

Forecasting - Part I

h -Step ahead prediction for **AR** processes

STAT 464 / 864 | Fall 2024

Discrete Time Series Analysis

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Properties of $P(\cdot | W)$ | Review from Week 8 Video Lecture

$$P(Y|W) = \mu_Y + a^T(W - \mu_W)$$

P1) Linearity $P(c_0 + c_1U + c_2V|W) = c_0 + c_1P(U|W) + c_2P(V|W)$

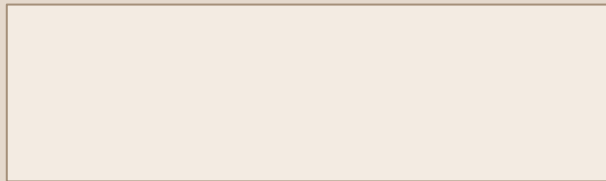
P2) If Y is uncorrelated with the predictors $\{W_i\}$, $P(Y|W) = E[Y]$

P3) Predictors uncorr. with residuals $\text{Cov}(W_i, Y - P(Y|W)) = 0,$

P4) Best LP of any predictor W_i is itself $P(W_i|W) = W_i$

Forecasting

$$P_N X_{N+h} = P(X_{N+h} | (X_N, \dots, X_1)^T)$$



Correlation
matrix of **W**

1-step ahead prediction | AR(1)

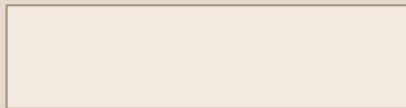
$$X_t = \phi X_{t-1} + Z_t$$

$$Z_t \sim wn(0, \sigma^2)$$

$$\begin{matrix} W = \\ Y = \\ \mu_Y = \\ \rho_X(h) = \end{matrix} \begin{bmatrix} \phi^0 & \phi^1 & \phi^2 & \dots & \phi^{N-1} \\ \phi^1 & & & \dots & \vdots \\ \phi^2 & & \ddots & & \phi^2 \\ \vdots & \vdots & \ddots & \ddots & \phi^1 \\ \phi^{N-1} & \dots & \phi^2 & \phi^1 & \phi^0 \end{bmatrix} \begin{bmatrix} \\ \\ \\ \\ \end{bmatrix} = \begin{bmatrix} \\ \\ \\ \\ \end{bmatrix}$$

ϕ times the first column of R is ρ → Solution:

1-step ahead prediction | AR(1)



$$X_t = \phi X_{t-1} + Z_t$$

$$Z_t \sim wn(0, \sigma^2)$$

$$a = (\phi, 0, \dots, 0)^T$$

$$|\phi| < 1$$

Another way to derive:

$$P_N X_{N+1} =$$

=

=

$$P(X_{N+1} | X_N, \dots, X_1) =$$

1-step ahead prediction | AR(1)

$$X_{N+h} = \phi X_{N+h-1} + Z_{N+h}$$

$$P_N X_{N+h} = \phi P_N X_{N+h-1} + P_N Z_{N+h}$$

$$= \phi P_N X_{N+h-1}$$

$$= \phi(\phi P_N X_{N+h-2})$$

$$= \vdots$$

$$= \phi^h P_N X_{N+h-h} = \phi^h X_N$$

$$X_t = \phi X_{t-1} + Z_t$$

$$Z_t \sim wn(0, \sigma^2)$$

$$a = (\phi, 0, \dots, 0)^T$$

$$|\phi| < 1$$

$$a^T W = \phi X_N$$

1-step ahead prediction | AR(p)

$$X_{N+1} = Z_{N+1} + \sum_{i=1}^p \phi_i X_{N+1-i}$$

AR(p) process

$$1 - \phi_1 z - \phi_2 z^2 - \dots - \phi_p z^p$$

$$P_N X_{N+1} =$$

$$=$$

h-step ahead prediction | AR(p)

$$P_N X_{N+h} =$$

MMSE | AR(1) – 1-step ahead

$$MMSE =$$

$$\gamma_X(0)$$

$$\gamma_X(1)$$

MMSE | AR(1) – h-step ahead

$$MMSE = \text{Var}(Y) - a^T \gamma$$

$$\text{Var}(Y) =$$

$$a =$$

$$a^T \gamma =$$

$$MMSE =$$

MMSE | AR(p) – 1-step ahead

$$MMSE = \text{Var}(Y) - a^T \gamma$$

$$X_{N+1} = \boxed{Z_{N+1}} + \sum_{i=1}^p \phi_i X_{N+1-i}$$

$$X_{N+1} - P_N X_{N+1} =$$

$$MMSE = \text{E}[(X_{N+1} - P_N X_{N+1})^2] =$$

We learned something today, in Time Series 🕒



What do we tell quin?