

## The model

The generalized linear regression model can be stated as:

$$I_i = \beta_0 + \beta_1 x_{1i} + \beta_2 x_{2i} + \beta_3 x_{3i} + \beta_4 x_{4i} + u_i \quad (1)$$

and

$$\Pr(Y_i = y_i) = f(I_i)$$

where:  $I_i$  is index variables.

$y_i$  is counted number 0, 1, 2,...

$x_{ki}$  is independent variable  $k$ .

$f(\cdot)$  is either Poisson or Negative Binomial probability distribution function.

$u_i$  is disturbance term.

## Requirements (using Data set file – assign12.dta):

- 1 Estimate models for  $y_i$  assuming that the model is traditional linear regression model. Interpret your estimated result.

```
. use "C:\Users\Jillllin\OneDrive\Desktop\Thammasat\EE426\Data\assign12.dta"
```

```
. reg y x1 x2 x3 x4
```

Source	SS	df	MS	Number of obs	=	232
Model	44.7298499	4	11.1824625	F(4, 227)	=	5.96
Residual	425.748598	227	1.87554449	Prob > F	=	0.0001
				R-squared	=	0.0951
				Adj R-squared	=	0.0791
Total	470.478448	231	2.03670324	Root MSE	=	1.3695

y	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
x1	.1016201	.0435073	2.34	0.020	.0158904 .1873499
x2	.1345044	.0462142	2.91	0.004	.0434407 .225568
x3	-.0748194	.0480457	-1.56	0.121	-.1694919 .0198531
x4	.1684563	.0688243	2.45	0.015	.0328401 .3040725
_cons	.9568064	.107007	8.94	0.000	.7459523 1.16766

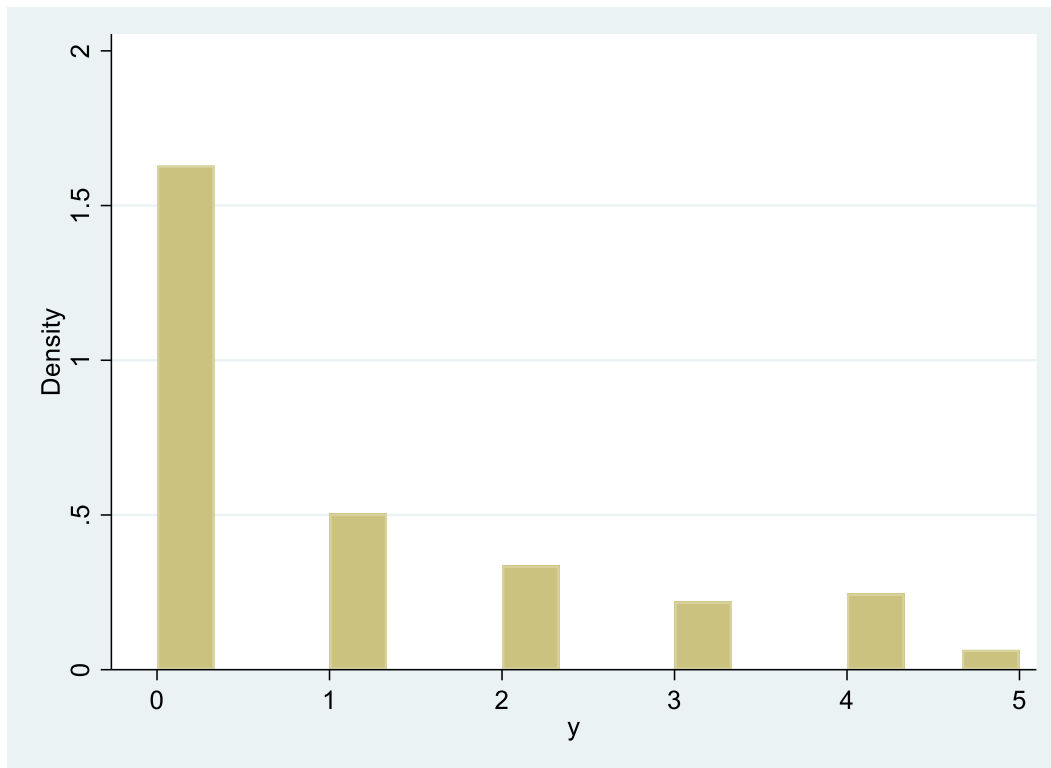
```
. est store ols
```

From estimated result, overall significance of both model and individual variables is significant except x3. However, the R-square is quite low. This means that the model has low explanatory power.

- 2 Create histogram for  $y_i$ . Determine whether there is limitation of dependent variable in this case. If yes, what type of limitation is it?

```
. histogram y
```

```
(bin=15, start=0, width=.33333333)
```



We can see that the distribution of  $y$  is highly skewed. Hence, normal OLS should not be applied here. Furthermore, the distribution looks like poisson distribution.

**3 Estimate models for  $y_i$  assuming that the probability functions follow Poisson probability distribution. Perform GOF test and determine whether Poisson is appropriated in this case. Interpret the estimated result (sign and meaning (in term of incidence-rate ratios), overall test, individual test, pseudo  $R^2$ , marginal effects).**

```
. poisson y x1 x2 x3 x4 ,nolog
```

```
Poisson regression                               Number of obs   =          232
                                                LR chi2(4)      =          43.33
                                                Prob > chi2     =          0.0000
Log likelihood = -342.88107                    Pseudo R2       =          0.0594
```

	y	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
	x1	.0971474	.0306762	3.17	0.002	.0370231 .1572717
	x2	.1293024	.0330916	3.91	0.000	.064444 .1941607
	x3	-.0715533	.0342177	-2.09	0.037	-.1386187 -.0044879
	x4	.1734482	.0507707	3.42	0.001	.0739395 .2729569
	_cons	-.1284876	.0849064	-1.51	0.130	-.294901 .0379259

```
. estat gof
```

```
Deviance goodness-of-fit = 409.4921
Prob > chi2(227)         = 0.0000

Pearson goodness-of-fit = 423.3541
Prob > chi2(227)         = 0.0000
```

```
. poisson y x1 x2 x3 x4 ,ir nolog
```

```
Poisson regression                               Number of obs   =      232
                                                LR chi2(4)      =      43.33
                                                Prob > chi2     =      0.0000
Log likelihood = -342.88107                    Pseudo R2      =      0.0594
```

y	IRR	Std. Err.	z	P> z	[95% Conf. Interval]	
x1	1.102023	.0338059	3.17	0.002	1.037717	1.170314
x2	1.138034	.0376594	3.91	0.000	1.066566	1.214291
x3	.9309467	.0318548	-2.09	0.037	.8705599	.9955222
x4	1.189399	.0603866	3.42	0.001	1.076742	1.313844
_cons	.8794245	.0746687	-1.51	0.130	.7446053	1.038654

Note: \_cons estimates baseline incidence rate.

```
. est store poisson
```

From GOF test, we can see that  $H_0$  is rejected. This indicates that poisson model should not be employed here. However, after using poisson model, we can see higher significance of all variables.

#### 4 Estimate models for $y_i$ assuming that the probability functions follow Negative Binomial probability distribution. Determine whether Negative Binomial regression model is appropriated in this case. Interpret your estimated result (sign and meaning (in term of incidence-rate ratios), overall test, individual test, pseudo $R^2$ , marginal effects).

```
. nbreg y x1 x2 x3 x4, nolog
```

```
Negative binomial regression                               Number of obs   =      232
                                                LR chi2(4)      =      21.24
Dispersion = mean                                         Prob > chi2     =      0.0003
Log likelihood = -317.49278                    Pseudo R2      =      0.0324
```

y	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
x1	.1285534	.0506934	2.54	0.011	.0291962	.2279106
x2	.151011	.0506477	2.98	0.003	.0517434	.2502785
x3	-.0672859	.0481376	-1.40	0.162	-.1616339	.0270621
x4	.1726312	.0707035	2.44	0.015	.034055	.3112075
_cons	-.1435596	.1177204	-1.22	0.223	-.3742874	.0871682
/lnalpha	.0479945	.2389531			-.4203449	.5163339

```
-----+-----
alpha | 1.049165 .2507012 .6568202 1.675872
-----+-----
```

```
LR test of alpha=0: chibar2(01) = 50.78 Prob >= chibar2 = 0.000
```

```
. nbreg y x1 x2 x3 x4,ir nolog
```

```
Negative binomial regression      Number of obs   =      232
LR chi2(4)                       =      21.24
Dispersion = mean                 Prob > chi2     =      0.0003
Log likelihood = -317.49278       Pseudo R2      =      0.0324
```

```
-----+-----
y | IRR Std. Err. z P>|z| [95% Conf. Interval]
-----+-----
x1 | 1.137182 .0576476 2.54 0.011 1.029627 1.255973
x2 | 1.163009 .0589037 2.98 0.003 1.053105 1.284383
x3 | .9349279 .0450052 -1.40 0.162 .8507526 1.027432
x4 | 1.188428 .084026 2.44 0.015 1.034641 1.365072
_cons | .8662692 .1019776 -1.22 0.223 .6877792 1.09108
-----+-----
/lnalpha | .0479945 .2389531 - .4203449 .5163339
-----+-----
alpha | 1.049165 .2507012 .6568202 1.675872
-----+-----
```

Note: Estimates are transformed only in the first equation.

Note: \_cons estimates baseline incidence rate.

```
LR test of alpha=0: chibar2(01) = 50.78 Prob >= chibar2 = 0.000
```

From LR test,  $H_0$  is rejected. This suggests that Negative Binomial model is more appropriated in this case. Overall significance is good except  $x_3$ . However, pseudo  $R^2$  is quite low.

**5 Estimate models for  $y_i$  assuming that the model is Zero Inflated Poisson ( $x_{1i}$ ,  $x_{2i}$ , and  $x_{3i}$  are independent variables in Poisson model and  $x_{4i}$  is independent variable in Inflated (Logit) model). Interpret your estimated result. Determine which model (Linear regression model, Poisson, Negative Binomial, or ZIP) is the most appropriated model in this case? Why? (provide the tests)**

```
. zip y x1 x2 x3, inflate(x4) forcevuong nolog
```

```
Zero-inflated Poisson regression      Number of obs   =      232
Nonzero obs                          =      106
Zero obs                              =      126
```

```
Inflation model = logit              LR chi2(3)      =      10.35
Log likelihood = -312.6158           Prob > chi2     =      0.0158
```

```
-----+-----
y | Coef. Std. Err. z P>|z| [95% Conf. Interval]
-----+-----
y |
x1 | .0805446 .0398159 2.02 0.043 .0025068 .1585824
x2 | .0857883 .0372107 2.31 0.021 .0128567 .1587199
x3 | -.0672468 .0357098 -1.88 0.060 -.1372367 .002743
_cons | .4589728 .1106031 4.15 0.000 .2421947 .6757508
-----+-----
```

```

-----+-----
inflate |
      x4 | -.2738532  .1212311  -2.26  0.024  -.5114618  -.0362446
      _cons | -.3379298  .1908217  -1.77  0.077  -.7119334  .0360738
-----+-----

```

Vuong test of zip vs. standard Poisson: z = 3.92 Pr>z = 0.0000  
Warning: The Vuong test is not appropriate for testing zero-inflation.

From Vuong test, Ho is rejected. This suggests that ZIP model is more appropriated than poisson model. Overall significance is good except x3.

**6 According to the above (1-5), determine the most appropriated model for this case. Give explanation why?**

```
est tab ols poisson nbreg zip, star(.1 .05 .01)
```

```

-----+-----
Variable |      ols      poisson      nbreg      zip
-----+-----
-
      x1 | .10162013**
      x2 | .13450436***
      x3 | -.07481941
      x4 | .16845626**
      _cons | .9568064***
-----+-----
Y
      x1 |      .09714739***      .12855344**      .08054462**
      x2 |      .12930238***      .15101096***      .08578827**
      x3 |      -.0715533**      -.06728591      -.06724685*
      x4 |      .17344816***      .17263123**
      _cons |      -.12848756      -.14355961      .45897275***
-----+-----
      /lnalpha |      .0479945
-----+-----
inflate |
      x4 |      -.27385319**
      _cons |      -.33792981*
-----+-----

```

legend: \* p<.1; \*\* p<.05; \*\*\* p<.01

From the table, we can see that ZIP has the most significance overall. Furthermore, Vuong test also suggest that ZIP model is more appropriated than poisson model.