d

> summary(iv_inv)

- **d.** Obtain the 2SLS estimates of the investment equation using all eight exogenous and predetermined variables as IVs and software designed for 2SLS. Compare the estimates to the OLS estimates in part (a). Do you find any important differences?
- e. Estimate the second-stage model $I_t = \beta_1 + \beta_2 \hat{P}_t + \beta_3 P_{t-1} + \beta_4 K_{t-1} + e_{2t}$ by OLS. Compare the estimates and standard errors from this estimation to those in part (d). What differences are there?

```
call:
ivreg(formula = i \sim p + plag + klag | g + w2 + tx + time + plag +
   klag + elag, data = df)
Residuals:
          1Q Median
   Min
                       3Q
-3.2909 -0.8069 0.1423 0.8601 1.7956
Coefficients:
          Estimate Std. Error t value Pr(>|t|)
(Intercept) 20.27821
                          2.419 0.02707 *
                  8.38325
          0.15022
                   0.19253 0.780 0.44598
plag
                   0.18093 3.404 0.00338 **
          0.61594
klag
                   0.04015 -3.930 0.00108 **
Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' '1
Residual standard error: 1.307 on 17 degrees of freedom
                        Adjusted R-squared: 0.8646
Multiple R-Squared: 0.8849,
Wald test: 41.2 on 3 and 17 DF, p-value: 5.148e-08
> print(compare_slopes, n = Inf)
# A tibble: 8 x 6
  model term
                  estimate std.error statistic p.value
                     <db7>
                              <db7>
                                        <db7>
                                                <db7>
  <chr> <chr>
      (Intercept)
                    10.1
                              5.47
                                        1.85 0.0814
2 OLS
                     0.480
                             0.0971
                                       4.94 0.000125
                     0.333
                                       3.30 0.00421
3 OLS
       plag
                             0.101
4 OLS
       klag
                    -0.112
                             0.0267
                                       -4.18
                                             0.000624
5 2SLS (Intercept)
                    20.3
                             8.38
                                       2.42
                                             0.0271
6 2SLS
                     0.150
                             0.193
                                       0.780 0.446
       plag
7 2SLS
                                       3.40
                     0.616
                             0.181
                                             0.00338
OLS estimate result have smultaneity bias
For example, when investment and profits
                    -0.158
                                       -3.93
are both high in certain years, ols may overestimate the relationship between
 the two. The 2565 method addresses
the endogeneity but may sacrifice
```

```
> summary(stage2)
    Call:
     lm(formula = i \sim phat + plag + klag, data = df)
     Residuals:
             1Q Median 3Q
       Min
     -3.8778 -1.0029 0.3058 0.7275 2.1831
    Coefficients:
             Estimate Std. Error t value Pr(>|t|)
     (Intercept) 20.27821 9.97663 2.033 0.05802 .
                            0.656 0.52084
           0.15022
0.61594
                      0.22913
                             2.861 0.01083 *
     plag
                      0.21531
     klag
             -0.15779
                     0.04778 -3.302 0.00421 **
    Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
    Residual standard error: 1.556 on 17 degrees of freedom
     Multiple R-squared: 0.837, Adjusted R-squared: 0.8082
     F-statistic: 29.09 on 3 and 17 DF, p-value: 6.393e-07
All the signs of the slopes
remain unchanged
However, the standard errors
have changed
But this might reduce
the precision of
 estimates.
```

Made with Goodnotes