate \( \lambda \) per unit time. Deduce the cumulative distribution function of \( T_1 \), the time to the first event in the process. Identify the distribution of \( T_1 \).

(ii) For \( s < t \), write down the joint probability that one event occurs in the interval [0, \( s \)] and none in \((s, t]\). Hence show that \( P(T_1 < s \mid N(t) = 1) = \frac{s}{t} \), for \( s < t \), and deduce that \( T_1 \mid N(t) = 1 \) is uniformly distributed on \([0, t]\). Write down \( E[N(t)] \) and \( E[T_1 \mid N(t) = 1] \).

You are now given the result that, under the condition that \( n \) events occur in time \( t \), the times \( S_1, S_2, \ldots, S_n \) at which these \( n \) events occur are jointly distributed as independent uniform random variables on \([0, t]\).

In a factory manufacturing tablet computers, assembled items arrive at the packaging point as a Poisson process of rate \( \lambda \) per unit time. At a fixed time \( \tau \) after the previous dispatch time, all packaged items are dispatched from the factory.

(iii) Explain why the expected total waiting time of all the items dispatched at time \( \tau \) is \( \frac{1}{2} \lambda \tau^2 \).

(iv) Investment funds are available to introduce two additional dispatch times at \( u \) and \( v \) respectively, where \( 0 < u < v < \tau \).

(a) Show that the expected total wait of all items dispatched at times \( u, v \) and \( \tau \) will now be

\[
h(u, v) = \frac{\lambda}{2} \left[ u^2 + (v-u)^2 + (\tau-v)^2 \right].
\]

(b) Deduce the optimal choice of \( u \) and \( v \) to minimise the expected total wait and show that it does not depend on \( \lambda \).
4. (i) An equilibrium M/G/1 queue has arrival rate $\lambda$ and a random service time whose mean is $\frac{1}{\mu}$ and variance is $\sigma^2$. The expected number left in the system by a departing customer is denoted by $E(N_S)$. Explain why the expected time a customer spends in the system from arrival to departure is given by $\frac{1}{\lambda} E(N_S)$. 

(ii) You are given that

$$E(N_S) = \frac{\rho(2-\rho)+\lambda^2\sigma^2}{2(1-\rho)}, \text{ where } \rho = \frac{\lambda}{\mu} < 1.$$ 

Aircraft operating a round-the-clock airlift of supplies to a famine-stricken area arrive randomly at average intervals of 4 hours at an airfield which has facilities for unloading and refuelling them one at a time. Calculate an aircraft's expected turnround time from arrival to departure under each of the following models.

(a) Unloading and refuelling takes exactly 3 hours in total for each plane. 

(b) The time for unloading plus refuelling is the sum of two independent times each having mean and standard deviation equal to $1\frac{1}{2}$ hours.

(c) Unloading and refuelling operations happen simultaneously so that the combined time, $X$, is the maximum of two independent random variables having exponential distributions with mean $1\frac{1}{2}$ hours. In this case you should first show that the probability density function of $X$ is

$$f(x) = \frac{4}{\pi} e^{-\frac{\pi}{x}} - \frac{4}{\pi} e^{-\frac{4}{x}}, \quad x > 0.$$ 

[You may use without proof the result

$$\int_0^\infty x^r e^{-x} dx = r!$$

for $r = 0, 1, 2, \ldots$.]
5. (i) Define the lag $k$ autocorrelation $\rho_k$ of a time series.

(ii) The approximate formula

$$SE(r_k) \approx \sqrt{\frac{1+2\sum_{j=1}^{k-1} r_j^2}{n}}$$

is commonly used in the identification of time series models. What is $r_k$? Under what conditions on $\rho_k$ is the above formula valid?

Describe the use of $SE(r_k)$ in the identification of possible models for a time series of length $n$. (1)

A time series of 100 wind speed measurements (24 hour averages) has been plotted in the computer output shown on the next page. The table below gives the sample autocorrelation function (acf) and sample partial autocorrelation function (pacf) for the first 16 lags.

<table>
<thead>
<tr>
<th>Lag</th>
<th>acf</th>
<th>pacf</th>
<th>Lag</th>
<th>acf</th>
<th>pacf</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>0.654</td>
<td>0.654</td>
<td>9</td>
<td>-0.126</td>
<td>-0.043</td>
</tr>
<tr>
<td>2</td>
<td>0.308</td>
<td>-0.210</td>
<td>10</td>
<td>-0.005</td>
<td>0.193</td>
</tr>
<tr>
<td>3</td>
<td>0.105</td>
<td>-0.002</td>
<td>11</td>
<td>0.087</td>
<td>0.026</td>
</tr>
<tr>
<td>4</td>
<td>0.086</td>
<td>0.123</td>
<td>12</td>
<td>0.094</td>
<td>-0.114</td>
</tr>
<tr>
<td>5</td>
<td>-0.038</td>
<td>-0.254</td>
<td>13</td>
<td>0.030</td>
<td>-0.025</td>
</tr>
<tr>
<td>6</td>
<td>-0.179</td>
<td>-0.095</td>
<td>14</td>
<td>-0.043</td>
<td>-0.110</td>
</tr>
<tr>
<td>7</td>
<td>-0.242</td>
<td>-0.016</td>
<td>15</td>
<td>-0.059</td>
<td>-0.019</td>
</tr>
<tr>
<td>8</td>
<td>-0.177</td>
<td>0.013</td>
<td>16</td>
<td>-0.024</td>
<td>0.092</td>
</tr>
</tbody>
</table>

(iii) Calculate $SE(r_k)$ for lags 1, 2 and 3. What standard error should be used for the pacf? Discuss which models of the ARIMA class it might be appropriate to attempt to fit to these data. (4)

(iv) The output also includes some results obtained by fitting two ARIMA models, A and B, to the wind speed data. The software used fits an ARIMA($p$, $d$, $q$) model to $x_t$ by first calculating $y_t$ by taking $d$th differences of $x_t$, and then fitting the model

$$y_t = \sum_{i=1}^{p} \phi_i y_{t-i} + \theta_0 + A_t + \sum_{i=1}^{q} \theta_i A_{t-i},$$

where $A_t$ is white noise.

State which models have been fitted and write down explicitly their model equations. (5)

(v) Discuss what the output tells you about the suitability of each of the models, A and B. (4)

Computer output is on the next page.
ARIMA Model A: Wind

Final Estimates of Parameters

<table>
<thead>
<tr>
<th>Type</th>
<th>Coef</th>
<th>SE Coef</th>
<th>T</th>
<th>P</th>
</tr>
</thead>
<tbody>
<tr>
<td>MA   1</td>
<td>-0.8084</td>
<td>0.0897</td>
<td>-9.02</td>
<td>0.000</td>
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<tr>
<td>MA   2</td>
<td>-0.4759</td>
<td>0.0896</td>
<td>-5.31</td>
<td>0.000</td>
</tr>
<tr>
<td>Constant</td>
<td>11.1651</td>
<td>0.6649</td>
<td>16.79</td>
<td>0.000</td>
</tr>
<tr>
<td>Mean</td>
<td>11.1651</td>
<td>0.6649</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

Number of observations: 100
Residuals: SS = 827.243 (backforecasts excluded)
MS = 8.528  DF = 97

Modified Box-Pierce (Ljung-Box) Chi-Square statistic

<table>
<thead>
<tr>
<th>Lag</th>
<th>Chi-Square</th>
<th>DF</th>
<th>P-Value</th>
</tr>
</thead>
<tbody>
<tr>
<td>12</td>
<td>10.1</td>
<td>9</td>
<td>0.339</td>
</tr>
<tr>
<td>24</td>
<td>12.8</td>
<td>21</td>
<td>0.915</td>
</tr>
<tr>
<td>36</td>
<td>30.5</td>
<td>33</td>
<td>0.590</td>
</tr>
<tr>
<td>48</td>
<td>45.3</td>
<td>45</td>
<td>0.457</td>
</tr>
</tbody>
</table>

ARIMA Model B: Wind

Final Estimates of Parameters

<table>
<thead>
<tr>
<th>Type</th>
<th>Coef</th>
<th>SE Coef</th>
<th>T</th>
<th>P</th>
</tr>
</thead>
<tbody>
<tr>
<td>AR   1</td>
<td>-0.0525</td>
<td>0.0989</td>
<td>-0.53</td>
<td>0.597</td>
</tr>
<tr>
<td>AR   2</td>
<td>-0.2091</td>
<td>0.0982</td>
<td>-2.13</td>
<td>0.036</td>
</tr>
<tr>
<td>AR   3</td>
<td>-0.2844</td>
<td>0.1008</td>
<td>-2.82</td>
<td>0.006</td>
</tr>
<tr>
<td>Constant</td>
<td>0.0133</td>
<td>0.3208</td>
<td>0.04</td>
<td>0.967</td>
</tr>
</tbody>
</table>

Differencing: 1 regular difference
Number of observations: Original series 100, after differencing 99
Residuals: SS = 966.857 (backforecasts excluded)
MS = 10.177  DF = 95

Modified Box-Pierce (Ljung-Box) Chi-Square statistic

<table>
<thead>
<tr>
<th>Lag</th>
<th>Chi-Square</th>
<th>DF</th>
<th>P-Value</th>
</tr>
</thead>
<tbody>
<tr>
<td>12</td>
<td>18.4</td>
<td>8</td>
<td>0.019</td>
</tr>
<tr>
<td>24</td>
<td>20.7</td>
<td>20</td>
<td>0.416</td>
</tr>
<tr>
<td>36</td>
<td>45.4</td>
<td>32</td>
<td>0.058</td>
</tr>
<tr>
<td>48</td>
<td>62.7</td>
<td>44</td>
<td>0.033</td>
</tr>
</tbody>
</table>
6. For daily sales forecasting a forecaster regularly uses the ARIMA(0, 1, 1) model
\[ Y_t = Y_{t-1} + A_t - \beta A_{t-1}, \]
where \(0 < \beta < 1\) and \(A_t\) is white noise with variance \(\sigma^2\).

(i) Show that \(\hat{Y}_t(h)\), the minimum mean square error forecast of \(Y_{t+h}\) at \(t\), is constant for all lags \(h \geq 1\). Show also that \(\hat{Y}_t(1)\) is a function of \(Y_t\) and a forecast error \(a_t\). Explain how \(a_t\) would be calculated.

(ii) Show that \(Y_t\) can be written in MA(\(\infty\)) form as
\[ Y_t = \sum_{i=0}^{\infty} \psi_i A_{t-i} \quad \text{with} \quad \psi_0 = 1 \quad \text{and} \quad \psi_i = 1 - \beta \quad \text{for} \quad i \geq 1. \]
Deduce that the \(h\)-step-ahead forecast error variance is
\[ V(h) = \sigma^2 \{1 + (h - 1)(1 - \beta)^2\}, \quad \text{for} \quad h \geq 1. \]

(iii) On Monday, in forecasting Tuesday’s sales, the forecaster calculated the one-step-ahead (i.e. one-day-ahead) forecast to be 812 with a standard error of 4. At the same time her forecast of Wednesday’s sales had a standard error of 5.

(a) Find the value of \(\beta\) used and the variance of the underlying white noise.

(b) Calculate also the variance of her forecast of the total sales for Tuesday and Wednesday and give a 90% prediction interval for her forecast of this total.
7. The time series $Y_t$ is second order stationary with mean $\mu$, variance $\gamma_0$ and autocorrelation function (acf) $\rho_k$, for $k = 0, 1, 2, \ldots$.

(i) Define $\rho_k$ and explain what is meant by the phrase *second order stationary*.

(ii) Find expressions for the mean, variance and acf of $Z_t = Y_t + \omega Y_{t-1}$, where $\omega$ is a constant, in terms of $\mu$, $\gamma_0$ and the autocorrelations of $Y_t$. Deduce that $Z_t$ is also second order stationary.

(iii) Deduce also that the time series $V_t = Y_t - 2Y_{t-1} + Y_{t-2}$ is second order stationary.

(iv) If the acf of $Y_t$ is $\alpha^k$, for $k \geq 0$, find the values of $\omega$ such that $Y_t + \omega Y_{t-1}$ is white noise.
8. The spectral density function \( f(\omega) \) of a stationary time series model having autocovariance function \( \{\gamma_k\} \) (for \( k = 0, 1, 2, \ldots \)) may be written as

\[
f(\omega) = \frac{1}{2\pi} \sum_{k=-\infty}^{\infty} \gamma_k e^{-i\omega k}, \text{ for } -\pi < \omega < \pi.
\]

(i) Show that the normalised spectral density, \( g(\omega) = \frac{2\pi f(\omega)}{\gamma_0} \), may be written equivalently as

\[
g(\omega) = 1 + \sum_{k=1}^{\infty} \rho_k (e^{-i\omega k} + e^{i\omega k})
\]

where \( \rho_k \) is the autocorrelation function of the time series.

(ii) Deduce that \( g(\omega) = g(-\omega) \), for \(-\pi < \omega < \pi\).

(iii) For the time series model \( X_t = \alpha X_{t-1} + A_t \) for \(|\alpha| < 1\), where \( A_t \) is white noise, show that

\[
\rho_k = \alpha^k, \text{ for } k = 0, 1, 2, \ldots ,
\]

and that

\[
g(\omega) = \frac{1 - \alpha^2}{1 - 2\alpha \cos \omega + \alpha^2}.
\]

(iv) In the case \( \alpha = -0.7 \), sketch \( g(\omega) \) for \( 0 < \omega < \pi \) and compare it with the spectral density function of white noise. Comment on what the shape of \( g(\omega) \) tells you about how a time series plot of a realisation of \( X_t \) would compare to white noise.