



```

name: <unnamed>
log: A:\_maestria_unibo_(operacional)\4_econometrics_1\4_problem_sets\2_ps2\3_
> log\log.smcl
log type: smcl
opened on: 27 Oct 2021, 19:59:47

1 .
2 . /**#####
3 . /** A.
4 . /**#####
5 .
6 . use 1_data\dataset_1.dta

7 .
8 . /**=====
9 . /** A.1.
10. /**=====
11. /**----- Calculate the mean value for each TV over all sample.
12.
13.
14. /**----- Calculate mean values of each target variable (TV).
15.
16. bysort state_code: egen mean_etr = mean(expos_to_robots)

17. bysort state_code: egen mean_emp = mean(d_emppriv_1990_2011)

18.
19. /**----- Remove duplicated data to display min and max values.
20.
21. quietly by state_code: gen dup = cond(_N==1,0,_n)

22. drop if dup > 1
    (674 observations deleted)

23.
24. /**----- Identify the states that have the min and max values for each TV.
25.
26. sort mean_etr

27. gen min_etr_state = 1 if _n == 1
    (47 missing values generated)

28. gen max_etr_state = 1 if _n == _N
    (47 missing values generated)

29. table state_code (min_etr_state), statistic (mean mean_etr) nototals

```

	min_etr_state 1
state_1 NV	<b>.5719783</b>

```

30. table state_code (max_etr_state), statistic (mean mean_etr) nototals

```

	max_etr_state 1
state_1 MI	<b>5.246422</b>

```

31.
32. sort mean_emp

33. gen min_emp_state = 1 if _n == 1
   (47 missing values generated)

34. gen max_emp_state = 1 if _n == _N
   (47 missing values generated)

35. table state_code (min_emp_state), statistic (mean mean_emp) nototals

```

	min_emp_state 1
state_1 NC	<b>-4.594934</b>

```

36. table state_code (max_emp_state), statistic (mean mean_emp) nototals

```

	max_emp_state 1
state_1 ND	<b>7.900403</b>

```

37.
38. /*----- Calculate the mean value for each TV over state means.
39.
40. summarize mean_etr

```

Variable	Obs	Mean	Std. dev.	Min	Max
mean_etr	<b>48</b>	<b>1.852853</b>	<b>.9572189</b>	<b>.5719783</b>	<b>5.246422</b>

```

41. summarize mean_emp

```

Variable	Obs	Mean	Std. dev.	Min	Max
mean_emp	<b>48</b>	<b>.3809222</b>	<b>2.482054</b>	<b>-4.594934</b>	<b>7.900403</b>

```

42.
43. /*----- Two-way catterplot with regression line.
44.
45. clear all

46. use 1_data\dataset_1.dta

47.
48. graph twoway (lfitci d_emppriv_1990_2011 expos_to_robots, xtitle("Exposure to robots
> ")) (scatter d_emppriv_1990_2011 expos_to_robots, msize(tiny) legend(size(small) lab
> el(3 "Change in the share of private employment over total population. 1990-2011.))
> )

49.
50. graph save "scatter_line.gph", replace
   file scatter_line.gph saved

```

```

51. graph save 4_graphs\scatter_line.gph, replace
    file 4_graphs\scatter_line.gph saved

52. graph export 4_graphs\scatter_line.png, as(png) replace
    file 4_graphs\scatter_line.png saved as PNG format

53. graph close

54.
55. /**=====
56. /** B.
57. /**=====
58.
59. clear all

60. use 1_data\dataset_1.dta

```

```

61.
62. /**=====
63. /** B.2. Simple linear regression.
64. /**=====
65.
66. reg d_emppriv_1990_2011 expos_to_robots, rob

```

```

Linear regression              Number of obs   =      722
                              F(1, 720)         =      99.20
                              Prob > F           =      0.0000
                              R-squared          =      0.1421
                              Root MSE       =      2.955

```

d_emppriv_~2011	Coefficient	Robust std. err.	t	P> t	[95% conf. interval]	
expos_to_robots	<b>-1.09071</b>	<b>.1095083</b>	<b>-9.96</b>	<b>0.000</b>	<b>-1.305704</b>	<b>-.8757165</b>
_cons	<b>2.774654</b>	<b>.2344016</b>	<b>11.84</b>	<b>0.000</b>	<b>2.314461</b>	<b>3.234846</b>

```

67. outreg2 using 5_tables\reg1, tex replace
    5_tables\reg1.tex
    dir : seeout

```

```

68.
69. /**=====
70. /** B.3. Regression with additional controls.
71. /**=====
72.
73. reg d_emppriv_1990_2011 expos to_robots ipums logpop_1990 ipums_female_1990 ipums_ab
> ove65_1990 ipums_highschool_1990 ipums_somcolle_1990 ipums_college_1990 ipums_mas
> ters_1990 ipums_white_1990 ipums_black_1990 ipums_hispanic_1990 ipums_asian_1990, ro
> b
note: ipums_college_1990 omitted because of collinearity.
note: ipums_hispanic_1990 omitted because of collinearity.

```

```

Linear regression              Number of obs   =      722
                              F(10, 711)        =      46.76
                              Prob > F           =      0.0000
                              R-squared          =      0.3847
                              Root MSE       =      2.5183

```

d_emppriv_1990_2011	Coefficient	Robust std. err.	t	P> t	[95% conf. interval]	
expos_to_robots	<b>-.555736</b>	<b>.0887673</b>	<b>-6.26</b>	<b>0.000</b>	<b>-.7300135</b>	<b>-.38145</b>
ipums_logpop_1990	<b>-.4474018</b>	<b>.0894417</b>	<b>-5.00</b>	<b>0.000</b>	<b>-.6230033</b>	<b>-.27180</b>
ipums_female_1990	<b>.6339813</b>	<b>15.73012</b>	<b>0.04</b>	<b>0.968</b>	<b>-30.24906</b>	<b>31.517</b>

```

> 02
ipums_above65_1990 | 11.57593 5.439096 2.13 0.034 .8973153 22.254
> 54
ipums_highschool_1990 | -23.29328 6.426127 -3.62 0.000 -35.90973 -10.676
> 83
ipums_somcollege_1990 | -12.70352 8.72179 -1.46 0.146 -29.82706 4.4200
> 27
ipums_college_1990 | 0 (omitted)
ipums_masters_1990 | -76.55848 15.78116 -4.85 0.000 -107.5417 -45.575
> 23
ipums_white_1990 | -1.879022 .6873117 -2.73 0.006 -3.228426 -.52961
> 92
ipums_black_1990 | -7.003238 .9646186 -7.26 0.000 -8.897079 -5.1093
> 96
ipums_hispanic_1990 | 0 (omitted)
ipums_asian_1990 | -28.80399 8.598956 -3.35 0.001 -45.68638 -11.921
> 61
_cons | 27.46541 11.39099 2.41 0.016 5.101417 49.82
> 94

```

```

74. outreg2 using 5_tables\reg2, tex replace
5_tables\reg2.tex
dir : seeout

```

```

75.
76. /**----- Dropping some variables because of multicollinearity.
77.
78. reg d_emppriv_1990_2011 expos to robots ipums_logpop_1990 ipums_female_1990 ipums_ab
> ove65_1990 ipums_somcollege_1990 ipums_college_1990 ipums_masters_1990 ipums_black_
> 1990 ipums_hispanic_1990 ipums_asian_1990, rob

```

```

Linear regression
Number of obs      =      722
F(10, 711)         =      46.76
Prob > F            =      0.0000
R-squared           =      0.3847
Root MSE           =      2.5183

```

```

-----
d_emppriv_1990_2011 | Coefficient   Robust      t    P>|t|    [95% conf. interval
> 1]
-----
expos_to_robots | -.555736 .0887673 -6.26 0.000 -.7300135 -.38145
> 85
ipums_logpop_1990 | -.4474018 .0894417 -5.00 0.000 -.6230033 -.27180
> 03
ipums_female_1990 | .6339831 15.73012 0.04 0.968 -30.24906 31.517
> 02
ipums_above65_1990 | 11.57593 5.439096 2.13 0.034 .8973147 22.254
> 54
ipums_somcollege_1990 | 10.58976 3.28857 3.22 0.001 4.133291 17.046
> 23
ipums_college_1990 | 23.29329 6.426127 3.62 0.000 10.67683 35.909
> 74
ipums_masters_1990 | -53.26521 11.6925 -4.56 0.000 -76.22117 -30.309
> 24
ipums_black_1990 | -5.124215 1.046752 -4.90 0.000 -7.179309 -3.0691
> 22
ipums_hispanic_1990 | 1.879022 .6873117 2.73 0.006 .5296187 3.2284
> 25
ipums_asian_1990 | -26.92497 8.413793 -3.20 0.001 -43.44382 -10.406
> 12
_cons | 2.293104 7.817483 0.29 0.769 -13.05501 17.641
> 22

```

```
79. outreg2 using 5_tables\reg2, tex replace
   5_tables\reg2.tex
   dir : seeout
```

```
80.
81. /**=====
82. /** B.4. Regression with the square of expos_to_robots.
83. /**=====
84.
85. gen expos_2 = expos_to_robots*expos_to_robots

86. reg d_emppriv_1990_2011 expos_to_robots expos_2 ipums_logpop_1990 ipums_female_1990
   > ipums_above65_1990 ipums_somecollege_1990 ipums_college_1990 ipums_masters_1990 ipum
   > s_black_1990 ipums_hispanic_1990 ipums_asian_1990, rob
```

```
Linear regression                Number of obs    =      722
                                F(11, 710)         =     47.47
                                Prob > F            =     0.0000
                                R-squared           =     0.3934
                                Root MSE        =     2.5022
```

	Coefficient	Robust std. err.	t	P> t	[95% conf. interval]	
d_emppriv_1990_2011						
expos_to_robots	-1.32494	.2462026	-5.38	0.000	-1.808312	-.84156
expos_2	.1055034	.0272236	3.88	0.000	.0520549	.15895
ipums_logpop_1990	-.404733	.0928466	-4.36	0.000	-.5870197	-.22244
ipums_female_1990	4.906006	15.72564	0.31	0.755	-25.96831	35.780
ipums_above65_1990	10.56819	5.561721	1.90	0.058	-.3512001	21.487
ipums_somecollege_1990	8.748911	3.285391	2.66	0.008	2.298667	15.199
ipums_college_1990	20.84298	6.385557	3.26	0.001	8.306145	33.379
ipums_masters_1990	-50.07539	11.56498	-4.33	0.000	-72.78104	-27.369
ipums_black_1990	-5.610725	1.052319	-5.33	0.000	-7.676753	-3.5446
ipums_hispanic_1990	1.222744	.7225081	1.69	0.091	-.1957636	2.6412
ipums_asian_1990	-26.26183	8.474227	-3.10	0.002	-42.89937	-9.6242
_cons	1.340348	7.813079	0.17	0.864	-13.99915	16.679

```
87. outreg2 using 5_tables\reg3, tex replace
   5_tables\reg3.tex
   dir : seeout
```

```
88.
```

```

89. /**=====
90. /** B.5. Testing the marginal effect of expos_to_robots (and its quadratic term).
91. /**=====
92. /**----- Evaluating the marginal effect of exposure to robots at the mean of expos_t
> o_robots.
93.
94. summarize expos_to_robots

```

Variable	Obs	Mean	Std. dev.	Min	Max
expos_to_r~s	<b>722</b>	<b>1.800013</b>	<b>1.101769</b>	<b>.4188264</b>	<b>10.26309</b>

```
95. scalar mean_etr = r(mean)
```

```
96. di mean_etr
1.800013
```

```
97. lincom _b[expos_to_robots] + (2 * _b[expos_2] * mean_etr)
```

```
( 1) expos_to_robots + 3.600026*expos_2 = 0
```

d_emppr~2011	Coefficient	Std. err.	t	P> t	[95% conf. interval]
(1)	<b>-.9451252</b>	<b>.1541927</b>	<b>-6.13</b>	<b>0.000</b>	<b>-1.247853 - .6423971</b>

```

98.
99. /**=====
100 /** B.6. Heteroskedasticity tests.
101 /**----- Runnign full regression without rob.
102
103 reg d_emppriv_1990_2011 expos_to_robots expos_2 ipums_logpop_1990 ipums_female_1990
> ipums_above65_1990 ipums_somecollege_1990 ipums_college_1990 ipums_masters_1990 ipum
> s_black_1990 ipums_hispanic_1990 ipums_asian_1990

```

Source	SS	df	MS	Number of obs	=	<b>722</b>
Model	<b>2882.94304</b>	<b>11</b>	<b>262.085731</b>	F(11, 710)	=	<b>41.86</b>
Residual	<b>4445.29832</b>	<b>710</b>	<b>6.26098355</b>	Prob > F	=	<b>0.0000</b>
				R-squared	=	<b>0.3934</b>
				Adj R-squared	=	<b>0.3840</b>
Total	<b>7328.24136</b>	<b>721</b>	<b>10.1639963</b>	Root MSE	=	<b>2.5022</b>

d_emppriv_1990_2011	Coefficient	Std. err.	t	P> t	[95% conf. interval]
> 1]					
expos_to_robots	<b>-1.32494</b>	<b>.2619301</b>	<b>-5.06</b>	<b>0.000</b>	<b>-1.83919 - .81068</b>
> 98					
expos_2	<b>.1055034</b>	<b>.0330962</b>	<b>3.19</b>	<b>0.001</b>	<b>.0405253 .17048</b>
> 15					
ipums_logpop_1990	<b>-.404733</b>	<b>.0852326</b>	<b>-4.75</b>	<b>0.000</b>	<b>-.5720711 -.23739</b>
> 48					
ipums_female_1990	<b>4.906006</b>	<b>15.32995</b>	<b>0.32</b>	<b>0.749</b>	<b>-25.19145 35.003</b>
> 46					
ipums_above65_1990	<b>10.56819</b>	<b>4.401876</b>	<b>2.40</b>	<b>0.017</b>	<b>1.925936 19.210</b>
> 44					
ipums_somecollege_1990	<b>8.748911</b>	<b>3.116931</b>	<b>2.81</b>	<b>0.005</b>	<b>2.629407 14.868</b>
> 42					
ipums_college_1990	<b>20.84298</b>	<b>5.934436</b>	<b>3.51</b>	<b>0.000</b>	<b>9.191836 32.494</b>
> 12					
ipums_masters_1990	<b>-50.07539</b>	<b>12.29295</b>	<b>-4.07</b>	<b>0.000</b>	<b>-74.21027 -25.94</b>
> 05					
ipums_black_1990	<b>-5.610725</b>	<b>1.126267</b>	<b>-4.98</b>	<b>0.000</b>	<b>-7.821937 -3.3995</b>
> 13					
ipums_hispanic_1990	<b>1.222744</b>	<b>.9225356</b>	<b>1.33</b>	<b>0.185</b>	<b>-.5884797 3.0339</b>
> 68					
ipums_asian_1990	<b>-26.26183</b>	<b>10.24617</b>	<b>-2.56</b>	<b>0.011</b>	<b>-46.37824 -6.1454</b>
> 17					
_cons	<b>1.340348</b>	<b>7.462319</b>	<b>0.18</b>	<b>0.858</b>	<b>-13.3105 15.99</b>

&gt; 12

104 rvfplot, yline(0)

105 estat imtest, white

White's test

H0: Homoskedasticity

Ha: Unrestricted heteroskedasticity

chi2(76) = **168.18**Prob > chi2 = **0.0000**

Cameron &amp; Trivedi's decomposition of IM-test

Source	chi2	df	p
Heteroskedasticity	<b>168.18</b>	<b>76</b>	<b>0.0000</b>
Skewness	<b>17.13</b>	<b>11</b>	<b>0.1042</b>
Kurtosis	<b>9.52</b>	<b>1</b>	<b>0.0020</b>
Total	<b>194.83</b>	<b>88</b>	<b>0.0000</b>

106

107 /\*\*#####

108 /\*\* C.

109 /\*\*#####

110

111 clear all

112 use 1\_data\dataset\_2.dta

113 summarize

Variable	Obs	Mean	Std. dev.	Min	Max
czone	<b>23,770</b>	<b>21937.98</b>	<b>11352.08</b>	<b>100</b>	<b>39400</b>
d_yrwag~2011	<b>23,770</b>	<b>.0813208</b>	<b>.6081289</b>	<b>-4.947901</b>	<b>4.888123</b>
state_code	<b>23,770</b>	<b>24.8496</b>	<b>13.81532</b>	<b>1</b>	<b>48</b>
expos to r~s	<b>23,770</b>	<b>1.842783</b>	<b>1.157471</b>	<b>.4188264</b>	<b>10.26309</b>
education	<b>23,770</b>	<b>2.812873</b>	<b>1.318574</b>	<b>1</b>	<b>5</b>
race	<b>23,770</b>	<b>3.287042</b>	<b>1.424595</b>	<b>1</b>	<b>5</b>
female	<b>23,770</b>	<b>.4982751</b>	<b>.5000075</b>	<b>0</b>	<b>1</b>

114

115 /\*\*=====

116 /\*\* C.1. Creating a dummy variable for whites.

117 /\*\*=====

118

119 codebook race

**race****(unlabeled)**Type: Numeric (**long**)Label: **race\_num**Range: [**1,5**]Unique values: **5**Units: **1**Missing .: **0/23,770**

```

Tabulation: Freq.    Numeric    Label
            3,347        1    asian
            4,426        2    black
            5,274        3    hispanic
            3,503        4    other
            7,220        5    white

```

```

120 gen white = 0 if race != 5
    (7,220 missing values generated)

```

```

121 replace white = 1 if race == 5
    (7,220 real changes made)

```

```

122 tab race white

```

race	white		Total
	0	1	
asian	3,347	0	3,347
black	4,426	0	4,426
hispanic	5,274	0	5,274
other	3,503	0	3,503
white	0	7,220	7,220
Total	16,550	7,220	23,770

```

123
124 /**=====
125 /** C.2. Regression.
126 /**=====
127
128 reg d_yrwage_ln_1990_2011 c.expos_to_robots##i.female white##i.education, rob

```

```

Linear regression
                                Number of obs   =    23,770
                                F(12, 23757)     =    189.29
                                Prob > F         =    0.0000
                                R-squared        =    0.0368
                                Root MSE     =    .59698

```

		Coefficient	Robust std. err.	t	P> t	[95% conf. inter	
d_yrwage_ln_1990_2011							
> val]							
	expos_to_robots	-.0452585	.0044846	-10.09	0.000	-.0540486	-.036
> 4685							
	1.female	.1428625	.0140142	10.19	0.000	.1153937	.170
> 3312							
	female#c.expos_to_robots						
	1	.0078448	.0063929	1.23	0.220	-.0046856	.020
> 3752							
	1.white	-.0301436	.0146348	-2.06	0.039	-.0588287	-.001
> 4585							
	education						
	high school	-.0310826	.0167027	-1.86	0.063	-.063821	.001
> 6558							
	some college	-.0251477	.0171907	-1.46	0.144	-.0588425	.008
> 5471							
	college	.1015417	.0191859	5.29	0.000	.0639361	.139
> 1472							
	master	.1624391	.0202855	8.01	0.000	.1226782	.202
> 2001							
	white#education						
	1#high school	.0469742	.0180229	2.61	0.009	.011648	.082
> 3003							
	1#some college	.104628	.0184406	5.67	0.000	.0684833	.140



```

> 7727
      1#college |      .0593417   .0204027   2.91   0.004   .0193511   .099
> 3323
      1#master  |      .0211344   .0217293   0.97   0.331   -.0214564   .063
> 7251
      _cons    |      .0541556   .0160536   3.37   0.001   .0226895   .085
> 6218

```

```

129 outreg2 using 5_tables\reg4, tex replace
    5_tables\reg4.tex
    dir : seeout

```

```

130
131 /*=====
132 /* C.3. Chow test for testing differences between whites and non-whites.
133 /*=====
134
135 /* Not included.
136
137 /*=====
138 /* C.4 Test of the difference in marginal effect of robots on men and women.
139 /*=====
140
141 test 1.female#c.expos_to_robots

( 1)  1.female#c.expos_to_robots = 0

      F( 1, 23757) =      1.51
      Prob > F =      0.2198

```

```

142
143 /*=====
144 /* C.5. Regression with dummy variables for education levels.
145 /*=====
146
147 tabulate education, generate(edu)

```

education	Freq.	Percent	Cum.
less than high school	4,810	20.24	20.24
high school	5,643	23.74	43.98
some college	5,732	24.11	68.09
college	4,355	18.32	86.41
master	3,230	13.59	100.00
Total	23,770	100.00	

```

148 reg d_yrwage_ln_1990_2011 expos_to_robots female edu2 edu3 edu4 edu5

```

Source	SS	df	MS	Number of obs	=	23,770
Model	315.098903	6	52.5164838	F(6, 23763)	=	147.25
Residual	8475.1713	23,763	.356654097	Prob > F	=	0.0000
				R-squared	=	0.0358
				Adj R-squared	=	0.0356
Total	8790.2702	23,769	.369820783	Root MSE	=	.59721

d_yrwage_l~2011	Coefficient	Std. err.	t	P> t	[95% conf. interval]	
expos_to_robots	-.0414809	.0033469	-12.39	0.000	-.048041	-.0349208
female	.1572163	.0077475	20.29	0.000	.1420307	.1724019
edu2	-.0176851	.0117199	-1.51	0.131	-.0406569	.0052867
edu3	.0027235	.0116779	0.23	0.816	-.0201659	.0256129
edu4	.1202807	.0124924	9.63	0.000	.0957947	.1447667
edu5	.1674346	.0135864	12.32	0.000	.1408045	.1940648
_cons	.0381767	.0112288	3.40	0.001	.0161675	.0601859

```

149
150 /**=====
151 /** C.6. Regression with interaction between robots and dummies for education.
152 /**=====
153
154 reg d_yrwage_ln_1990_2011 c.expos_to_robots##i.education female race, rob

```

```

Linear regression              Number of obs   =    23,770
                              F(11, 23758)      =     89.82
                              Prob > F         =     0.0000
                              R-squared        =     0.0370
                              Root MSE     =     .59693

```

		Coefficient	Robust std. err.	t	P> t	[95% conf. in terval]
>	d_yrwage_ln_1990_2011					
>	expos_to_robots	-.0557862	.0083043	-6.72	0.000	-.0720631 -.
>	0395093					
>	education					
>	high school	-.0100199	.0223958	-0.45	0.655	-.0539171 .
>	0338772					
>	some college	-.0223304	.0226929	-0.98	0.325	-.0668099 .
>	0221491					
>	college	.0529709	.0248537	2.13	0.033	.0042561 .
>	1016858					
>	master	.0995622	.0233687	4.26	0.000	.0537581 .
>	1453663					
>	education#c.expos_to_robots					
>	high school	-.0037766	.0103959	-0.36	0.716	-.0241532
>	.0166					
>	some college	.0140973	.0102976	1.37	0.171	-.0060867 .
>	0342813					
>	college	.0367525	.0115503	3.18	0.001	.0141132 .
>	0593917					
>	master	.0363174	.0108219	3.36	0.001	.0151058 .
>	0575291					
>	female	.1573782	.0077434	20.32	0.000	.1422006 .
>	1725558					
>	race	.0040804	.0026976	1.51	0.130	-.0012071 .
>	0093679					
>	_cons	.0505455	.0214288	2.36	0.018	.0085436 .
>	0925473					

```

155 outreg2 using 5_tables\reg5, tex replace
    5_tables\reg5.tex
    dir : seeout

```

```

156
157 /**#####
158 /** n. Close log.
159 /**#####
160
161 log close
    name: <unnamed>
    log: A:\_maestria_unibo_(operacional)\4_econometrics_1\4_problem_sets\2_ps2\3_
> log\log.smcl
    log type: smcl
    closed on: 27 Oct 2021, 19:59:53

```