Master 2 M0 2017-2018 Analyse des séries financières

## Final exam, February 2018

3h00, without any documents.

In all the sequel, $\left(\xi_{t}\right)_{t \in \mathbf{Z}}$ is a sequence of centered independent and identically distributed random variables with a symmetric distribution (the distributions of $\xi_{0}$ and $-\xi_{0}$ are the same) continuous with respect to Lebesgue measure, such as $\mathbb{E}\left(\xi_{0}^{2}\right)=1$.

## 1. Preliminaries:

(a) Let $\left(X_{t}\right)_{t \in \mathbf{Z}}$ be a stationary time series. Prove that $\left(X_{t}^{2}\right)_{t \in \mathbf{Z}}$ is a stationary time series.
(b) Assume that $\left(Y_{t}^{2}\right)_{t \in \mathbf{Z}}$ is a stationary time series. Prove that $\left(Y_{t}\right)_{t \in \mathbf{Z}}$ is not necessary a stationary time series.
(c) Let $\left(u_{t}\right)_{t \in \mathbf{Z}}$ be a stationary time series, independent to $\left(\xi_{t}\right)_{t \in \mathbf{Z}}$. Prove that $\left(\xi_{t} u_{t}\right)_{t \in \mathbf{Z}}$ is a stationary time series.
(d) For a random variable $Z$, define $\operatorname{sign}(Z)=\mathbb{I}_{Z>0}-\mathbb{I}_{Z<0}$. Prove that $\left(\operatorname{sign}\left(\xi_{t}\right)\right)_{t \in \mathbf{Z}}$ is a white noise, independent of $\left(\left|\xi_{t}\right|\right)_{t \in \mathbf{Z}}$. Let $\left(Y_{t}\right)_{t \in \mathbf{Z}}$ be a time series defined by $Y_{t}=\xi_{t} G\left(\left(\left|Y_{t-i}\right|\right)_{i \in \mathbf{N}^{*}}\right)$ for any $t \in \mathbf{Z}$ where $G: \mathbf{R}^{\mathbf{N}} \rightarrow(0, \infty)$ is a fixed function. Assume that $\left(Y_{t}^{2}\right)_{t \in \mathbf{Z}}$ is a causal (with respect to $\left.\left(\left(\xi_{s}\right)_{s \leq t}\right)_{t \in \mathbf{Z}}\right)$ stationary process. Prove that $\left(Y_{t}\right)_{t \in \mathbf{Z}}$ is also a causal stationary time series.
2. Main theoretical part: If it exists, we consider a sequence $\left(X_{t}\right)_{t \in \mathbf{Z}}$ such as:

$$
\begin{equation*}
X_{t}=\alpha X_{t-1}+\varepsilon_{t} \quad \text { with } \quad \varepsilon_{t}=\xi_{t} \sqrt{a_{0}+a_{1} X_{t-1}^{2}} \quad \text { for any } t \in \mathbf{Z} \tag{1}
\end{equation*}
$$

where $\left.\left(\alpha, a_{0}, a_{1}\right) \in\right]=\mathbf{R} \times(0, \infty) \times[0, \infty)$ are unknown parameters.
(a) In this question, we assume $\alpha=a_{1}=0$. Which kind of process is $\left(X_{t}\right)_{t \in \mathbf{Z}}$ and provide a condition of the existence of a stationary causal second order solution. Compute $\mathbb{E}\left(X_{0}\right)$ and $r_{X}(k)=\operatorname{cov}\left(X_{0}, X_{k}\right)$ for $k \in \mathbf{N}$.
(b) In this question, we assume $a_{1}=0$ and $\alpha \neq 0$. Which kind of process is $\left(X_{t}\right)_{t \in \mathbf{Z}}$ and provide a condition of the existence of a stationary causal second order solution. Compute $\mathbb{E}\left(X_{0}\right)$ and $r_{X}(k)$ for $k \in \mathbf{N}$.
(c) In this question, we assume $\alpha=0$ and $a_{1}>0$. Which kind of process is $\left(X_{t}\right)_{t \in \mathbf{Z}}$ and provide condition of the existence of a stationary causal second order solution. Compute $\mathbb{E}\left(X_{0}\right)$ and $r_{X}(k)$ for $k \in \mathbf{N}$.
(d) Now and until the end, we study the general case $\left.\left(\alpha, a_{0}, a_{1}\right) \in\right]=\mathbf{R} \times(0, \infty) \times[0, \infty)$. Prove that $\left(X_{t}\right)_{t \in \mathbf{Z}}$ is an affine causal process. Prove that the function $x \rightarrow \sqrt{1+x^{2}}$ is Lipshitzian and deduce that a sufficient condition for $\left(X_{t}\right)_{t \in \mathbf{Z}}$ to be a causal stationary second order process is:

$$
\begin{equation*}
|\alpha|+\sqrt{a_{1}}<1 \tag{2}
\end{equation*}
$$

(e) For $|\alpha|<1$, prove that if $\left(X_{t}\right)_{t \in \mathbf{Z}}$ is a causal stationary second order process then $\left(\varepsilon_{t}\right)_{t \in \mathbf{Z}}$ defined by

$$
\begin{equation*}
\varepsilon_{t}=\xi_{t} \sqrt{a_{0}+a_{1}\left(\sum_{i=0}^{\infty} \alpha^{i} \varepsilon_{t-1-i}\right)^{2}} \quad \text { for any } t \in \mathbf{Z} \tag{3}
\end{equation*}
$$

is a causal stationary second order process and a weak white noise. Show that if $\left(\varepsilon_{t}\right)_{t \in \mathbf{Z}}$ is a causal stationary second order process then

$$
\begin{equation*}
a_{1}+\alpha^{2}<1 \tag{4}
\end{equation*}
$$

Compare this condition with (2).
(f) In Doukhan et al. (2016), it was established that under (4), then $\left(\varepsilon_{t}\right)_{t \in \mathbf{Z}}$ is a causal stationary second order process. Extend this property to $\left(X_{t}\right)_{t \in \mathbf{Z}}$. Under (4), compute $\mathbb{E}\left(X_{0}\right)$ and $r_{X}(k)$ for $k \in \mathbf{N}$.
(g) Deduce also $\mathbb{E}\left(X_{t} \mid\left(X_{t-s}\right)_{s \in \mathbf{N}^{*}}\right)$ and $\operatorname{var}\left(X_{t} \mid\left(X_{t-s}\right)_{s \in \mathbf{N}^{*}}\right)$. Is $\left(X_{t}\right)_{t \in \mathbf{Z}}$ a conditionaly heteroskedastic process?
(h) Assume now that $\left(X_{1}, \cdots, X_{N}\right)$ is observed and let $\theta={ }^{t}\left(\alpha, a_{0}, a_{1}\right)$. Provide the expression of the quasi-maximum likelihood estimator $\hat{\theta}$ of $\theta$. Is $\widehat{\theta}$ a consistent estimator? What is its convergence rate?
(i) Provide forecasting of $X_{N+1}$ and $X_{N+1}^{2}$.
3. Numerical part: We study with R software the open daily historical data of Bitcoin from January 282014 to January 282018.
(a) First the following commands have been executed with figures exhibited below:

```
Bit=read.csv("C:/Users/Admin/Dropbox/Enseignement/M2 MO/TP/BTC-USD.csv")
Bit0=Bit$Open; n=length(Bit0)
plot.ts(Bit0); plot.ts(log(Bit0))
Y=log(Bit0); X1=c(1:n); X2=X1^2
Y.lm=lm(Y~}\textrm{X}1+\textrm{X}2); summary(Y.lm
```

Command lm realizes a least squares linear regression. Here there are the graphs and main numerical results:


Coefficients:
Estimate Std. Error t value $\operatorname{Pr}(>|t|)$
$\begin{array}{lrrrr}\text { (Intercept) } & 6.541 \mathrm{e}+00 & 1.645 \mathrm{e}-02 & 397.61 & <2 \mathrm{e}-16 * * * \\ \text { X1 } & -4.034 \mathrm{e}-03 & 5.396 \mathrm{e}-05 & -74.76 & <2 \mathrm{e}-16 * * * \\ \text { X2 } & 4.332 \mathrm{e}-06 & 3.711 \mathrm{e}-08 & 116.73 & <2 \mathrm{e}-16 * * *\end{array}$

Residual standard error: 0.2054 on 1404 degrees of freedom
Multiple R-squared: 0.9635,Adjusted R-squared: 0.9635
F-statistic: 1.854e+04 on 2 and 1404 DF, p-value: < 2.2e-16
Question II.1: Explain what is done.
(b) New commands are then executed:

```
plot.ts(Y.lm$residuals)
Fit=arima(Y.lm$residuals, order = c(1,0,2))
acf(Fit$residuals)
Box.test(Fit$residuals, lag = 20,"Ljung-Box", fitdf=3)
```

Here there are graphs and numerical results:



Box-Ljung test
data: Fit\$residuals
X-squared $=25.557, \mathrm{df}=17, \mathrm{p}$-value $=0.08292$

Question II.2: Explain what is done (notably why we use fitdf=3) and explain which conclusions you deduce.
(c) The following sequence of commands is then executed:

```
pred=predict(Fit,n.ahead=1); pred[1]
exp(pred$pred[1]+sum(Y.lm$coeff*c(1,(n+1),(n+1)^2)))
```

The results are the following:

```
>-0.09092553
```

> 11582.19

Question II.3: What is done here and what are your conclusions?
(d) Finally, the following sequence of commands is executed:

```
LogRetBit=log(Bit0[2:n]/Bit0[1:(n-1)])
plot.ts(LogRetBit); acf(LogRetBit)
library(fGarch)
FitLogRet1=garchFit(~garch(1,1),data=LogRetBit,trace=FALSE)
summary(FitLogRet1)
FitLogRet2=garchFit(~ garch (1,2), data=LogRetBit,trace=FALSE)
summary(FitLogRet2)
```

The graphs and results are the following:

> Error Analysis:

|  | Estimate | Std. Error | t value $\operatorname{Pr}(>\|t\|)$ |  |  |
| :--- | ---: | ---: | ---: | ---: | :--- |
| mu | $1.646 \mathrm{e}-03$ | $7.012 \mathrm{e}-04$ | 2.347 | 0.018911 | $*$ |
| omega | $2.894 \mathrm{e}-05$ | $7.565 \mathrm{e}-06$ | 3.825 | 0.000131 | $* * *$ |
| alpha1 | $1.690 \mathrm{e}-01$ | $2.155 \mathrm{e}-02$ | 7.845 | $4.44 \mathrm{e}-15$ | $* * *$ |
| beta1 | $8.365 \mathrm{e}-01$ | $1.774 \mathrm{e}-02$ | 47.139 | $<2 \mathrm{e}-16$ | $* * *$ |

Standardised Residuals Tests:

Statistic p-Value

| Jarque-Bera Test | R | Chi^2 | 2253.533 | 0 |
| :--- | :--- | :--- | :--- | :--- |
| Shapiro-Wilk Test | $R$ | W | 0.9142998 | 0 |
| Ljung-Box Test | $R$ | $Q(10)$ | 31.40408 | 0.0005030119 |
| Ljung-Box Test | $R$ | $Q(15)$ | 34.67742 | 0.002732705 |
| Ljung-Box Test | $R$ | $Q(20)$ | 42.10945 | 0.002676158 |
| Ljung-Box Test | $R \wedge 2$ | $Q(10)$ | 7.152754 | 0.7109496 |
| Ljung-Box Test | $R \wedge 2$ | $Q(15)$ | 11.73218 | 0.6991771 |
| Ljung-Box Test | $R \wedge 2$ | $Q(20)$ | 16.21253 | 0.7033551 |
| LM Arch Test | $R$ | $T R \wedge 2$ | 9.629902 | 0.648393 |

Information Criterion Statistics:
AIC BIC SIC HQIC
$-3.971264-3.956332-3.971280-3.965683$
> Error Analysis:

| Estimate S | Std. Error t | value $\operatorname{Pr}(>\|t\|)$ |
| :---: | :---: | :---: |
| mu $\quad 1.711 \mathrm{e}-03$ | $7.038 \mathrm{e}-04$ | 2.4310 .015041 |
| omega 3.529e-05 | $1.068 \mathrm{e}-05$ | 3.3050 .000949 *** |
| alpha1 2.098e-01 | $3.277 \mathrm{e}-02$ | $6.4011 .54 \mathrm{e}-10$ *** |
| beta1 $4.464 \mathrm{e}-01$ | $1.945 \mathrm{e}-01$ | 2.2950 .021745 |
| beta2 $3.485 \mathrm{e}-01$ | $1.755 \mathrm{e}-01$ | 1.9850 .047127 |
| Standardised Residuals Tests: |  |  |
|  |  | Statistic p-Value |
| Jarque-Bera Test | R $\mathrm{Chi}^{\wedge} 2$ | 2360.2690 |
| Shapiro-Wilk Test | $\mathrm{R} \quad \mathrm{W}$ | 0.9149907 |
| Ljung-Box Test | $R \quad \mathrm{Q}(10)$ | 30.514120 .000705495 |
| Ljung-Box Test | $R \quad \mathrm{Q}(15)$ | 33.536330 .003954055 |
| Ljung-Box Test | $R \quad \mathrm{Q}(20)$ | 40.947990 .003782937 |
| Ljung-Box Test | R^2 Q (10) | 5.4516140 .8590436 |
| Ljung-Box Test | R^2 Q (15) | 8.6995280 .8926969 |
| Ljung-Box Test | R^2 Q (20) | 13.333150 .8626366 |
| LM Arch Test | R TR^2 | 7.5097030 .8221771 |
| Information Criterion Statistics: |  |  |
| AIC BIC | S SIC | HQIC |
| -3.974214-3.955549 | -3.974239 - | 3.967238 |

Question II.4: What is done here and which model could you chose?

