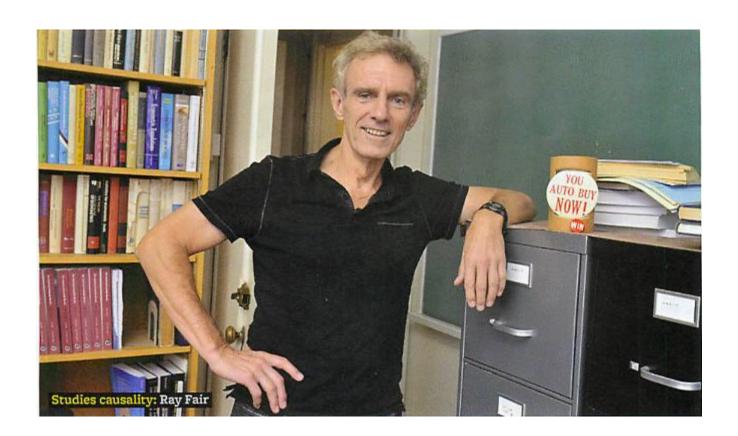
"A Theory of Extramarital Affairs"

Fair, 1978

Why this article?

Ray C. Fair is a macroeconomist at Yale; this article sticks out on his CV.

It's become a standard dataset in economics for illustrating the Tobit estimator.



Description of the Dataset

• The data comes from a survey conducted by the magazine *Psychology Today (PT)* in July, 1969. Responder bias?

• The outcome variable is how often the respondent engaged in extramarital affairs over the previous year:

```
0 = \text{none}, 1 = \text{once}, 2 = \text{twice}, 3 = 3 \text{ times}, 7 = 4-10 \text{ times}, 12 = \text{monthly}, 12 = \text{weekly}, 12 = \text{daily}
```

Note that it's a kind of weird count-ordinal hybrid (Fair's analysis is mostly qualitative anyway).

 The explanatory variables are: sex, age, how long the respondent has been married, how religious they are, their level of education, their occupation, and how they would subjectively rate their satisfaction with their marriage. Predictions?

Fair's Estimation Strategy

• "... because many values of y [...] are zero, it would clearly be incorrect to use ordinary least squares to estimate the equations. The obvious technique to use in this case is the Tobit estimator" (pg.57). There is also a little top-censoring: all frequencies equal to or greater than monthly are coded as 12. It seems Fair doesn't acknowledge this (see below).

• Recall Tobin's original problem: large numbers of households spent exactly \$0 on luxury goods; he wanted to distinguish the effect of the explanatory variables on people's choosing to buy *any* luxury goods (selection) from the effect of the explanatory variables on *how much* luxury goods someone buys, given that they buy any.

• In this case, 451 of the 601 respondents have a response variable of 0 (no extramarital affairs in the last year). The problem is structurally very similar to Tobin's: how do the explanatory variables affect a person's *selecting* a nonzero response variable, and how do they affect the *level* of the response variable, given that it's nonzero?

Fair's Results

	•
$\alpha = 0.1$	ı
u – 0. i	ı
** $\alpha = 0.05$	ļ
*** ~ - 0 01	ı
*** $\alpha = 0.01$	ı

Variable		Coeff. Est.	t-Stat.	Coeff. Est.	t-Stat.
Constant		7.60	1.92 **	8.17	2.96 **
Occupation	27	.213	.67	.326	1.29 *
Education	26	.0252	.11		
Marital happiness	₹g	-2.27	-5.48 ***	-2.28	-5.61 ***
Age	22	193	-2.37***	179	-2.26 **
No. years married	Z3	.533	3.63 ***	-554	4.13 ***
Children	Z4	1.02	.79		
Degree of religiosity	25	-1.70			-4.14 ***
Sex		.945	.88		
	$\dot{\sigma}_{PT}^2$	8.26	2.5.20	8.25	
No. observations	19790	601‡		601 t	85.50

There are two sets of coefficient estimates: if you look at the second column, you see that it drops some of the variables from the first model.

Replication in R

Replication in R

```
Estimate Std. Error z value Pr(>|z|)
(Intercept)
              7.60849
                         3.90599 1.948 0.051426
                                   0.890 0.373548
              0.94579
                         1.06287
sex
             -0.19270
                         0.08097 -2.380 0.017316
age
             0.53319
                         0.14661
years_married
                                   3.637 0.000276
children.
              1.01918
                         1.27957
                                   0.797 0.425741
how_religious -1.69900
                      0.40548 -4.190 2.79e-05
education
            0.02536
                         0.22767
                                   0.111 0.911304
occupation
                         0.32116
              0.21298
                                   0.663 0.507220
                         0.41541 -5.472 4.44e-08
rate_marriage -2.27328
```

Other than rounding, the point estimates are identical to Fair's (thankfully). Because the degrees of freedom is so high, R uses a normal approximation rather than a t-distribution: this has no effect on the statistical significance (R's star notation is a bit different from mine).

Fair seems not to have been aware of Heckman selection models (Heckman's first paper on the subject was about two years earlier). In fact, Heckman models are perhaps more useful for the purpose discussed earlier, since they have a distinct selection equation and outcome equation.

This allows us to move towards a *theory* of censoring / selection.

```
### Estimate a Heckman model.
library(sampleSelection)
# We have to define a binary outcome variable for the selection equation.
thing <- numeric(601)
for (i in 1:length(data$y)) {
 if (data$y[i] > 0) {
   thing[i] <- 1
 if (data$y[i] == 0) {
   thing[i] <- 0
# Then estimate the model.
model2 <- heckit(selection = thing ~ sex + age + years_married + children + how_religious +
                   education + occupation + rate_marriage,
                 outcome = y ~ sex + age + years_married + children + how_religious +
                   education + occupation + rate_marriage,
                 data = data)
summary(model2)
```

```
Estimate Std. Error t value Pr(>|t|)
            0.77940
                    0.51255 1.521 0.128895
(Intercept)
            0.17346
                   0.13799 1.257 0.209253
sex
           age
years_married 0.05434 0.01881 2.889 0.004005 **
children .
            0.21664
                   0.16517 1.312 0.190154
education
            0.01126 0.02952 0.382 0.702930
            0.01367 0.04140 0.330 0.741421
occupation
rate_marriage -0.27179
                    0.05347 -5.083 5.03e-07 ***
Outcome equation:
           Estimate Std. Error t value Pr(>|t|)
(Intercept)
          4.57504
                    11.15810 0.410
                                     0.682
                   4.55357 0.583
sex
           2.65669
                                   0.560
                   0.57499 -0.895
age
           -0.51448
                                   0.371
years_married 1.35905
                   1.21799 1.116
                                   0.265
children.
            3.49256
                   6.06633 0.576
                                   0.565
how_religious -4.42642
                    4.13525 -1.070
                                    0.285
education
            0.04795
                    0.63916 0.075
                                   0.940
occupation
            0.56871
                   0.87044 0.653
                                   0.514
rate_marriage -5.90339
                     5.63577 -1.047
                                     0.295
Multiple R-Squared:0.1876,
                          Adjusted R-Squared:0.1353
  Error terms:
           Estimate Std. Error t value Pr(>|t|)
            27.089 27.318
invMillsRatio
                             0.992
                                     0.322
sigma
             23.261
                         NA
                                NA
                                        NA
rho
              1.165
                         NA
                                NA
                                        NA
```

Takeaways:

- The same variables are estimated to have a significant effect on selection as in Fair's original analysis: age, years married, religiosity, and self-described marital satisfaction.
- As with the analysis of the Mroz data in the lecture: *none* of these variables is significant in the outcome equation. This suggests that these features influence *whether* to engage in an extramarital affair: not the *extent* of such behaviour conditional upon it being nonzero.
- The inverse Mills ratio (lamda) is not statistically significantly different from zero: we can't reject the null hypothesis that there is no selection bias effect.

The End.