

8/25/2010. Thursday Time Series. ①

White noise: The time series generated from uncorrelated variables is used as a model for noise in engineering application, where it is called white noise.

$$W_t \sim WN(0, \sigma^2)$$

The designation white originates from the analogy with white light & indicates that all possible periodic oscillations are present with equal strength.

White independent noise: require the noise to be iid random variables. with mean 0 & variance σ^2 . $W_t \sim iid(0, \sigma^2)$

Gaussian white noise: We are independent variables, with mean 0 & variance σ^2 . $W_t \sim iid N(0, \sigma^2)$

Example 1.9. Moving average.

replace W_t by an average of its current value and its immediate neighbors in the past and future.

$$Let \quad U_t = \frac{1}{3} (W_{t-1} + W_t + W_{t+1}).$$

expect a smoother version of the time series. the slower oscillations are more apparent & some of the faster oscillations are taken over.

norm (random variable)

filter: linear filtering on a T.S.
filter(x, filter.method = c("convolution", "recursive"), sides = 2, circular = FALSE, init)

par(mfrow = c(2, 2)) # 2x2 pictures on one plot.

mfrow = c(2, 1) # 2x1 pictures on one plot.

plot.ts

plot time series

Sides = 1. the fiber coefficients are for ③
base values only.

Sides = 2. they are centered around lag 0.
Length of the fibers should be odd, but
if it is even, more of the fibers is forward
in time than backward.

Example 1.1.1. Random walk.

$$X_t = \delta + X_{t-1} + W_t \quad * \quad \begin{array}{c} | \\ | \\ | \end{array} \text{ noise.}$$

$t = 1, 2, \dots$ initial condition $X_0 = 0$. W_t which

δ : drift. $\delta = 0$ $X_t = X_{t-1} + W_t$ is

called a random walk.

The term random walk comes from the
fact that, when $\delta = 0$, the value of the
time series at time t is the value
of the series at time $t-1$ plus a
complete random movement determined by
 W_t .

using induction.

$$X_t = \delta + X_{t-1} + W_t.$$

④

$$X_0 = 0.$$

$$X_1 = \delta + 0 + W_1$$

$$X_2 = \delta + X_1 + W_2 = \delta + \delta + W_1 + W_2 \\ = 2\delta + W_1 + W_2.$$

$$X_3 = \delta + X_2 + W_3$$

$$= \delta + 2\delta + W_1 + W_2 + W_3$$

$$= 3\delta + \sum_{i=1}^3 W_i$$

⋮

$$\therefore X_t = \delta t + \sum_{j=1}^t W_j$$

$t = 1, 2, \dots$

R code:

```
> set.seed(154)
```

```
> W <- rnorm(200, 0, 1); X = cumsum(W)
```

```
> Wd = W + .2; Xd = cumsum(Wd)
```

```
> plot.ts(Xd, ylim = c(-5, 55))
```

```
> lines(X)
```

```
> lines(.2 * (1:200), lty = "dashed")
```

Example 1.17. Auto covariance of a (5)
 moving average. $W_t \sim (0, \sigma^2 = 1)$

$$V_t = \frac{1}{3} (W_{t-1} + W_t + W_{t+1})$$

$$E(V_t) = 0$$

$$\begin{aligned} V_V(s, t) &= E[(V_s - 0)(V_t - 0)] \\ &= \frac{1}{9} E[(W_{s-1} + W_s + W_{s+1})(W_{t-1} + W_t + W_{t+1})] \end{aligned}$$

Let $s - t = h$.

$h = 0, \pm 1, \pm 2, \dots$

$h = 0$.

$$\begin{aligned} V_V(t, t) &= \frac{1}{9} E[(W_{t-1} + W_t + W_{t+1})(W_{t-1} + W_t + W_{t+1})] \\ &= \frac{1}{9} [E(W_{t-1}W_{t-1}) + E(W_tW_t) + E(W_{t+1}W_{t+1})] \\ &= \frac{3}{9} \end{aligned}$$

$E(X) = 0$

$V(X) = 1$

$$V(X) = E(X^2) - [E(X)]^2$$

$\therefore E(X^2) = 1$

$$h=1.$$

(6)

$$\begin{aligned} \gamma_V(t+1, t) &= \frac{1}{9} E [(W_t + W_{t+1} + W_{t+2}) (W_{t-1} + W_t + W_{t+1})] \\ &= \frac{1}{9} E [(W_t W_t) + E(W_{t+1} W_{t+1})] \\ &= \frac{2}{9}. \end{aligned}$$

Similar computations give $\gamma_V(t-1, t) = 2/9$

$$\gamma_V(t+2, t) = \gamma_V(t-2, t) = 1/9.$$

$$\gamma_V(s, t) = \begin{cases} \frac{3}{9} & |s-t| = 0 \\ \frac{2}{9} & |s-t| = 1 \\ \frac{1}{9} & |s-t| = 2 \\ 0 & |s-t| \geq 3 \end{cases}$$

① Shows that the smoothing operation introduces a covariance function that decreases as the separation between the two time points increases and disappears completely when the time points are separated by three or more time points.

② depends on the time separation or lag and not on the absolute location of the points along the axis.

①

Example 1.18. Autocovariance of a Random walk.

$$X_t = \sum_{j=1}^t w_j. \quad \text{we have.}$$

$$V_x(s, t) = \text{Cov}(X_s, X_t) = \text{Cov}\left(\sum_{j=1}^s w_j, \sum_{k=1}^t w_k\right)$$
$$= \min\{s, t\} \sigma_w^2.$$

We uncorrelated ~~var.~~ i.i.d.

① Autocovariance function of a random walk depends on the time values s, t .
not on the time separation or lag.

② $\text{Var}(X_t) = \gamma_x(t, t) = t \sigma_w^2$ ↑
We have bound as time t ↑.