

# FinQuiz Formula Sheet CFA Program Level II

## Reading 4: Introduction to Regression

1. Sample Cov (X, Y) =  $\frac{\sum_{i=1}^n (X_i - \bar{X})(Y_i - \bar{Y})}{n-1}$
2. Correlation Coefficient =  $r_{XY} = \frac{\text{cov}_{XY}}{(s_X)(s_Y)}$  or  $r = \frac{\text{cov}(X,Y)}{\sqrt{\text{var}(X)}\sqrt{\text{var}(Y)}}$
3. t-test (for normally distributed variables) =  $t = \frac{r\sqrt{n-2}}{\sqrt{1-r^2}}$  t distribution with (n - 2) deg. of freedom
4. Linear Regression =  $Y_i = b_0 + b_1X_i + \epsilon_i$ 
  - Intercept ( $b_0$ ) =  $\bar{b}_0 = \bar{y} - b_1\bar{x}$
  - Slope or regression coefficient =  $b_1 = \frac{\text{cov}(x,y)}{\text{var}(x)}$  or  $= \frac{\sum(x-\bar{x})(y-\bar{y})}{\sum(x-\bar{x})^2}$
5. Standard Error of Estimate SEE =  $S_E = \sqrt{\frac{SSE}{n-k-1}} = \sqrt{\frac{\sum_{i=1}^n (y_i - \hat{y})^2}{n-k-1}}$
6. Coefficient of Determination ( $R^2$ ) =

$$= \frac{SST - SSE}{SST} = \frac{RSS}{SST} \text{ where, } 0 \leq R^2 \leq 1$$

(for single independent variable  $R^2 = r^2$ )

7. SST = SSE + SSR(or RSS)
8. Hypothesis Testing:
  - Null and Alternative hypotheses
  - $H_0: b_1 = 0$  (no linear relationship)
  - $H_1: b_1 \neq 0$  (linear relationship does exist)
  - Test statistic =  $t = \frac{\hat{b}_1 - b_1}{S_{b_1}}$
  - Confidence Interval =  $b_1 \pm t_c S_{b_1}$

9. ANOVA (Analysis of variance) =

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

Total df = n-1	$SST = \sum_{i=1}^n (y_i - \bar{y})^2$		
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Source of Variability	DoF	Sum of Squares	Mean Sum of Squares
Regression (Explained)	1	RSS	MSR = RSS/1
Error (Unexplained)	n-2	SSE	MSE = SSE/n-2
Total	n-1	SST = RSS + SSE	

10. F-Statistic or F-Test =  $\frac{MSR}{MSE} = \frac{(\frac{RSS}{k})}{(\frac{SSE}{n-k-1})}$   
 (df numerator = k = 1)  
 (df denominator = n - k - 1 = n - 2)

11. Prediction Intervals =  $\hat{Y} \pm t_c S_f$

where  $s_f^2 = s^2 \left[ 1 + \frac{1}{n} + \frac{(X-\bar{X})^2}{(n-1)s_x^2} \right]$  and

$$s_f = \sqrt{s_f^2}$$

**Reading 5: Multiple Regression & Issues in Regression Analysis**

- $Y_i = b_0 + b_1X_{1i} + b_2X_{2i} + \dots + b_kX_{ki} + \varepsilon_i, i = 1, 2, \dots, n$
- Prediction equation  $= \hat{Y}_i = \hat{b}_0 + \hat{b}_1X_{1i} + \hat{b}_2X_{2i} + \dots + \hat{b}_kX_{ki} + \varepsilon_i, i$
- Adjusted  $R^2 = \bar{R}^2 = 1 - \left( \frac{n-1}{n-k-1} \right) (1 - R^2)$
- Breusch-Pagan test
  - $H_0 =$  No conditional Heteroskedasticity exists
  - $H_A =$  Conditional Heteroskedasticity exists
  - Test statistic  $= n \times R^2_{\text{residuals}}$
- 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	
Conclusive	Inconclusive	Inconclusive	Conclusive	Conclusive	Inconclusive
0	$d_L$	$d_U$	$4-d_U$	$4-d_L$	4

**Reading 6: Time Series Analysis**

- Linear Trend Models  $= y_t = b_0 + b_1t + \varepsilon_t$

- Predicted/fitted value of  $y_t$  in period  $(T + 1) = \hat{y}_{t+1} = \hat{b}_0 + \hat{b}_1(T + 1)$
- Log-Linear Trend Models  $= y_t = e^{b_0 + b_1t}$
  - 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)  $= x_t = b_0 + b_1 x_{t-1} + b_2 x_{t-2} + \dots + b_p x_{t-p} + \varepsilon_t$
  - Mean reverting level of  $x_t = \frac{b_0}{1 - b_1}$
  - Chain Rule of Forecasting:
    - One-period ahead forecast  $= \hat{x}_{t+1} = \hat{b}_0 + \hat{b}_1 x_t$
    - Two-period ahead forecast  $= \hat{x}_{t+2} = \hat{b}_0 + \hat{b}_1 x_{t+1}$
  - 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  $y_t = x_t - x_{t-1} = b_0 + \varepsilon_t$
  - Using Dickey-Fuller Test  $= x_t - x_{t-1} = b_0 + (b_1 - 1) x_{t-1} + \varepsilon_t$

- Smoothing Past Values with n-Period Moving Average  $= \frac{x_t + x_{t-1} + x_{t-2} + \dots + x_{t-(n-1)}}{n}$
- 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  $= x_t = b_0 + b_1 x_{t-1} + b_2 x_{t-12} + \varepsilon_t$
- ARCH model  $= \hat{\varepsilon}_t^2 = \alpha_0 + \alpha_1 \hat{\varepsilon}_{t-1}^2 + \mu_t$  where  $\mu_t$  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 7: Machine Learning**

LASSO:

- Penalty term (when  $\lambda > 0$ )  $= \lambda \sum_{k=1}^K |\bar{b}_k|$
- $\sum_{i=1}^n (Y_i - Y_i)^2 + \lambda \sum_{k=1}^K |\hat{b}_k|$
- When  $\lambda = 0$ , LASSO penalized regression = OLS regression

**Reading 8: Big Data Projects**

$$1. X_{i(\text{normalized})} = \frac{X_i - X_{\min}}{X_{\max} - X_{\min}}$$

where  $X_i =$  value of observation

**Performance Metrics:**

$$2. \text{ Accuracy} = (TP + TN) / (TP + FP + TN + FN)$$

$$\text{F1 score} = (2 * P * R) / (P + R)$$

$$3. \text{ Receiver Operating Characteristic (ROC):}$$

False positive rate (FPR) = FP / (TN + FP)  
and

True positive rate (TPR) = TP / (TP + FN),  
which is same as recall

$$4. \text{ Root Mean Square Error (RMSE):}$$

$$\sum_{i=1}^n \frac{(\text{Predicted}_i - \text{Actual}_i)^2}{n}$$

Reading 9: Excerpt from 'Probabilistic Approaches, Scenario Analysis, Decision Tree & Simulations'

**Reading 10: Currency Exchange Rates**

$$1. \text{ Bid-offer Spread} = \text{Offer price} - \text{Bid price}$$

$$2. \text{ Fwd rate} = \text{Spot Exchange rate} + \frac{\text{Forward points}}{10,000}$$

$$3. \text{ Forward premium/discount (in \%)} = \frac{\text{spot exchange rate} - (\text{forward points} / 10,000)}{\text{spot exchange rate}} - 1$$

$$4. \text{ To convert spot rate into forward quote:}$$

- Spot exchange rate  $\times$  (1 + % premium)
- Spot exchange rate  $\times$  (1 - % discount)

$$5. \text{ Covered interest rate parity:}$$

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

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

Using day count convention:

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

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

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

6. Uncovered Interest Rate Parity :

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

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

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:

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

7. 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

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

8. Purchasing Power parity (PPP)

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

$$\bullet S_{f/d} = P_f / P_d$$

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

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

11. 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

12. Fisher effect:

$$\bullet i_d = r_d + \pi_d^e$$

$$\bullet i_f = r_f + \pi_f^e$$

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

$$\bullet (r_f - r_d) = (i_f - i_d) - (\pi_f^e - \pi_d^e)$$