

Answers to Exam No. 1 on Topics from Chapters 2 through 9

$$1. \quad SR_t = RFR_t + a MR_t - a RFR_t + v_t = RFR_t(1 - a) + a MR_t + v_t$$

Therefore, $b_1 = 0$, $b_2 = a$, and $b_3 = 1 - a$. The relevant restrictions are $b_1 = 0$ and $b_2 + b_3 = 1$.

2. First regress SR_t against a constant, MR_t , and RFR_t , and save the error sum of squares as $ESSA$. Next generate $Y_t = SR_t - RFR_t$ and $X_t = MR_t - RFR_t$. Then regress Y_t against X_t without a constant term and save the error sum of squares as $ESSB$.

$$3. \quad \text{Compute } F_c = \frac{(ESSB - ESSA)/2}{ESSA/(52 - 3)}.$$

4. Under the null hypothesis, F_c has the F-distribution with 2 d.f. for the numerator and 49 d.f. for the denominator.

5. Look up the F-table and find $F_{2,49}^*$, the point at which the area to the right is 0.05. From the table, we interpolate it to be approximately 3.19. Reject the null hypothesis that $b_1 = 0$ and $b_2 + b_3 = 1$ if $F_c > 3.19$.

6. The transformed model is:

$$(C) \quad SR_t - rSR_{t-1} = b_1(1 - r) + b_2(MR_t - rMR_{t-1}) + b_3(RFR_t - rRFR_{t-1}) + e_t$$

Step 1 Fix the value of r . Then generate $SR^* = SR_t - rSR_{t-1}$, $MR^* = MR_t - rMR_{t-1}$ and $RFR^* = RFR_t - rRFR_{t-1}$.

Step 2 Regress SR^* against a constant, MR^* , and RFR^* , and get \hat{b}_1 , \hat{b}_2 , and \hat{b}_3 .

Step 3 Vary r at broad steps from -0.99 through +0.99, say, at steps of length 0.1. Choose the \hat{r} that minimizes ESSC as the starting point of a Cochrane-Orcutt iteration.

Step 4 Repeat Step 2 after using this \hat{r} in Step 1 and compute

$$\hat{u}_t = SR_t - \hat{b}_1 - \hat{b}_2 MR_t - \hat{b}_3 RFR_t.$$

Step 5 Get new estimate $\hat{r} = \frac{\sum \hat{u}_t \hat{u}_{t-1}}{\sum \hat{u}_{t-1}^2}$.

Step 6 Repeat Steps 1, 2, 4, and 5, using new \hat{r} values and iterate \hat{r} from two successive iterations do not change by more than say 0.001.

Step 7 Using this final \hat{r} estimate Model C.

7.

Testing

- Step 1 Regress SR_t against a constant, MR_t , and RFR_t , and save the error term as \hat{u}_t .
- Step 2 Regress \hat{u}_t against a constant, \hat{u}_{t-1} , \hat{u}_{t-2} , \hat{u}_{t-3} , \hat{u}_{t-4} , MR_t , and RFR_t using observations 5 through 52.
- Step 3 Compute $LM = 52 R^2$, where R^2 is the unadjusted goodness of fit in Step 2.
- Step 4 Reject the null hypothesis of no serial correlation if $LM > 9.48773$, the point on the Chi-square distribution with 4 d.f. with an area of 0.05 to the right of it.

Estimation

- Step 1 Regress SR_t against a constant, MR_t , and RFR_t , and save the error term as \hat{u}_t .
- Step 2 Regress \hat{u}_t against a constant, \hat{u}_{t-1} , \hat{u}_{t-2} , \hat{u}_{t-3} , and \hat{u}_{t-4} , using observations 5 through 52. This gives the AR parameters \hat{r}_1 , \hat{r}_2 , \hat{r}_3 , and \hat{r}_4 .
- Step 3 Generate the new variables
- $$SR^* = SR_t - \hat{r}_1 SR_{t-1} - \hat{r}_2 SR_{t-2} - \hat{r}_3 SR_{t-3} - \hat{r}_4 SR_{t-4}$$
- $$MR^* = MR_t - \hat{r}_1 MR_{t-1} - \hat{r}_2 MR_{t-2} - \hat{r}_3 MR_{t-3} - \hat{r}_4 MR_{t-4}$$
- $$RFR^* = RFR_t - \hat{r}_1 RFR_{t-1} - \hat{r}_2 RFR_{t-2} - \hat{r}_3 RFR_{t-3} - \hat{r}_4 RFR_{t-4}$$
- Step 4 Regress SR^* against a constant, MR^* , and RFR^* , and get \hat{b}_1 , \hat{b}_2 , and \hat{b}_3 . Note That, to get \hat{b}_1 you should divide the constant term obtain in Step 4 by $1 - \hat{r}_1 - \hat{r}_2 - \hat{r}_3 - \hat{r}_4$.
- Step 5 Compute $\hat{u}_t = SR_t - \hat{b}_1 - \hat{b}_2 MR_t - \hat{b}_3 RFR_t$
- Step 6 Go back to Step 2 and iterate until the stopping rule given below applies.
- Step 7 Stop when the error sum of squares from Step 4 does not change by more than a pre-specified percent, say 0.01.