Time Series Homework 2

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Problem 1

For real valued random variables X, and Y, denote mixing coefficient $\alpha(X,Y)$ as

$$\alpha = \sup_{(x,y)\in R^2} |P(X > x, Y > y) - P(X > x)P(Y > y)|.$$

Now we see that

$$\begin{split} E|XY| &= E|\int_0^\infty (I_{X>x} - I_{X<-x}) dx \int_0^\infty (I_{Y>y} - I_{Y<-y}) dy| \\ &\leq E\int_0^\infty I_{|X|>x} dx \int_0^\infty I_{|Y|>y} dy \\ &= \int_0^\infty \int_0^\infty E[I_{|X|>x} I_{|Y|>y}] dx dy. \end{split} \tag{Fubini's theorem)}$$

We want to show

$$E[I_{|X|>x}I_{|Y|>y}] \le \min\{1, P(|X|>x), P(|Y|>y)\}. \tag{1}$$

This is true by Holder's inequality:

$$\int I_{|X|>x}I_{|Y|>y} \le \int |I_{|X|>x}| \int \max |I_{|Y|>y}| = P(|X|>x).$$

and we can change the order of X, and Y to bound by P(|Y| > y). We can show

$$\begin{aligned} & \min\{1, P(|X| > x), P(|Y| > y)\} \\ &= \int_0^1 I_{u < P(|X| > x)I_{u < P(|Y| > y)}du} \\ &= \int_0^1 I_{x < Q_x(u)I_{y < Q_y(u)}du}. \end{aligned}$$

Combining above all together, we have that

$$\begin{split} &E|XY|\\ &\leq \int_0^\infty \int_0^\infty \int_0^1 I_{x < Q_x(u)I_{y < Q_y(u)}} du dx dy\\ &= \int_0^1 Q_X(u)Q_Y(u) du \end{split} \tag{Fubini's theorem.}$$

Problem 2

The sequence of strong mixing coefficients $\{\alpha_n, n \geq 0\}$ of $\{X_i, i \in N\}$ is defined by

$$\alpha_0 = 1/2$$
 and $\alpha_n = \sup_{|i-j| > n} \alpha(\sigma(X_i), \sigma(X_j))$ for $n > 0$.

By definition, it is non-increasing and right continuous. For $x \ge 0$, define $\alpha(x) = \alpha_{|x|}$.

$$\alpha^{-1}(u) = \inf\{k \in N : \alpha_k \le u\} = \sum_{i=0}^{\infty} I_{u < \alpha_i}..$$
 (count)

Now by Holder's inequality

$$M_{2,\alpha}(Q_k) \le \left(\int_0^1 [\alpha^{-1}(u)]^{r/(r-2)} du\right)^{1-2/r} \left(\int_0^1 Q_k^{2*\frac{r}{2}}(u) du\right)^{2/r}.$$

The second integral on r.h.s. is equal to $||X_k||_r^2$ because $Q_k(U)$ has the same distribution as X_k . Since α is non-increasing right continuous,

$$x \ge \alpha^{-1}(u)$$
 iff $\alpha(x) \ge u$.

This implies that

$$\begin{split} & \int_0^1 [\alpha^{-1}(u)]^{r/(r-2)} du \\ & = \int_0^1 \int_0^\infty I_{x \le [\alpha^{-1}(u)]^{r/(r-2)}} dx du \\ & = \int_0^\infty \int_0^1 I_{u \le \alpha(x)^{1-2/r}} du dx \\ & = \int_0^\infty \alpha(x)^{1-2/r} dx. \end{split}$$

When $1 \le x^{1-2/r} < i+1$, $i^{r/(r-2)} \le x < (i+1)^{r/(r-2)}$ and $\alpha(x^{1-2/r}) = \alpha_i$. Then

$$\int_{0}^{\infty} \alpha(x)^{1-2/r} dx$$

$$= \sum_{i=0}^{\infty} ((i+1)^{r/(r-2)} - i^{r/(r-2)}) \alpha_{i}$$

$$\leq \frac{r}{r-2} \sum_{i\geq 0} (i+1)^{2/(r-2)} \alpha_{i}.$$
 (by MVT)

Combining above,

$$M_{2,\alpha}(Q_k) \leq (\frac{r}{r-2})^{1-2/r} \bigg(\sum (i+1)^{2/(r-2)} \alpha_i \bigg)^{1-2/r} ||X_k||_r^2.$$

In particular, DMR conditions holds if

$$\sum_{i>0} (i+1)^{2/(r-2)} \alpha_i < \infty.$$

If $\alpha_i = O[i^{-r/(r-2)}(logi)^{-\theta}]$, then

$$(i+1)^{2/(r-2)}\alpha_i = O[(1/i)^{r/(r-2)}\frac{1}{(logi)^{\theta}}] \le O[\frac{1}{i(logi)^{\theta}}]$$

converges when $\theta > 1$ because

$$\int \frac{1}{x(\log x)^{\theta}} dx = \int 1/u^{\theta} du < \infty.$$

when $\theta > 1$

1 Problem 3.

Norm convergence of a sequence $\{X_n\}$ of elements of L^2 to the limit X means

$$||X_n - X||^2 = E|X_n - X|^2 \to 0$$

as $n \to \infty$. This is called mean-square convergence.

The space of complex-valued random variable X with finite second moment is a complex Hilbert space. To show mean-square convergence of $S_m := \sum_{j=1}^m \theta^j X_{n+1-j}$ it is sufficient to show the sum is Cauchy.

$$||S_l - S_m||^2 = E \sum_{m+1 \le i \le l} \sum_{m+1 \le j \le l} E\theta^8 X_{n-(i-1)\theta^j X_{n-(j-1)}}$$
$$= \sum_{m+1 \le i \le l} \sum_{m+1 \le j \le l} \theta^{i+j} (\mu^2 + \gamma(j-i)).$$

Since $|\theta| < 1$, we can find a large integer N such that if $m, l \geq N$, then

$$\theta^{i+j}(\mu^2 + \gamma(j-i)) < \epsilon/(l-m)^2$$

for any pre-specified ϵ so that $\{S_i, i \geq 1\}$ is a Cauchy sequence.

2 Problem 4.

2.1 a

Note that in a separable Hilbert space, $\{x_1, ...\}$ are complete orthonormal bases.

$$P_{\bar{sp}\{x_1,\dots,x_n\}}(x) = \sum_{i=1}^n \langle x, x_i \rangle x_i.$$

as n goes to infinity, it converges to x because

$$P_{\bar{s}p\{x_1,\dots,x_\infty\}}(x) = \sum_{i=1}^{\infty} \langle x, x_i \rangle x_i = x$$

as shown in the lecture.

2.2 b

Random variables in a stationary process has finite second moment, so it is a L2 space which is a Hilbert space.

Let S_r be a linear space of $X_{n-1},...,X_{n-r}$, and \widehat{S}_r be a projection of X_n onto S_r such that

$$E(X_n - \widehat{S}_r)S_r = 0.$$

If we put $Y_r := X_n - \widehat{S}_r$, then

$$E(X - \widehat{S}_r)S_r = cov(Y_r, S_r).$$

By continuity of inner product in Hilbert space,

$$cov(Y_r, S_r) \to cov(Y, S),$$

where $Y := X_n - S$, S is a linear space of $X_{n-j}, 1 \le j \le \infty$.