

FinQuiz Formula Sheet CFA Program Level II

Reading 1: Introduction to Linear Regression

- Linear Regression = $Y_i = b_0 + b_1X_i + \varepsilon_i$,
- Sum of square error (SSE) = $\sum_{i=1}^n (y_i - \hat{y})^2$
- $\hat{b}_1 = \frac{\text{cov}(x,y)}{\text{var}(x)} = \frac{\sum(X_i - \bar{X})(Y_i - \bar{Y})}{\sum(X_i - \bar{X})^2}$
- $\hat{b}_0 = \bar{Y} - \hat{b}_1 \bar{X}$
- Sample correlation $r = \frac{\text{Covariance of } Y \text{ and } X}{(\text{standard deviation of } Y)(\text{standard deviation of } X)}$
- $\text{Cov}_{XY} = \frac{\sum_{i=1}^n (Y_i - \bar{Y})(X_i - \bar{X})}{n-1}$
- Standard deviation of $Y = S_Y = \sqrt{\frac{\sum_{i=1}^n (Y_i - \bar{Y})^2}{n-1}}$
- Squared residuals $E(\varepsilon_i^2) = \sigma_i^2$, $i = 1, \dots, n$
- Sum of square regression (SSR) = $\sum_{i=1}^n (\hat{y}_i - \bar{y})^2$
- Coefficient of Determination (R^2) = $\frac{\text{Sum of square regression}}{\text{Sum of square total}} = \frac{\sum_{i=1}^n (\hat{y}_i - \bar{y})^2}{\sum_{i=1}^n (y_i - \bar{y})^2}$
(for single independent variable $R^2 = r^2$)
- Mean square regression MSR = $\frac{\text{Sum of square regression}}{k} = \frac{\sum_{i=1}^n (\hat{y}_i - \bar{y})^2}{1}$
- Mean square error MSE = $\frac{\text{sum of squares error}}{n-k-1}$
- F-Statistic or F-Test = $\frac{MSR}{MSE} = \frac{(\frac{\text{Sum of square regression}}{k})}{(\frac{\text{Sum of squares error}}{n-k-1})}$
(df numerator = $k = 1$)
(df denominator = $n - k - 1 = n - 2$)

14. ANOVA

ANOVA	SS	MSS	F
Regression df = 1	$SSR = \sum_{i=1}^n (\hat{y}_i - \bar{y})^2$	$\frac{SSR}{k}$	$\frac{SSR/k}{SSE/(n-k-1)}$
Error df = n-2	$SSE = \sum_{i=1}^n (y_i - \hat{y})^2$	$\frac{SSE}{n-k-1}$	
Total df = n-1	$SST = \sum_{i=1}^n (y_i - \bar{y})^2$		

- Test statistic $t = \frac{\hat{b}_1 - B_1}{s_{\hat{b}_1}}$
- Standard error of slope coefficient = $s_{\hat{b}_1} = \frac{s_e}{\sqrt{\sum_{i=1}^n (X_i - \bar{X})^2}}$
- Standard error of the intercept = $s_{\hat{b}_1} = \sqrt{\frac{1}{n} + \frac{\bar{X}^2}{\sum_{i=1}^n (X_i - \bar{X})^2}}$

18. Forecasted value of dependent variable =
 $\hat{Y}_f = \hat{b}_0 + \hat{b}_1 X_f$

19. Standard error of the intercept = $s_f =$

$$s_e \sqrt{1 + \frac{1}{n} + \frac{(X_f - \bar{X})^2}{\sum_{i=1}^n (X_i - \bar{X})^2}}$$

20. Log-lin Model = $\ln Y_i = b_0 + b_1 X_i$

21. Lin-log Model = $Y_i = b_0 + \ln X_i$

22. Log-log Model = $\ln Y_i = b_0 + b_1 \ln X_i$

Reading 2: Multiple Regression

1. $Y_i = b_0 + b_1 X_{1i} + b_2 X_{2i} + \dots + b_k X_{ki} + \varepsilon_i, i = 1, 2, \dots, n$

2. Prediction equation = $\hat{Y}_i = \hat{b}_0 + \hat{b}_1 X_{1i} + \hat{b}_2 X_{2i} + \dots + \hat{b}_k X_{ki} + \varepsilon_i, i$

3. Adjusted $R^2 = \bar{R}^2 = 1 - \left(\frac{n-1}{n-k-1}\right) (1 - R^2)$

4. Breusch-Pagan test

- H_0 = No conditional Heteroskedasticity exists
- H_A = Conditional Heteroskedasticity exists
- Test statistic = $n \times R^2_{\text{residuals}}$

5. Durbin-Waston Test = $DW =$

$$\frac{\sum_{t=2}^T (\hat{\varepsilon}_t - \hat{\varepsilon}_{t-1})^2}{\sum_{t=1}^T \hat{\varepsilon}_t^2}$$

- For Large Sample size DW Statistic
 $(d) = d \approx 2 (1 - r)$

Reject H_0 , conclude Positive Serial Correlation		Do not reject H_0		Reject H_0 , conclude Negative Serial Correlation	
0	d_L	d_U	$4 - d_U$	$4 - d_L$	4

Reading 3: Time Series Analysis

1. Linear Trend Models = $y_t = b_0 + b_1 t + \varepsilon_t$

- Predicted/fitted value of y_t in period $(T + 1) = \hat{y}_{t+1} = \hat{b}_0 + \hat{b}_1 (T + 1)$

2. Log-Linear Trend Models = $y_t = e^{b_0 + b_1 t}$

3. Autoregressive Time-Series Models:

- First order autoregressive AR (1) = $x_t = b_0 + b_1 x_{t-1} + \varepsilon_t$
- pth-order autoregressive AR (p) = $x_t = b_0 + b_1 x_{t-1} + b_2 x_{t-2} + \dots + b_p x_{t-p} + \varepsilon_t$

4. Mean reverting level of $x_t = \frac{b_0}{1 - b_1}$

5. Chain Rule of Forecasting:

- One-period ahead forecast =
 $\hat{x}_{t+1} = \hat{b}_0 + \hat{b}_1 x_t$
- Two-period ahead forecast =
 $\hat{x}_{t+2} = \hat{b}_0 + \hat{b}_1 x_{t+1}$

6. Random Walks and Unit Roots:

- Random Walk without drift = $x_t = x_{t-1} + \varepsilon_t$ where, $b_0 = 0$ and $b_1 = 1$.

- Correcting Random Walk = $y_t = x_t - x_{t-1}$
- Random walk with a drift = $x_t = b_0 + x_{t-1} + \varepsilon_t$ where, $b_0 \neq 0$ and $b_1 = 1$
- By taking first difference $y_t = x_t - x_{t-1} = b_0 + \varepsilon_t$

7. Using Dickey-Fuller Test = $x_t - x_{t-1} = b_0 + (b_1 - 1) x_{t-1} + \varepsilon_t$

8. Smoothing Past Values with n-Period Moving Average =

$$\frac{x_t + x_{t-1} + x_{t-2} + \dots + x_{t-(n-1)}}{n}$$

9. Correcting Seasonality in Time Series Models:

- For quarterly data = $x_t = b_0 + b_1 x_{t-1} + b_2 x_{t-4} + \varepsilon_t$
- For monthly data = $x_t = b_0 + b_1 x_{t-1} + b_2 x_{t-12} + \varepsilon_t$

10. ARCH model =

$\hat{\varepsilon}_t^2 = \alpha_0 + \alpha_1 \hat{\varepsilon}_{t-1}^2 + \mu_t$ where μ_t is an error term

- Predicting variance of errors in period $t+1 = \hat{\sigma}_{t+1}^2 = \hat{\alpha}_0 + \alpha_1 \hat{\varepsilon}_t^2$

Reading 4: Machine Learning

LASSO:

1. Penalty term (when $\lambda > 0$) = $\lambda \sum_{k=1}^K |\bar{b}_k|$

- $\sum_{i=1}^n (Y_i - \hat{Y}_i)^2 + \lambda \sum_{k=1}^K |\hat{\beta}_k|$
- When $\lambda = 0$, LASSO penalized regression = OLS regression

Reading 5: Big Data Projects

- $X_{i(normalized)} = \frac{X_i - X_{min}}{X_{max} - X_{min}}$

where X_i = value of observation

Performance Metrics:

- Accuracy = (TP + TN)/(TP + FP + TN + FN)
F1 score = (2*P*R)/(P + R)
- Receiver Operating Characteristic (ROC):
False positive rate (FPR) = FP/(TN + FP) and
True positive rate (TPR) = TP/(TP + FN), which is same as recall
- Root Mean Square Error (RMSE):

$$\sum_{i=1}^n \frac{(Predicted_i - Actual_i)^2}{n}$$

Reading 6: Currency Exchange Rates

- Bid-offer Spread = Offer price – Bid price
- Fwd rate = Spot Exchange rate + $\frac{\text{Forward points}}{10,000}$
- Forward premium/discount (in %) = $\frac{\text{spot exchange rate} - (\text{forward points}/10,000)}{\text{spot exchange rate}} - 1$

- To convert spot rate into forward quote:
 - Spot exchange rate \times (1 + % premium)
 - Spot exchange rate \times (1 - % discount)
- Covered interest rate parity:

- $(1 + i_d) = S_{f/d} (1 + i_f) \left(\frac{1}{F_{f/d}} \right)$

- $F_{f/d} = S_{f/d} \left(\frac{1+i_f}{1+i_d} \right)$

- Using day count convention:

$$\left(1 + i_d \left[\frac{\text{Actual}}{360} \right] \right) =$$

$$S_{f/d} \left(1 + i_f \left[\frac{\text{Actual}}{360} \right] \right) \left(\frac{1}{F_{f/d}} \right)$$

- $F_{f/d} = S_{f/d} \left(\frac{1 + i_f \left[\frac{\text{Actual}}{360} \right]}{1 + i_d \left[\frac{\text{Actual}}{360} \right]} \right)$

- Uncovered Interest Rate Parity :

- $i_f - \% \Delta S_{f/d}^e = i_d$

- $\% \Delta S_{f/d}^e = i_f - i_d$

- Forward premium or discount:

- For one year horizon =

$$F_{f/d} - S_{f/d} =$$

$$S_{f/d} \left(\frac{i_f - i_d}{1 + i_d} \right) \cong S_{f/d} (i_f - i_d)$$

- Using day count convention:

$$F_{f/d} - S_{f/d} = S_{f/d} \left(\frac{\left[\frac{\text{Actual}}{360} \right]}{1 + i_d \left[\frac{\text{Actual}}{360} \right]} \right) (i_f - i_d)$$

- Forward discount or premium as % of spot rate:

$$\frac{F_{f/d} - S_{f/d}}{S_{f/d}} \cong (i_f - i_d)$$

If uncovered interest rate parity holds

- $= \frac{F_{f/d} - S_{f/d}}{S_{f/d}} = \% \Delta S_{f/d}^e \cong (i_f - i_d)$

- Purchasing Power parity (PPP)

- $P_f = S_{f/d} \times P_d$
- $S_{f/d} = P_f / P_d$

- Relative version of PPP = $\% \Delta S_{f/d} = \pi_f - \pi_d$

- Ex ante version of PPP = $\% \Delta S_{f/d}^e = \pi_f^e - \pi_d^e$

- Real Exchange Rate

$$q_{f/d} = \left(\frac{S_{f/d} P_d}{P_f} \right) = S_{f/d} \left(\frac{P_d}{P_f} \right)$$

$$q_{f/d} = S_{f/d} \left(\frac{CPI_d}{CPI_f} \right)$$

or

- Fisher effect:

- $i_d = r_d + \pi_d^e$
- $i_f = r_f + \pi_f^e$