

## AR, MA and ARMA

$$1. \text{AR}(1) \Rightarrow \text{MA}(\infty)$$

$$\text{AR}(1): r_t = \phi_0 + \phi_1 \underline{r_{t-1}} + a_t$$

Mean-adjusted AR(1):

$$\bullet x_t = \phi_1 x_{t-1} + a_t$$

$$\text{where } x_t = r_t - \mu, \mu = \bar{E}(r_t) = \frac{\phi_0}{1-\phi_1}$$

$$x_t = \phi_1 \underline{x_{t-1}} + a_t$$

$$= \phi_1 (\phi_1 x_{t-2} + a_{t-1}) + a_t$$

$$= \phi_1^2 x_{t-2} + \phi_1 a_{t-1} + a_t$$

$$= \phi_1^2 (\phi_1 x_{t-3} + a_{t-2}) + \phi_1 a_{t-1} + a_t$$

$$= \phi_1^3 x_{t-3} + \phi_1^2 a_{t-2} + \phi_1 a_{t-1} + a_t$$

$\vdots$

$$= \underline{\phi_1^k x_{t-k}} + \phi_1^{k-1} a_{t-(k-1)} + \phi_1^{k-2} a_{t-(k-2)}$$

$$+ \dots + \phi_1 a_{t-1} + a_t$$

$$\text{when } k \rightarrow \infty, \quad \phi_1^k x_{t-k} \rightarrow 0$$

$$(|\phi_1| < 1 \Rightarrow \phi_1^k \xrightarrow{k \rightarrow \infty} 0)$$

$$x_t = \sum_{k=0}^{\infty} \phi_1^k a_{t-k}$$

$$= a_t + \phi_1 a_{t-1} + \phi_1^2 a_{t-2} + \phi_1^3 a_{t-3} + \dots$$

$$x_t = r_t - \mu$$

$$\Rightarrow \boxed{r_t = \mu + \sum_{k=0}^{\infty} \phi_1^k a_{t-k}}$$

$$AR(1) \Rightarrow MA(\infty)$$

General Form of Linear TS Model:

$$r_t = \mu + \sum_{k=0}^{\infty} \psi_k a_{t-k}, \quad \psi_0 = 1$$

(MA( $\infty$ ))

$$AR(1) : \psi_k = \phi_1^k$$

$$\bullet \quad \begin{matrix} AR(p) \\ ARMA(p, q) \end{matrix} \Rightarrow MA(\infty)$$

$$2. \quad MA(1) \Rightarrow AR(\infty)$$

Check Week 7 InClass Notes

Invertibility of MA(1) :  $|\theta_1| < 1$

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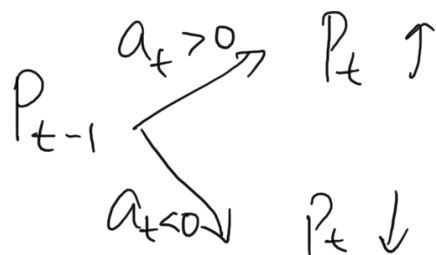
$$\begin{aligned} & \cdot \text{MAC}(q) \Rightarrow \text{AR}(\infty) \\ & \text{ARMA}(p, q) \end{aligned}$$

## Unit - Root Nonstationary Models

### 1. Random Walk

$$P_t = P_{t-1} + a_t, \quad \text{Var}(a_t) = \sigma_a^2, \quad \underline{P_0 \text{ given}}$$

( $P_t$  is log stock price,  $r_t = P_t - P_{t-1} = a_t$ )



① AR(1) model with  $\phi_1 = 1$

$$P_t = \phi_1 P_{t-1} + a_t$$

Unit Root:

$$(1 - B)P_t = a_t \quad B: \text{backshift operator (lag)}$$

$$\underline{1 - x = 0} \Rightarrow x = 1$$

② Mean:

$$\begin{aligned}
P_t &= \underline{P_{t-1}} + a_t \\
&= P_{t-2} + a_{t-1} + a_t \\
&= P_{t-3} + a_{t-2} + a_{t-1} + a_t \\
&\vdots \\
&= P_{t-t} + a_{t-t+1} + a_{t-t+2} + \dots + a_{t-1} + a_t \\
&= P_0 + a_1 + a_2 + \dots + a_{t-1} + a_t
\end{aligned}$$

$$\begin{aligned}
E(P_t) &= E(P_0 + a_1 + a_2 + \dots + a_{t-1} + a_t) \\
&= E(P_0) + E(a_1 + a_2 + \dots + a_{t-1} + a_t) \\
&= P_0
\end{aligned}$$

③ Variance:

$$\begin{aligned}
\text{Var}(P_t) &= \text{Var}(P_{t-1} + a_t) \\
&= \text{Var}(P_{t-1}) + \text{Var}(a_t) + 2 \underbrace{\text{Cov}(P_{t-1}, a_t)}_{=0} \\
&= \text{Var}(P_{t-1}) + \sigma_a^2
\end{aligned}$$

$\text{Var}(P_t) \neq \text{Var}(P_{t-1}) \Rightarrow P_t$  is Non stationary

$$\begin{aligned}
\text{Var}(P_t) &= \text{Var}(P_0 + a_1 + a_2 + \dots + a_{t-1} + a_t) \\
&= \text{Var}(a_1) + \text{Var}(a_2) + \dots + \text{Var}(a_{t-1}) + \text{Var}(a_t) \\
&= t \cdot \sigma_a^2 \quad \text{depends on time } t
\end{aligned}$$

when  $t \rightarrow \infty$ ,  $\text{Var}(P_t) \rightarrow \infty$

(4) Autocovariance:

$$\begin{aligned} \text{cov}(P_t, P_s) & \quad \underline{s < t} \\ &= \text{cov}(P_{t-1} + a_t, P_s) \\ &= \text{cov}(P_{t-2} + a_{t-1} + a_t, P_s) \\ &\vdots \\ &= \text{cov}(P_s + \underbrace{a_{s+1} + a_{s+2} + \dots + a_{t-1} + a_t}_{\sim}, P_s) \\ &= \text{cov}(P_s, P_s) \\ &= \text{Var}(P_s) = s \cdot \sigma_a^2 \\ \bullet \text{ cov}(P_t, P_s) &= \min\{t, s\} \cdot \sigma_a^2 \end{aligned}$$

(5) Autocorrelation:

$$\begin{aligned} \rho_{t,s} &= \frac{\text{cov}(P_t, P_s)}{\sqrt{\text{Var}(P_t) \cdot \text{Var}(P_s)}} \\ &= \frac{s \cancel{\sigma_a^2}}{\sqrt{t \cancel{\sigma_a^2} \cdot s \cancel{\sigma_a^2}}} = \sqrt{\frac{s}{t}} < 1 \end{aligned}$$

•  $\rho_{t,s}$  is close to 1

if  $S$  is not far away from  $t$

- Strong memory

(6) Forecast: at time origin  $n$

$$\begin{aligned} \bullet \hat{P}_n(1) &= \bar{E}(P_{n+1} | \bar{F}_n) & P_t &= P_{t-1} + a_t \\ & & \downarrow & \quad \downarrow \quad \downarrow \\ & & n+1 & \quad n+1 \quad n+1 \\ &= \bar{E}(P_n + a_{n+1} | \bar{F}_n) \\ &= \bar{E}(P_n | \bar{F}_n) + \bar{E}(a_{n+1} | \bar{F}_n) \\ &= P_n & & = 0 \end{aligned}$$

$$\begin{aligned} \hat{e}_n(1) &= P_{n+1} - \hat{P}_n(1) \\ &= P_n + a_{n+1} - P_n \\ &= a_{n+1} \end{aligned}$$

$$\text{Var}(\hat{e}_n(1)) = \text{Var}(a_{n+1}) = \sigma_a^2$$

$$\begin{aligned} \bullet \hat{P}_n(2) &= \bar{E}(P_{n+2} | \bar{F}_n) \\ &= \bar{E}(P_{n+1} + a_{n+2} | \bar{F}_n) \\ &= \bar{E}(P_n + a_{n+1} + a_{n+2} | \bar{F}_n) \\ &= \bar{E}(P_n | \bar{F}_n) + \bar{E}(a_{n+1} + a_{n+2} | \bar{F}_n) \\ &= P_n \end{aligned}$$

$$\begin{aligned}
\hat{e}_n(2) &= P_{n+2} - \hat{P}_n(2) \\
&= P_n + a_{n+1} + a_{n+2} - P_n \\
&= a_{n+1} + a_{n+2}
\end{aligned}$$

$$\begin{aligned}
\text{Var}(\hat{e}_n(2)) &= \text{Var}(a_{n+1} + a_{n+2}) \\
&= \text{Var}(a_{n+1}) + \text{Var}(a_{n+2}) + 2\text{Cov}(a_{n+1}, a_{n+2}) \\
&= 2\sigma_a^2 \qquad \qquad \qquad = 0
\end{aligned}$$

$$\begin{aligned}
\bullet \hat{P}_n(L) &= E(P_{n+L} | \bar{F}_n) \\
&= E(P_n + a_{n+1} + a_{n+2} + \dots + a_{n+L} | \bar{F}_n) \\
&= E(P_n | \bar{F}_n) + E(a_{n+1} + a_{n+2} + \dots + a_{n+L} | \bar{F}_n) \\
&= P_n
\end{aligned}$$

$$\begin{aligned}
\hat{e}_n(L) &= P_{n+L} - \hat{P}_n(L) \\
&= a_{n+1} + a_{n+2} + \dots + a_{n+L}
\end{aligned}$$

$$\begin{aligned}
\text{Var}(\hat{e}_n(L)) &= \text{Var}(a_{n+1} + a_{n+2} + \dots + a_{n+L}) \\
&= L \cdot \sigma_a^2
\end{aligned}$$

$$\text{When } L \rightarrow \infty, \quad \text{Var}(\hat{e}_n(L)) \rightarrow \infty$$

## 2. Random Walk with Drift

$$P_t = \underline{\mu} + P_{t-1} + a_t, \mu \neq 0, P_0 \text{ given}$$
$$\text{Var}(a_t) = \sigma_a^2$$

① has an unit root

② Nonstationary

③ strong memory

④ has a time trend with slope  $\mu$

$$\begin{aligned} P_t &= \mu + P_{t-1} + a_t \\ &= \mu + (\mu + P_{t-2} + a_{t-1}) + a_t \\ &= 2\mu + P_{t-2} + a_{t-1} + a_t \\ &= 2\mu + (\mu + P_{t-3} + a_{t-2}) + a_{t-1} + a_t \\ &= 3\mu + P_{t-3} + a_{t-2} + a_{t-1} + a_t \\ &\vdots \\ &= \underline{t \cdot \mu} + P_0 + a_1 + a_2 + \dots + a_{t-1} + a_t \end{aligned}$$

## 3. Differencing

1st difference:  $r_t = P_t - P_{t-1} = \Delta P_t$



- $P_t = P_{t-1} + a_t \Rightarrow r_t = a_t$

$$P_t : \text{ARIMA}(0, 1, 0)$$

- $P_t = \mu + P_{t-1} + a_t \Rightarrow r_t = \mu + a_t$

$$P_t : \text{ARIMA}(0, 1, 0)$$

- $P_t = \mu + P_{t-1} + \phi_1 \Delta P_{t-1} + a_t + \theta_2 a_{t-2}$

$$P_t : \text{ARIMA}(1, 1, 2)$$

$$\Rightarrow r_t = \mu + \phi_1 r_{t-1} + a_t + \theta_2 a_{t-2}$$

$$r_t = \Delta P_t = P_t - P_{t-1}$$

- $P_t = \mu + \phi_1 P_{t-1} + \phi_2 \Delta P_{t-1} + a_t + \theta_2 a_{t-2}$

$$P_t : \text{ARIMA}(2, 0, 2)$$

$$\Rightarrow P_t = \mu + \phi_1 P_{t-1} + \phi_2 (P_{t-1} - P_{t-2}) + a_t + \theta_2 a_{t-2}$$

$$= \mu + (\phi_1 + \phi_2) P_{t-1} - \phi_2 P_{t-2} + a_t + \theta_2 a_{t-2}$$

when  $\phi_1 = 1$ ,  $P_t : \text{ARIMA}(1, 1, 2)$

$\phi_1 \neq 1$ ,  $P_t : \text{ARIMA}(2, 0, 2)$

#### 4. Unit-Root test

- Test for  $AR(1)$  for  $P_t$ :

$$P_t = \phi_1 P_{t-1} + a_t$$

$$\text{or } P_t = \mu + \phi_1 P_{t-1} + a_t$$

$$H_0: \phi_1 = 1 \quad \text{vs} \quad H_1: \phi_1 < 1$$

$H_0$ : there is a unit root for  $P_t$

Dickey-Fuller test

$P \text{ value} > 5\% \Rightarrow P_t \text{ has a unit root}$

- Test for  $AR(p)$  for  $P_t$ :

$$P_t = C_t + \beta P_{t-1} + \sum_{i=1}^{p-1} \phi_i \Delta P_{t-i} + a_t$$

$$\begin{pmatrix} \Delta P_{t-1} = P_{t-1} - P_{t-2} \\ \Delta P_{t-2} = P_{t-2} - P_{t-3} \\ \vdots \end{pmatrix}$$

$$\bullet H_0: \beta = 1 \quad \text{vs} \quad H_1: \beta < 1$$

$H_0$ : there is a unit root

$$H_0 \Rightarrow \Delta P_t = C_t + \sum_{i=1}^{p-1} \phi_i \Delta P_{t-i} + a_t$$

$$\Delta P_t \sim AR(p-1)$$

$$P_t \sim ARIMA(p-1, 1, 0)$$

- $H_1: \beta < 1 \Rightarrow P_t \sim \text{ARIMA}(p, 0, 0)$   
AR(p)

$$P_t = C_t + \beta P_{t-1} + \phi_1 \Delta P_{t-1} + \phi_2 \Delta P_{t-2} + \dots + \phi_{p-1} \Delta P_{t-(p-1)} + a_t$$

$$= C_t + \beta P_{t-1} + \phi_1 (\underbrace{P_{t-1} - P_{t-2}}_{\Delta P_{t-1}}) + \phi_2 (\underbrace{P_{t-2} - P_{t-3}}_{\Delta P_{t-2}}) + \dots + \phi_{p-1} (\underbrace{P_{t-(p-1)} - P_{t-p}}_{\Delta P_{t-(p-1)}}) + a_t$$

$$= C_t + (\beta + \phi_1) P_{t-1} + (\phi_2 - \phi_1) P_{t-2} + \dots + (-\phi_{p-1}) \underline{P_{t-p}} + a_t$$

$$P_t \sim \text{AR}(p)$$

Augmented Dickey-Fuller test

P value  $> 5\%$   $\Rightarrow P_t$  has a unit root

- Alternative form of the model

$$P_t = C_t + \beta P_{t-1} + \sum_{i=1}^{p-1} \phi_i \Delta P_{t-i} + a_t$$

$\underbrace{P_t - P_{t-1}}_{\Delta P_t} \quad \dots \quad \underbrace{P_{t-1} - P_{t-2}}_{\Delta P_{t-1}} \quad \dots \quad \underbrace{P_{t-p+1} - P_{t-p}}_{\Delta P_{t-p+1}}$

$$\Rightarrow \underline{\Delta P_t} = C_t + (\beta - 1) \underline{P_{t-1}} + \sum_{i=1}^p \phi_i \Delta \underline{P_{t-i}} + a_t$$

$$= C_t + \beta_r \underline{P_{t-1}} + \sum_{i=1}^{p-1} \phi_i \Delta \underline{P_{t-i}} + a_t$$

$$\beta_r \triangleq \beta - 1$$

$$H_0: \beta_r = 0 \quad \text{vs} \quad H_1: \beta_r < 0$$

$$= \beta - 1$$

$$= \beta - 1$$