

```
-----  
    name:  PK  
    log:  
/Users/priyakoirala/Desktop/school/econometrics/projects/project5/koirala_project5.log  
    log type:  text  
    opened on:  18 May 2023, 19:39:43
```

```
.  
. /*=====
```

> The purpose of this exercise is to understand how borrower characteristics  
> relate to the probability of home loan approval.

>

> Open the LOANAPP.dta data set. It is a random sample of people who applied for  
> home mortgage loans. It contains information about the people's characteristics  
> and a binary variable for whether their loan application got approved or not.

> =====\*/

```
.  
. use "/Users/priyakoirala/Desktop/school/econometrics/projects/project5/loanapp.dta"  
.
```

```

. /*=====
> (Q1): Use a linear probability model to estimate the relationship between the
> probability of loan approval (approve) and the following variables:
>
> X1=bankruptcy
> X2=hh_expenditures
> X3=term
> X4=apr
> X5=Black
> X6=Hispanic
>
> What type of standard errors should you use?
> =====*/
.
. sum

```

Variable	Obs	Mean	Std. dev.	Min	Max
gdlin	1,983	.9152799	.278535	0	1
bankruptcy	1,983	.068583	.2528074	0	1
hh_expendi~s	1,983	24.80028	7.122213	1	72
term	1,983	341.0671	64.52418	6	480
apr	1,983	205.2156	156.3076	25	4316
black	1,983	.0983359	.2978433	0	1
hispan	1,983	.0559758	.229933	0	1
male	1,968	.8130081	.3900041	0	1
approve	1,983	.878467	.3268281	0	1

```

.
. reg approve i.bankruptcy c.hh_expenditures c.term c.apr i.black i.hisp, robust

```

Linear regression	Number of obs	=	1,983
	F(6, 1976)	=	20.71
	Prob > F	=	0.0000
	R-squared	=	0.1102
	Root MSE	=	.30877

	approve	Coefficient	Robust std. err.	t	P> t	[95% conf. interval]
1.bankruptcy		-.3052161	.0430494	-7.09	0.000	-.3896431 -.2207891
hh_expenditures		-.0027087	.0011926	-2.27	0.023	-.0050476 -.0003699
term		.0001036	.000112	0.92	0.355	-.0001161 .0003232
apr		.000101	.0000428	2.36	0.018	.0000171 .0001849
1.black		-.1840989	.032968	-5.58	0.000	-.2487546 -.1194433
1.hispan		-.1175171	.0398279	-2.95	0.003	-.1956262 -.039408
_cons		.9352098	.0461626	20.26	0.000	.8446774 1.025742

```

.
. /* For a Linear Probability Model, we should always use heteroskedasticity robust
> standard errors as the variance in these models are not constant, therefore, they
> are naturally heteroskedastic. */
.

```

```
. /*=====
> (Q2): What is the marginal effect of household expenditures on loan approval
> probability?
> =====*/
.
. /* Marginal Effect: -0.0027087.
>
> For every $1,000 increase in household expenditure per year, the probability of
> loan approval decreases by 0.0027. */
.
```

```
. /*=====
> (Q3): Estimate the same model as in (Q1) but use the probit functional form.
>
> What is the marginal effect of household expenditures on loan approval
> probability when all variables are held at their mean values?
> =====*/
. probit approve i.bankruptcy c.hh_expenditures c.term c.apr i.black i.hisp, robust
```

```
Iteration 0: log pseudolikelihood = -733.64726
Iteration 1: log pseudolikelihood = -650.20317
Iteration 2: log pseudolikelihood = -648.12078
Iteration 3: log pseudolikelihood = -648.11547
Iteration 4: log pseudolikelihood = -648.11547
```

```
Probit regression                                Number of obs = 1,983
                                                Wald chi2(6) = 160.06
                                                Prob > chi2 = 0.0000
Log pseudolikelihood = -648.11547                Pseudo R2 = 0.1166
```

		Robust				
approve	Coefficient	std. err.	z	P> z	[95% conf. interval]	
1.bankruptcy	-1.04896	.1214742	-8.64	0.000	-1.287046	-.8108753
hh_expenditures	-.014929	.0060127	-2.48	0.013	-.0267136	-.0031443
term	.0005265	.0006085	0.87	0.387	-.0006661	.0017192
apr	.0012518	.0004483	2.79	0.005	.0003731	.0021304
1.black	-.7127674	.1087012	-6.56	0.000	-.9258178	-.499717
1.hispan	-.5004237	.1437615	-3.48	0.000	-.7821911	-.2186563
_cons	1.37818	.2605525	5.29	0.000	.8675065	1.888853

Note: 0 failures and 1 success completely determined.

```
.
. margins, dydx(_all) at((means)_all)
```

```
Conditional marginal effects                                Number of obs = 1,983
Model VCE: Robust
```

```
Expression: Pr(approve), predict()
dy/dx wrt: 1.bankruptcy hh_expenditures term apr 1.black 1.hispan
At: 0.bankruptcy = .931417 (mean)
    1.bankruptcy = .068583 (mean)
    hh_expenditures = 24.80028 (mean)
    term = 341.0671 (mean)
    apr = 205.2156 (mean)
    0.black = .9016641 (mean)
    1.black = .0983359 (mean)
    0.hispan = .9440242 (mean)
    1.hispan = .0559758 (mean)
```

		Delta-method				
	dy/dx	std. err.	z	P> z	[95% conf. interval]	
1.bankruptcy	-.2939985	.0443986	-6.62	0.000	-.3810181	-.2069789
hh_expenditures	-.0026442	.0010571	-2.50	0.012	-.004716	-.0005723
term	.0000933	.0001077	0.87	0.386	-.0001178	.0003043
apr	.0002217	.0000779	2.85	0.004	.000069	.0003744

Note:  $dy/dx$  for factor levels is the discrete change from the base level.

```
. /* Marginal Effect: -0.0026442.
>
> For every $1,000 increase in household expenditure per year (holding the mean
> value for all of the variables in the sample) the probability of loan approval
> decreases by 0.0026.*/
.
```

```

. /*=====
> (Q4): Test whether loan term and loan APR are jointly statistically significant.
> Make sure to write the null and alternative hypotheses (not in Stata format).
> =====*/
.
. test term apr

( 1)  [approve]term = 0
( 2)  [approve]apr = 0

           chi2( 2) =      8.39
       Prob > chi2 =      0.0150

.
. /* H_0: beta3 = beta4 = 0
>    H_1: beta3 != 0 and/or beta4 != 0
>
>    p-value of Chi-squared statistic: |0.0150| < 0.05
>
>    We reject the null hypothesis of no joint statistical significance.
>
> We accept the alternate hypothesis and conclude that loan term and loan apr are
> jointly statistically significant. */
.

```

```
. /*=====
> (Q5): What is the predicted probability of loan approval for someone who has
> never filed for bankruptcy, has $25,000 in yearly household expenditures,
> wants a 360 month term mortgage at $205 APR, is not Black, and is not Hispanic?
> =====*/
.
. probit approve i.bankruptcy c.hh_expenditures c.term c.apr i.black i.hisp, robust
```

```
Iteration 0: log pseudolikelihood = -733.64726
Iteration 1: log pseudolikelihood = -650.20317
Iteration 2: log pseudolikelihood = -648.12078
Iteration 3: log pseudolikelihood = -648.11547
Iteration 4: log pseudolikelihood = -648.11547
```

```
Probit regression                                Number of obs = 1,983
                                                Wald chi2(6) = 160.06
                                                Prob > chi2 = 0.0000
Log pseudolikelihood = -648.11547              Pseudo R2 = 0.1166
```

-----							
		Robust					
approve	Coefficient	std. err.	z	P> z	[95% conf. interval]		
-----							
1.bankruptcy	-1.04896	.1214742	-8.64	0.000	-1.287046	-.8108753	
hh_expenditures	-.014929	.0060127	-2.48	0.013	-.0267136	-.0031443	
term	.0005265	.0006085	0.87	0.387	-.0006661	.0017192	
apr	.0012518	.0004483	2.79	0.005	.0003731	.0021304	
1.black	-.7127674	.1087012	-6.56	0.000	-.9258178	-.499717	
1.hispan	-.5004237	.1437615	-3.48	0.000	-.7821911	-.2186563	
_cons	1.37818	.2605525	5.29	0.000	.8675065	1.888853	
-----							

Note: 0 failures and 1 success completely determined.

```
.
. scalar yhat_nonblack = _b[_cons] + _b[hh_expenditures]*25 + _b[term]*360 + _b[apr]
> ]*205
```

```
.
. display yhat_nonblack
1.4511299
```

```
.
. /* The standard normal distribution for Phi coefficient 1.45 is 0.92647. */
```

```
.
. display normprob(_b[_cons] + _b[hh_expenditures]*25 + _b[term]*360 + _b[apr]*205)
.92662816
```

```
.
. /* The predicted probability of loan approval for someone with these
> characteristics is 0.926. */
.
```

```
. /*=====
> (Q6) What is the relationship between being Black and home loan approval for an
> applicant with otherwise the same characteristics as the applicant in (Q5)?
>
> Calculate the result by hand first, and then use Stata to confirm.
> =====*/
.
. scalar yhat_nonblack = _b[_cons] + _b[hh_expenditures]*25 + _b[term]*360 + _b[apr
> ]*205 + _b[1.black]
.
. display yhat_nonblack
.73836257
.
. /* The standard normal distribution for Phi coefficient 0.73 is 0.76730.
> The predicted probability of home loan approval for a Black applicant with
> otherwise the same characteristics in (Q5) is 0.767.*/
.
. display normprob(_b[_cons] + _b[hh_expenditures]*25 + _b[term]*360 + _b[apr]*205
> + _b[1.black])
.76985292
.
. /* Black applicants: 0.767
> Non-Black applicants: 0.926
> 0.767 - 0.926 = -0.159
>
> The difference in loan approval probability for two applicants with the same
> characteristics where one person is Black and the other is not Black is 0.159. */
.
. margins, dydx(_all) at(bankruptcy=0 hh_expenditures=25 term=360 apr=205 black=0 h
> isp=0)
```

Conditional marginal effects  
Model VCE: Robust

Number of obs = 1,983

```
Expression: Pr(approve), predict()
dy/dx wrt: 1.bankruptcy hh_expenditures term apr 1.black 1.hispan
At: bankruptcy      = 0
    hh_expenditures = 25
    term            = 360
    apr             = 205
    black           = 0
    hispan          = 0
```

		Delta-method				[95% conf. interval]	
		dy/dx	std. err.	z	P> z		
1.bankruptcy		-.2704078	.0431548	-6.27	0.000	-.3549896	-.185826
hh_expenditures		-.0020781	.0008337	-2.49	0.013	-.0037122	-.0004441
term		.0000733	.0000832	0.88	0.379	-.0000898	.0002364
apr		.0001743	.0000619	2.81	0.005	.0000528	.0002957
1.black		-.1567752	.0310937	-5.04	0.000	-.2177178	-.0958327
1.hispan		-.0975049	.0353522	-2.76	0.006	-.1667939	-.0282159

Note: dy/dx for factor levels is the discrete change from the base level.

```
. /* dy/dx for Black applicant: -0.1567752
>
> There seems to be a negative relationship with being Black and approved for a home
> loan. The probability of a Black applicant being approved for a loan is 0.157
> lower than a non-Black applicant with the same characteristics. */
```



```
. /*=====
> (Q7) Now estimate a logit model where Y=approve and include the following Xs:
> X1=gdlin
> X2=hh_expenditures
> X3=black
> X4=Hispanic
>
> What is the marginal effect of household expenditures on loan approval
> probability for someone who meets the loan guidelines, has $30,000 per year
> in expenditures, is not Black, and is Hispanic?
> =====*/
```

```
. logit approve i.gdlin c.hh_expenditures i.black i.hisp, robust
```

```
Iteration 0: log pseudolikelihood = -733.64726
Iteration 1: log pseudolikelihood = -566.40839
Iteration 2: log pseudolikelihood = -498.48485
Iteration 3: log pseudolikelihood = -495.4255
Iteration 4: log pseudolikelihood = -495.42102
Iteration 5: log pseudolikelihood = -495.42102
```

Logistic regression

Log pseudolikelihood = -495.42102

Number of obs = 1,983  
Wald chi2(4) = 346.80  
Prob > chi2 = 0.0000  
Pseudo R2 = 0.3247

	approve	Coefficient	Robust std. err.	z	P> z	[95% conf. interval]	
1.gdlin		3.804443	.2142151	17.76	0.000	3.384589	4.224297
hh_expenditures		-.0181353	.0144019	-1.26	0.208	-.0463624	.0100918
1.black		-.9566613	.2329437	-4.11	0.000	-1.413223	-.5000999
1.hispan		-.975948	.3167225	-3.08	0.002	-1.596713	-.3551833
_cons		-.4367085	.4143761	-1.05	0.292	-1.248871	.3754538

```
. margins, dydx(hh_expenditures) at (gdlin=1 hh_expenditures=30 black=0 hisp=1)
```

Conditional marginal effects  
Model VCE: Robust

Number of obs = 1,983

```
Expression: Pr(approve), predict()
dy/dx wrt: hh_expenditures
At: gdlin = 1
hh_expenditures = 30
black = 0
hispan = 1
```

		Delta-method dy/dx	std. err.	z	P> z	[95% conf. interval]	
hh_expenditures		-.0021328	.001865	-1.14	0.253	-.0057882	.0015226

```
. /* Marginal Effect:-0.0021328
```

```
>
> For every $1,000 increase in household expenditure per year, the probability of
> loan approval decreases by 0.0021 for an applicant with these characteristics. */
```

```

. /*=====
> (Q8) Is the marginal effect of household expenditures in (Q7) statistically
> significant?
> =====*/
.
. /* H_0: beta2 = 0
>    H_1: beta2 != 0
>
>    t-test = |-1.14| < 1.96
>
> We fail to reject the null hypothesis at the 5% significance level. We conclude
> that the marginal effect of household expenditures in (Q7) is not statistically
> significant. */
.

```

```

. /*=====
> (Q9) What is the predicted probability of loan approval for a person with the
> characteristics described in (Q7)?
>
> Please calculate by hand. Then you may use Stata to confirm.
> =====*/
.
. scalar yhat_hisp = _b[_cons] + _b[1.gdlin] + _b[hh_expenditures]*30 + _b[1.hisp]

.
. display yhat_hisp
1.8477279

.
. /* 1/(1+e^-1.85) = 0.86
>
> The predicted probability of loan approval for an applicant with the
> characteristics described in (Q7) is 0.86. */
.
. display 1/[1+exp(-1*( _b[_cons]+_b[1.gdlin]+_b[hh_expenditures]*30+_b[1.hisp]))]
.86386011
.

```

```

. /*=====
> (Q10) What is the predicted probability of loan approval for a person with the
> same characteristics as described in (Q7), except the person is non-Hispanic?
>
> Please calculate by hand.
> =====*/
.
. scalar yhat_nonhisp = _b[_cons] + _b[1.gdlin] + _b[hh_expenditures]*30

.
. display yhat_nonhisp
2.8236759

.
. /* 1/(1+e^-2.82) = 0.94
>
> The predicated probability of loan approval for a Non-Hispanic person with the
> same characteristics described in (Q7) is 0.94. */
.
. display 1/[1+exp(-1*(_b[_cons]+_b[1.gdlin]+_b[hh_expenditures]*30))]
.9439419

.
. /* Hispanic applicant: 0.86
> Non-Hispanic Applicant: 0.94
>
> 0.86 - 0.94 = -0.08
>
> The difference in loan approval probability between two applicants with the same
> characteristics where one person is Hispanic and the other one is Non-Hispanic is
> 0.08. */
.
. margins, dydx(hisp) at(gdlin=1 hh_expenditures=30 black=0 hisp=1)

Conditional marginal effects                                Number of obs = 1,983
Model VCE: Robust

Expression: Pr(approve), predict()
dy/dx wrt: 1.hispan
At: gdlin          = 1
    hh_expenditures = 30
    black           = 0
    hispan          = 1

-----
      |               Delta-method
      |      dy/dx   std. err.      z    P>|z|    [95% conf. interval]
-----+-----
1.hispan |   -.0800818   .0359376   -2.23   0.026   -.1505182   -.0096454
-----

Note: dy/dx for factor levels is the discrete change from the base level.

.
. /* dy/dx for Hispanic applicant: -.0800818
>
> Therefore, according to our estimation model, the probability of a Hispanic
> applicant being approved for a loan is 0.08 lower than a non-Hispanic.*/
.

```

```

. /*=====
> (Q11) Based on your answers to the questions in this project, do you think there
> is discrimination in the market for home loans?
> =====*/
.
> /* Based on the answers to the questions in this project, it seems that there is
> some discrimination in the market for home loans. Primarily, there seems to be a
> bias against Hispanic and Black applicants.
>
> We simulated models by changing the characteristics of applicants such as if the
> applicant has filed bankruptcy before, their household expenditures and the
> borrowing guidelines. We then compared the models from Hispanic applicants to non-
> Hispanic applicants, and black applicants to non-Black applicants and the
> difference suggests that Hispanic and Black applicants are less likely to be
> approved for a loan possessing the same characteristics as their non-Black or
> non-Hispanic counterparts.
>
> However, we should also note that there are other variables we may not have
> accounted for in our sample, and therefore, this result could also be the cause of
> omitted variable bias. For example, credit history, capital, and work history may
> be some other factors that lenders investigate when approving someone for a loan.
> In order to get a more accurate estimate from our models we may need more
> information regarding the applicants.
>
> Based on the project alone, it seems to be that Black and Hispanic applicants face
> discrimination in the loan approval process. */
>
> /*=====
> =====*/
>

```

end of do-file