

Economics 413: Economic Forecast & Analysis

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Final – Answers

1. For $t < t'$, $Y_t = c + \varepsilon_t + \theta\varepsilon_{t-1}$ and for $t \geq t'$, $Y_t = (c + \psi) + \varepsilon_t + \theta\varepsilon_{t-1}$
 - (a) $E(Y_t) = c$ and $E(Y_t) = (c + \psi)$, respectively
 - (b) $V(Y_t) = (1 + \theta^2)\sigma^2$ for each as the MA(1)'s only differ by means
 - (c) Similarly, $\gamma_1 = \theta\sigma^2$ and $\gamma_j = 0$ for all $j > 1$ for each
 - (d) $h = 1$, $\hat{Y}_{t+1|t} = E(Y_{t+1}|\Omega_t) = E[(c + \psi) + \varepsilon_{t+1} + \theta\varepsilon_t|\Omega_t] = (c + \psi) + \theta\varepsilon_t$ and $\hat{Y}_{t+h|t} = E(Y_{t+h}|\Omega_t) = E[(c + \psi) + \varepsilon_{t+h} + \theta\varepsilon_{t+h-1}|\Omega_t] = (c + \psi) \forall h > 1$
 - (e) $e_{t+1} = Y_{t+1} - \hat{Y}_{t+1|t} = \varepsilon_{t+1} \Rightarrow \hat{Y}_{t+1|t} \pm 2\sigma$ and $e_{t+h} = Y_{t+h} - \hat{Y}_{t+h|t} = \varepsilon_{t+h} + \theta\varepsilon_{t+h-1} \Rightarrow \hat{Y}_{t+h|t} \pm 2\sigma\sqrt{(1 + \theta^2)} \forall h > 1$
2.
 - (a) ARIMA(0,1,1)
 - (b) The ACF should have high persistence (lots of spikes near one) and the PACF should have a single large significant spike (near one)
 - (c) The figure should look like a random walk
 - (d) The point estimates should be $y_t + \hat{\theta}\varepsilon_t$ for $h = 1, 2, 3$
 - (e) The forecast intervals should be $y_t + \hat{\theta}\varepsilon_t \pm 2\hat{\sigma}$, $y_t + \hat{\theta}\varepsilon_t \pm 2\hat{\sigma}\sqrt{(2 + \theta^2)}$ and $y_t + \hat{\theta}\varepsilon_t \pm 2\hat{\sigma}\sqrt{(3 + \theta^2)}$, respectively
3.
 - (a) $\mathcal{L}(\cdot) = -\frac{T}{2} \ln 2\Pi - \frac{T}{2} \ln \sigma^2 - \frac{1}{2\sigma^2} \sum_{t=2}^T \varepsilon_t^2$, where $\varepsilon_t = y_t - c - \phi y_{t-1} - \theta\varepsilon_{t-1}$ and $\varepsilon_t = y_t - c - \phi y_{t-1}$ for the true model and your model, respectively
 - (b) The sample ACF and PACF should look like white noise for the true model and the sample ACF and PACF should look like an AR(1) for your model
 - (c) The null in each case is that of no serial correlation. We should fail to reject the null in the true model and reject the null in your model.
 - (d) The point forecasts will be off by $\hat{\theta}\varepsilon_t$ for $h = 1$, but will be the same for $h = 2$ and 3
 - (e) The variance will be too small for your model as compared to the true model and hence the forecast intervals will suggest more confidence than they should

4. (a) y , our outcome of interest (growth rate of GDP for Sierra Leone)
- (b) OLS, but the model is likely estimated via MLE
- (c) $t = 1, 2, \dots, T$
- (d) \hat{c} - coefficient estimate
- (e) $\hat{\phi}$ - coefficient estimate
- (f) $\hat{\theta}$ - coefficient estimate
- (g) $se(\hat{c})$ - variation in the estimate
- (h) $se(\hat{\phi})$ - variation in the estimate
- (i) $se(\hat{\theta})$ - variation in the estimate
- (j) $t = (\hat{c} - 0) / se(\hat{c})$
- (k) $t = (\hat{\phi} - 0) / se(\hat{\phi})$
- (l) $t = (\hat{\theta} - 0) / se(\hat{\theta})$
- (m) fail to reject null ($H_0 : c = 0$) at 5% level
- (n) fail to reject null ($H_0 : \phi = 0$) at 5% level
- (o) fail to reject null ($H_0 : \theta = 0$) at 5% level
- (p) $1 - SSR/SST$
- (q) $1 - [SSR / (T - 3)] / [SST / (T - 1)]$
- (r) $\hat{\sigma} = \sqrt{\frac{1}{T-3} SSR}$
- (s) $SSR = \sum_{t=1}^T \hat{\varepsilon}_t^2$
- (t) $\mathcal{L}(\cdot) = -\frac{T}{2} \ln 2\pi - \frac{T}{2} \ln \hat{\sigma}^2 - \frac{1}{2\hat{\sigma}^2} \sum_{t=2}^T \hat{\varepsilon}_t^2$
- (u) $H_0 : \phi = \theta = 0$, $F = \frac{(SSR_R - SSR_U)/3}{SSR_U/(T-3)}$
- (v) $\bar{y} = \frac{1}{T} \sum_{t=1}^T y_t$
- (w) $\hat{\sigma}_y = \sqrt{\frac{1}{T-1} \sum_{t=1}^T (y_t - \bar{y})^2}$
- (x) $AIC = T \ln(SSR/T) + 2(k + 1)$
- (y) $SC = T \ln(SSR/T) + (k + 1)$