Q 28 a

Qi = Qi + Qz Pi + Q3 PSi + Q4 DIi + Pdi

Qi = B1 + B2 Pi + B3 PFi + Csi

az Pi = Qi - Q1 - Q3 PSi - Q4 DIi - edi 索求

供给 BzPi= Qi-Bi-BaPFi-Psi

=) 索求
$$P_i = \frac{Q_i}{\alpha_2} - \frac{\alpha_1}{\alpha_2} - \frac{\alpha_3 P_5}{\alpha_2} - \frac{\alpha_4 DI_i}{\alpha_2} - \frac{e_{di}}{\alpha_2}$$

供給
$$P_i = \frac{Q_i}{B_2} - \frac{B_1}{B_2} - \frac{B_3PF_i}{B_2} - \frac{esi}{B_2}$$

需求
$$\frac{1}{\alpha_2} < 0$$
 $-\frac{\alpha_3}{\alpha_2} > 0$ $\frac{-\alpha_4}{\alpha_2} > 0$

供給
$$\frac{1}{\beta_2} > 0 - \frac{\beta_3}{\beta_2} > 0$$

(b) 索求

Min 1Q Median 3Q Max -39.661 -6.781 2.410 8.320 20.251

Coefficients:

Estimate Std. Error t value Pr(>|t|) (Intercept) -11.428 q -2.671 13.592 -0.841 0.40810 1.175 -2.273 0.03154 * 1.116 3.103 0.00458 ** ps di 1.116 3.103 0.00458 ** 2.747 4.875 4.68e-05 ***

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

Residual standard error: 13.17 on 26 degrees of freedom Multiple R-Squared: 0.5567, Adjusted R-Squared: 0.5056 Wald test: 17.37 on 3 and 26 DF, p-value: 2.137e-06

係數符號皆正確

供给

Min 1Q Median 3Q Max -9.7983 -2.3440 -0.6281 2.4350 11.1600

| Stimate Std. Error t value $\Pr(>|t|)$ | (Intercept) -58.7982 | 5.8592 -10.04 1.32e-10 | *** | q | 2.9367 | 0.2158 | 13.61 1.32e-13 | *** | pf | 2.9385 | 0.1560 | 18.97 | < 2e-16 | *** |

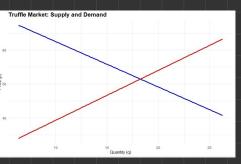
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 4.399 on 27 degrees of freedom Multiple R-Squared: 0.9486, Adjusted R-squared: 0.9448 Wald test: 232.7 on 2 and 27 DF, p-value: < 2.2e-16

```
> # 計算揮性
> ed <- (1 / delta_2) * (mean_p / mean_q)
> cat("需求的平均點價格彈性為:", round(ed, 3), "\n")
需求的平均點價格彈性為: -1.272
```

(d)

(C)



(G)

```
> cat("均衡數量 q* =", round(q_eq, 3), "\n")
均衡數量 q* = 18.25
> cat("均衡價格 p* =", round(p_eq, 3), "\n")
均衡價格 p* = 62.843
```

```
> p_hat <- c + b_ps * ps_star + b_di * di_star + b_pf * pf_star
> cat("Reduced-form 預測價格 P* =", round(p_hat, 3), "\n")
Reduced-form 預測價格 P* = 62.814
>
```

> # 預測 Q > q_hat <- c_q + bq_ps * ps_star + bq_di * di_star + bq_pf * pf_star > cat("Reduced-form 預測數量 Q* =", round(q_hat, 3), "\n") Reduced-form 預測數量 Q* = 18.26

```
(<del>5</del>)
```

Ħ .	A LIDDIE									
	model		term	estimate	std.error	statistic	p. value			
	<chr></chr>		<chr></chr>	<db7></db7>	<db7></db7>	<db7></db7>	<db7></db7>			
			(Intercept)							
2	Demand	25L5	q	-2.67	1.17	-2.27	3.15e- 2			
3	Demand	2SLS	ps	3.46	1.12	3.10	4.58e- 3			
4	Demand	25LS	di	13.4	2.75	4.87	4.68e- 5			
5	Demand	OLS	(Intercept)	-13.6	9.09	-1.50	1.46e- 1			
6	Demand	OLS	q	0.151	0.499	0.303	7.64e- 1			
7	Demand	OLS	ps	1.36	0.594	2.29	3.03e- 2			
8	Demand	OLS	di	12.4	1.83	6.77	3.48e- 7			
9	Supply	2SLS	(Intercept)	-58.8	5.86	-10.0	1.32e-10			
10	Supply	25L5	q	2.94	0.216	13.6	1.32e-13			
11	Supply	25L5	pf	2.96	0.156	19.0	3.88e-17			
12	Supply	OLS	(Intercept)	-52.9	5.02	-10.5	4.68e-11			
13	Supply	OLS	q	2.66	0.171	15.5	5.42e-15			
14	Supply	OLS	pf	2.92	0.148	19.7	1.47e-17			
>										
>	# 如需寫	λ csv	1:							
> .	# write.	.csv(all_results,	"regress"	ion_compar	ison_result	ts.csv", r	ow.names	= FALSE	Ξ)
-	7.00				and the same of th			CANADA STATE OF THE STATE OF TH	A 100 A	

OLS 的需求函數 自的符號錯誤且不顯著

其餘的符號正確且顯著

```
CH 11-30
```

Residuals:

(a)

Min 1Q Median 3Q

-2.56562 -0.63169 0.03687 0.41542 1.49226

coefficients:

Estimate Std. Error t value Pr(>|t|) (Intercept) 10.12579 5.46555 1.853 0.081374

4.939 0.000125 *** 0.47964 0.09711 3.302 0.004212 ** plag 0.33304 0.10086

-0.11179 0.02673 -4.183 0.000624 *** klag

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

a. 當期利潤 (p) 和一期前利潤 (plag) 都具有正向係數:當企業賺取較高利潤並擁有充足的內部資金時,會 增加其投資支出,這與預期中「利潤越高、投資越多」的正向利潤-投資關係一致。相對地,對於一期前資本 存量(klaq),其係數為負:即現有資本基礎越大,企業對新增資本的邊際需求越低,這與加速器模型 (accelerator model)的預測一致。

三個主要斜率係數在 1% 至 5% 顯著水準下皆與零有統計上顯著差異。

(b)

Residuals:

1Q Median 30 мах -3.9067 -1.3050 0.3226 1.3613 2.8881

coefficients:

1

2

Estimate Std. Error t value Pr(>|t|) (Intercept) 50.38442 31.63026 1.593 0.43902 0.39114 1.122 0.2820 g

w2 -0.07961 2.53382 -0.031 0.9754 tx -0.923100.43376 -2.128 0.0530 0.80250 0.51886 1.547 0.1459

plag klag -0.21610 0.11911 -1.814 0.0928 . time 0.31941 0.77813 0.410 0.6881 0.02200 0.28216 0.078 elag 0.9390

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

Res. Df RSS Df Sum of Sq F Pr(>F) 18 108.04

13 61.95 46.093 1.9345 0.1566

不具有共同的然充計器直著性

12.4 13.255556 -0.8555556 16.9 16.577368 0.3226319 18.4 19.282347 -0.8823465 19.4 20.960143 -1.5601433 20.1 19.766509 0.3334910 19.6 18.238731 1.3612688 19.8 17.573065 2.2269354 21.1 19.541720 1.5582796 10 21.7 20.375101 1.3248995 11 15.6 17.180415 -1.5804148 12 11.4 12.705026 -1.3050261

7.0 8.999780 -1.9997802 14 11.2 9.054102 2.1458976 15 12.3 12.671263 -0.3712632 16 14.0 14.421338 -0.4213385

17 17.6 14.711907 2.8880932 18 17.3 19.796405 -2.4964049

19 15.3 19.206691 -3.9066913 20 19.0 17.419605 1.5803947

21 21.1 20.305654 0.7943462 22 23 5 22 657273 0 8427268

```
(C)
```

```
Residuals:
Min 1Q Median 3Q Max
-1.04645 -0.56030 0.06189 0.25348 1.36700
```

Coefficients:

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

Residual standard error: 0.7331 on 16 degrees of freedom Multiple R-squared: 0.9659, Adjusted R-squared: 0.9574 F-statistic: 113.4 on 4 and 16 DF, p-value: 1.588e-11

√ < 0.001 點 3 , P是内生的

(g)

Residuals: Min 1Q Median 3Q Max -3.2909 -0.8069 0.1423 0.8601 1.7956

Coefficients:

| Estimate Std. Error t value Pr(>|t|) | (Intercept) 20.77821 | 8.38325 | 2.419 | 0.02707 * p | 0.15022 | 0.19253 | 0.780 | 0.44598 | plag | 0.61594 | 0.18093 | 3.404 | 0.0338 *** klag | -0.15779 | 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 Multiple R-Squared: 0.8849, Adjusted R-squared: 0.8646 Wald test: 41.2 on 3 and 17 DF, p-value: 5.148e-08

OLS 回饋的結果顯示,當期利潤 P_t的係數為 0.48,在 1% 顯著水準下是顯著的。但這可能反映了同時性偏誤 (simultaneity bias):高投資年份通常也會有高利潤,因此會設估 P_{_t}t 的影響力,導致斜率被高估。

• 2SLS 回歸使用工具變數法處理內生性,結果發現 P_i 的係數降為 0.15,且變得不顯著,顯示 OLS 所估計 的效果是偏誤的 •

同時,**plag (落後利潤) 與 klag (落後資本) **的條數在 2SLS 下仍顯著,甚至效果更強。

term estimate std.error statistic p.value model <db1> <db1> <db7> <db1> <chr> (Intercept) 10.1 5.47 1.85 0.0814 0.480 0.0971 4.94 0.000<u>125</u> OLS plag 0.333 3.30 0.004<u>21</u> OLS klag -0.112 0.0267 -4.18 0.000<u>624</u> OLS (Intercept) 9.82 5.49 1.79 0.0913 2SLS 0.489 0.0984 4.97 0.000<u>116</u> 25LS plag 0.325 0.102 3.19 0.005<u>38</u> 2SLS klag -0.110 0.0268 -4.12 0.000721 2SLS

(Intercept) 20.27821 9.97663 2.033 0.05802 .
phat 0.15022 0.22913 0.656 0.52084
plag 0.61594 0.21531 2.861 0.01083 *
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

第二階段 OLS 回歸 (使用 \hat{P}_t ,也就是第一階段的預測值) 得出的條數估計,與 ivreg 所執行的 2SLS 完全相同。

但是,在手動執行的兩階段 OLS(先回歸得到 \hat{P}_t ,再用它進行 OLS)中,標準誤($standard\ errors$)明顯更大,因此得到的 $p ext{-}value\ to 更大。$

這種差異是因為:

手動的第二階段 OLS 忽略了 $\hat{P_t}$ 的預測誤差 (也就是忘了它是估計出來的,不是固定的數) ,

而 ivreg 則正確地將這層不確定性納入計算,回傳的是 I正確的 I2SLS 標準誤。

(f)

(e)

> # 輸出結果

> TR2

[1] 0.7067662

> crit_val

[1] 9.487729

0-7068 < 9.4817

不拒絕 HO ,工具變數和結構方程式的設差項

不相關