

## Test Exercise 3: Answers to the Questions

(a) Use *general-to-specific* to come to a model. Start by regressing the *federal funds rate* on the other 7 variables and eliminate 1 variable at a time.

- Starting with the all the 7 explanatory variables and dropping the variables 1 at a time (w.r.t. the lowest **p-value**) we get the following:

	(Intercept)	INFL	PROD	UNEMPL	COMPRI	PCE	PERSINC	HOUST	
coeff	-0.2147	0.6968	-0.0576	0.1041	-0.0055	0.3426	0.2464	-0.0194	
p.val	0.3808	0	0.1489	0.2816	0.0636	0	1e-04	0	Drop UNEMPL
R <sup>2</sup>	0.639								

	(Intercept)	INFL	PROD	COMPRI	PCE	PERSINC	HOUST	
coeff	-0.2856	0.6942	-0.0248	-0.0065	0.3671	0.2512	-0.021	
p.val	0.2263	0	0.3357	0.0209	0	0	0	Drop PROD
R <sup>2</sup>	0.638							

	(Intercept)	INFL	COMPRI	PCE	PERSINC	HOUST	
coeff	-0.2363	0.7177	-0.0075	0.3398	0.2402	-0.0205	
p.val	0.3051	0	0.0046	0	1e-04	0	All variables are significant at 5% level
R <sup>2</sup>	0.637						

```
##
## Call:
## lm(formula = INTRATE ~ INFL + COMPRI + PCE + PERSINC + HOUST,
##     data = df)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -7.1918 -1.5298 -0.0974  1.3829  7.6603
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -0.236287   0.230212  -1.026  0.30509
## INFL         0.717720   0.056972  12.598 < 2e-16 ***
## COMPRI      -0.007499   0.002639  -2.842  0.00463 **
## PCE          0.339822   0.058989   5.761 1.29e-08 ***
## PERSINC      0.240157   0.059265   4.052 5.68e-05 ***
## HOUST       -0.020519   0.004382  -4.682 3.45e-06 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.188 on 654 degrees of freedom
## Multiple R-squared:  0.6374, Adjusted R-squared:  0.6347
## F-statistic: 230 on 5 and 654 DF, p-value: < 2.2e-16
```

**Final Model**

```
## Start: AIC=1041.17
## INTRATE ~ INFL + PROD + UNEMPL + COMMPRI + PCE + PERSINC + HOUST
##
##      Df Sum of Sq    RSS    AIC
## - UNEMPL 1      5.56 3125.4 1040.3
## <none>      3119.8 1041.2
## - PROD 1      9.99 3129.8 1041.3
## - COMMPRI 1     16.52 3136.3 1042.7
## - PERSINC 1     79.31 3199.1 1055.7
## - HOUST 1     82.70 3202.5 1056.4
## - PCE 1    117.21 3237.0 1063.5
## - INFL 1    603.72 3723.5 1155.9
##
## Step: AIC=1040.34
## INTRATE ~ INFL + PROD + COMMPRI + PCE + PERSINC + HOUST
##
##      Df Sum of Sq    RSS    AIC
## - PROD 1      4.44 3129.8 1039.3
## <none>      3125.4 1040.3
## - COMMPRI 1     25.67 3151.0 1043.7
## - PERSINC 1     82.88 3208.2 1055.6
## - HOUST 1    108.45 3233.8 1060.8
## - PCE 1    150.68 3276.0 1069.4
## - INFL 1    600.15 3725.5 1154.3
##
## Step: AIC=1039.28
## INTRATE ~ INFL + COMMPRI + PCE + PERSINC + HOUST
##
##      Df Sum of Sq    RSS    AIC
## <none>      3129.8 1039.3
## - COMMPRI 1     38.65 3168.4 1045.4
## - PERSINC 1     78.58 3208.4 1053.7
## - HOUST 1    104.91 3234.7 1059.0
## - PCE 1    158.82 3288.6 1070.0
## - INFL 1    759.48 3889.3 1180.7
```

#### Stepwise Backward Regression using AIC

```
##
## Call:
## lm(formula = INTRATE ~ INFL + COMMPRI + PCE + PERSINC + HOUST,
##     data = df)
##
## Coefficients:
## (Intercept)      INFL      COMMPRI          PCE      PERSINC
## -0.236287      0.717720     -0.007499      0.339822      0.240157
##      HOUST
## -0.020519
```

#### Final Model (using AIC)

```
## Start: BIC=1077.1
## INTRATE ~ INFL + PROD + UNEMPL + COMMPRI + PCE + PERSINC + HOUST
##
##      Df Sum of Sq    RSS    BIC
## - UNEMPL 1      5.56 3125.4 1071.8
## - PROD 1      9.99 3129.8 1072.7
## - COMMPRI 1     16.52 3136.3 1074.1
## <none>      3119.8 1077.1
## - PERSINC 1     79.31 3199.1 1087.2
## - HOUST 1     82.70 3202.5 1087.9
## - PCE 1    117.21 3237.0 1095.0
## - INFL 1    603.72 3723.5 1187.4
##
## Step: AIC=1071.79
## INTRATE ~ INFL + PROD + COMMPRI + PCE + PERSINC + HOUST
##
##      Df Sum of Sq    RSS    BIC
## - PROD 1      4.44 3129.8 1066.2
## - COMMPRI 1     25.67 3151.0 1070.7
## <none>      3125.4 1071.8
## - PERSINC 1     82.88 3208.2 1082.6
## - HOUST 1    108.45 3233.8 1087.8
## - PCE 1    150.68 3276.0 1096.4
## - INFL 1    600.15 3725.5 1181.2
##
## Step: AIC=1066.23
## INTRATE ~ INFL + COMMPRI + PCE + PERSINC + HOUST
##
##      Df Sum of Sq    RSS    BIC
## <none>      3129.8 1066.2
## - COMMPRI 1     38.65 3168.4 1067.8
## - PERSINC 1     78.58 3208.4 1076.1
## - HOUST 1    104.91 3234.7 1081.5
## - PCE 1    158.82 3288.6 1092.4
## - INFL 1    759.48 3889.3 1203.1
```

#### Stepwise Backward Regression using BIC

```
##
## Call:
## lm(formula = INTRATE ~ INFL + COMMPRI + PCE + PERSINC + HOUST,
##     data = df)
##
## Coefficients:
## (Intercept)      INFL      COMMPRI          PCE      PERSINC
## -0.236287      0.717720     -0.007499      0.339822      0.240157
##      HOUST
## -0.020519
```

#### Final Model (using BIC)

(b) Use *specific-to-general* to come to a model. Start by regressing the *federal funds rate* on only a constant and add 1 variable at a time. Is the model the same as in (a)?

- Starting regression only with constant and adding 1 variable at a time (the most significant one, i.e., with the least p-value) we obtain the following:

	INFL	PROD	UNEMPL	COMPRI	PCE	PERSINC	HOUST
intercept	1.644129e+00	5.394704e+00	4.542960e+00	5.401737e+00	-3.331470e-01	5.126281e+00	5.403626e+00
coef	9.446992e-01	-1.610555e-02	4.534938e-01	-1.152313e-02	8.289447e-01	1.034739e-01	-3.095185e-02
intercept.pval	1.932206e-23	1.591793e-139	4.255914e-95	6.372498e-169	2.371635e-01	3.017588e-95	5.054543e-173
coef.pval	2.291631e-119	5.870581e-01	1.978701e-10	6.133917e-03	7.367723e-80	1.502290e-01	4.313714e-07
R^2	5.598580e-01						

```
## Start: AIC=1698.89
## INTRATE ~ 1
##
## Df Sum of Sq RSS AIC
## + INFL 1 4833.0 3799.5 1159.3
## + PCE 1 3625.0 5007.5 1341.5
## + UNEMPL 1 515.5 8117.0 1660.2
## + HOUST 1 329.0 8303.5 1675.2
## + COMPRI 1 98.0 8534.5 1693.3
## + PERSINC 1 27.1 8605.4 1698.8
## <none> 8632.5 1698.9
## + PROD 1 3.9 8628.6 1700.6
##
## Step: AIC=1159.26
## INTRATE ~ INFL
##
## Df Sum of Sq RSS AIC
## + PERSINC 1 456.15 3343.4 1076.8
## + PCE 1 383.68 3415.9 1091.0
## + UNEMPL 1 281.56 3518.0 1110.4
## + PROD 1 129.22 3670.3 1138.4
## + COMPRI 1 16.08 3783.5 1158.5
## <none> 3799.5 1159.3
## + HOUST 1 4.95 3794.6 1160.4
##
## Step: AIC=1076.85
## INTRATE ~ INFL + PERSINC
##
## Df Sum of Sq RSS AIC
## + PCE 1 58.683 3284.7 1067.2
## + HOUST 1 38.112 3305.3 1071.3
## + UNEMPL 1 35.684 3307.7 1071.8
## + COMPRI 1 25.996 3317.4 1073.7
## <none> 3343.4 1076.8
## + PROD 1 0.494 3342.9 1078.8
##
## Step: AIC=1067.16
## INTRATE ~ INFL + PERSINC + PCE
##
## Df Sum of Sq RSS AIC
## + HOUST 1 116.254 3168.4 1045.4
## + COMPRI 1 49.991 3234.7 1059.0
## + PROD 1 11.147 3273.5 1066.9
## <none> 3284.7 1067.2
## + UNEMPL 1 7.229 3277.5 1067.7
##
## Step: AIC=1045.38
## INTRATE ~ INFL + PERSINC + PCE + HOUST
##
## Df Sum of Sq RSS AIC
## + COMPRI 1 38.647 3129.8 1039.3
## + PROD 1 17.416 3151.0 1043.7
## <none> 3168.4 1045.4
## + UNEMPL 1 0.210 3168.2 1047.3
##
## Step: AIC=1039.28
## INTRATE ~ INFL + PERSINC + PCE + HOUST + COMPRI
##
## Df Sum of Sq RSS AIC
## <none> 3129.8 1039.3
## + PROD 1 4.4425 3125.3 1040.3
## + UNEMPL 1 0.0077 3129.8 1041.3
```

#### Stepwise Forward Regression using AIC

As we can see, in both the cases (a) and (b) we obtain the same final model.

```
## Call:
## lm(formula = INTRATE ~ INFL + PERSINC + PCE + HOUST + COMPRI,
## data = df)
##
## Coefficients:
## (Intercept)      INFL      PERSINC      PCE      HOUST
## -0.236287      0.717720      0.240157      0.339822     -0.020519
## COMPRI
## -0.007499
```

#### Final Model (using AIC)

As can be seen from above, the **unrestricted model** is  $INTRATE = X_1\beta_1 + X_2\beta_2 + \epsilon$  and

the **restricted model** is  $INTRATE = X_1\beta_1 + \epsilon$ ,

where  $X_1 = X(INFL, COMPRI, PCE, PERSINC, HOUST)$  and  $X_2 = X(PROD, UNEMPL)$ .

Sum of square residual for the unrestricted model = 3119.796, whereas the Sum of square residual for the restricted model = 3129.795.

Now the  $F$ -statistic for our hypothesis test  $H_0 : \beta_2 = 0$  against  $H_1 : \beta_2 \neq 0$  is  $F = \frac{(R_1^2 - R_0^2)/g}{(1 - R_1^2)/(n - k)} = \frac{(0.6386 - 0.63744)/2}{(1 - 0.6386)/(660 - 8)} = 1.046375$ , and the  $p$ -value is 0.3517967 with degrees of freedoms 2, 653. Similar results we obtain from **wald test** and **anova**, so we can't reject our **null hypothesis** at 5% significance level, it confirms that we shall be better off without the extra variables (with the **restricted model**).

(c) Compare your model from (a) and the Taylor rule of equation  $i_t = \beta_1 + \beta_2 \pi_t + \beta_3 y_t + \epsilon_t$ . Consider  $R^2$ , AIC and BIC. Which of the models do you prefer?

- We compare both the unrestricted (with all 7 explanatory variables) and the restricted (final) model (with 5 explanatory variables) obtained from part (a) with Taylor model (with 2 explanatory variables). The results are as shown below.

	Taylor's (restricted)	unrestricted	restricted
$R^2$	0.57483	0.6386	0.63744
AIC	3013.42062	2916.16577	2914.2777
BIC	3031.38958	2956.59593	2945.72338

#### Taylor's (restricted) model

INTRATE ~ INFL + PROD

Residuals:

Min	1Q	Median	3Q	Max
-5.1842	-1.6619	0.0085	1.3745	7.9237

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	1.25075	0.17609	7.103	3.19e-12 ***
INFL	0.97441	0.03271	29.792	< 2e-16 ***
PROD	0.09475	0.01970	4.810	1.88e-06 ***

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 2.364 on 657 degrees of freedom  
Multiple R-squared: 0.5748, Adjusted R-squared: 0.5735  
F-statistic: 444.1 on 2 and 657 DF, p-value: < 2.2e-16

#### restricted model from (a)

INTRATE ~ INFL + COMMPRI + PCE + PERSINC + HOUST

Residuals:

Min	1Q	Median	3Q	Max
-7.1918	-1.5298	-0.0974	1.3829	7.6603

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	-0.236287	0.230212	-1.026	0.30509
INFL	0.717720	0.058972	12.598	< 2e-16 ***
COMMPRI	-0.007499	0.002639	-2.842	0.00463 **
PCE	0.339822	0.058989	5.761	1.29e-08 ***
PERSINC	0.240157	0.059265	4.052	5.68e-05 ***
HOUST	-0.020519	0.004382	-4.682	3.45e-06 ***

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 2.188 on 654 degrees of freedom  
(1 observation deleted due to missingness)  
Multiple R-squared: 0.6374, Adjusted R-squared: 0.6347  
F-statistic: 230 on 5 and 654 DF, p-value: < 2.2e-16

#### unrestricted model from (a)

INTRATE ~ INFL + PROD + UNEMPL + COMMPRI + PCE + PERSINC + HOUST

Residuals:

Min	1Q	Median	3Q	Max
-7.4286	-1.4409	-0.1142	1.3547	7.7212

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	-0.214701	0.244829	-0.877	0.3808
INFL	0.696766	0.062031	11.233	< 2e-16 ***
PROD	-0.057581	0.039848	-1.445	0.1489
UNEMPL	0.104141	0.096640	1.078	0.2816
COMMPRI	-0.005525	0.002973	-1.858	0.0636 .
PCE	0.342641	0.069230	4.949	9.49e-07 ***
PERSINC	0.246418	0.060525	4.071	5.25e-05 ***
HOUST	-0.019371	0.004660	-4.157	3.65e-05 ***

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 2.187 on 652 degrees of freedom  
Multiple R-squared: 0.6386, Adjusted R-squared: 0.6347  
F-statistic: 164.6 on 7 and 652 DF, p-value: < 2.2e-16

Now the  $F$ -statistic for our hypothesis test to test the Taylor's (restricted) model against the unrestricted model, we found  $F = 23.01$ , and the  $p$ -value is < 0.05. Similar results we obtain from wald test and anova, so we can reject our null hypothesis at 5% significance level, it confirms that Taylor's model is not better than the unrestricted model.

```
## Wald test
##
## Model 1: INTRATE ~ INFL + PROD + UNEMPL + COMMPRI + PCE + PERSINC + HOUST
## Model 2: INTRATE ~ INFL + PROD
##   Res.Df Df    F      Pr(>F)
## 1      652
## 2      657 -5 23.01 < 2.2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
## Analysis of Variance Table
##
## Model 1: INTRATE ~ INFL + PROD + UNEMPL + COMMPRI + PCE + PERSINC + HOUST
## Model 2: INTRATE ~ INFL + PROD
##   Res.Df    RSS Df Sum of Sq    F      Pr(>F)
## 1      652 3119.8
## 2      657 3670.3 -5    -550.51 23.01 < 2.2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Similarly with anova test we can see that the  $p$ -value is small, so we can reject the null hypothesis again and conclude that Taylor's model is not better than the restricted model (we can't do walds test because these models are not nested).

```
## Analysis of Variance Table
##
## Model 1: INTRATE ~ INFL + COMMPRI + PCE + PERSINC + HOUST
## Model 2: INTRATE ~ INFL + PROD
##   Res.Df    RSS Df Sum of Sq    F      Pr(>F)
## 1      654 3129.8
## 2      657 3670.3 -3    -540.51 37.648 < 2.2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

(d) Test the Taylor rule of equation  $i_t = \beta_1 + \beta_2 \pi_t + \beta_3 y_t + \epsilon$  using the RESET test, Chow break and forecast test (with in both tests as break date January 1985) and a Jarque-Bera test. What do you conclude?

- Sum of square residual for the unrestricted model =  $S1 = 3119.796$ , whereas the Sum of square residual for the restricted Taylor's model =  $S0 = 3670.306$  and the F-statistic for the RESET Test  $F = \frac{(S_0 - S_1)/g}{S_1/(n-k)} = \frac{(3119.796 - 3670.306)/1}{3119.796/(660-4)} = 115.7558$ , with  $p\text{-value}=0$ , we can reject the null hypothesis at 5% significance level, concluding that **we reject that the model is a linear regression model.**

$S_0$

- The total Sum of square residual for the Taylor's model for 1960:1 - 2014:12 is =  $3670.306$ , whereas the Sum of square residual before 1985 January (1960:1 - 1984:12) =  $S_1 = 1757.802$ , whereas the Sum of square residual for on or after 1985 January ((1985:1 - 2014:12)) =  $S_2 = 1813.22$ . Hence, the F-statistic for the Chow break Test  $F = \frac{(S_0 - (S_1 + S_2))/k}{(S_1 + S_2)/(n-2+k)} = 4.578194$ , with  $p\text{-value}=0.003568257$ , **we can reject the null hypothesis** at 5% significance level **that the model parameters do not suffer from the break**
- The F-statistic for the Chow forecast Test  $F = \frac{(S_0 - S_1)/n_2}{S_1/(n_1-k)} = 0.8976073$ , with  $p\text{-value}$  as 0.4422323, where we have  $n_1 = 300$ ,  $n_2 = 360$ . The  $p\text{-value}$  is not significant at 5% level, hence **we can't reject the null hypothesis that there is no structural change in the prediction period.**
- The Jarque-Bera test statistic is  $JB = \sqrt{\frac{n}{6}} S^2 + \sqrt{\frac{n}{24}} (K - 3)^2 = 9.171319$ , where skewness of residuals  $S = 0.3213945$  and kurtosis  $K = 3.164455$ , with  $p\text{-value}$  0.01019702 so that we can **reject the null hypothesis  $H_0$  that residuals are normally distributed** at 5% level of significance.

**Conclusion:** we can conclude that the model requires non-linear terms of the regressors and linear regression is not appropriate here (also rejecting the hypothesis that the error is normally distributed). The chow break test indicates that the model parameters suffer from the break, although surprisingly enough, chow forecast test indicates that there is no structural change during the prediction period.