

Forecasting - Part II b

h -Step ahead prediction for $MA(q)$ processes

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Discrete Time Series Analysis

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We learned something last time, in Time Series 🕒



Predicting $MAC(1)$ using infinitely many past values
so it gets messy if we avoid approximation
approximation should be sufficient

What do we tell quin?

Invertibility

$$X_{N+1} = Z_{N+1} + \theta Z_N$$

$$\tilde{P}_N X_{N+1} = \tilde{P}_N Z_{N+1} + \theta \tilde{P}_N Z_N.$$

Z_N is uncorrelated with all predictors $X_N, X_{N-1}, \dots \rightarrow \tilde{P}_N Z_{N+1} = 0$

Z_N is not uncorrelated with $X_N \rightarrow$ we can't say $\tilde{P}_N Z_N = 0$

Approach: treat X as an ARMA(1,1) process, solve for Z in terms of X

To do this, we must invert the operator $\theta(B)$

Definition: An ARMA(1,1) process with MA coefficient θ is invertible if $|\theta| < 1$

Suppose $\{X_t\}$ is an invertible ARMA(1,1) process:

$$\phi(B)X_t = \theta(B)Z_t$$

$$\theta(B) = (1 + \theta B)$$

$$|\theta| < 1$$

$$\phi(B) =$$

$$\theta(B)^{-1} =$$

$$Z_t =$$

$$=$$

$$=$$

$$= \begin{pmatrix} B^0 \end{pmatrix}$$

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$$=$$

Suppose $\{X_t\}$ is an invertible ARMA(1,1) process:

$$Z_{N+1} = X_{N+1} - (\theta + \phi) \sum_{i=1}^{\infty} (-\theta)^{j-1} X_{N+1-i}$$

$$\phi(B)X_t = \theta(B)Z_t$$

$$\theta(B) = (1 + \theta B)$$


$$|\theta| < 1$$

$$\tilde{P}_N Z_{N+1} =$$

$$0 =$$

$$\tilde{P}_N X_{N+1} =$$

Filter

$$a_j =$$


Algorithms...

Plot twist: there is a 3rd option
(Actually it's a modification of the 1st)

Solving these equations exactly
for $\mathbf{a}_1, \dots, \mathbf{a}_N$ is curiously difficult

- 1) Solve numerically
- 2) Approximation

Text discusses 2 **algorithms** for exact numerical estimates, using sample ACVFs.



Durbin-Levinson



Innovations

We'll be using this one, since it's more popular nowadays,
and it's implemented by the `forecast()` function from Workshop 4.

MMSE: ARMA(1,1)

$$\text{MSE} = \text{E}[(Y - P(Y|W))^2]$$

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

$=$

\approx



This term is small,
and decreases as N gets larger