Exercise 11-1.

Suppose one of your co-workers is expecting a baby and you are participating in an office pool to predict the date of birth. Assuming that bets are placed during the 30th week of pregnancy, what variables could you use to make the best prediction? You should limit yourself to variables that are known before the birth, which variables are best to predict the baby weight

```
In [1]:
# setup the data frame and import the packages
import sys
import numpy as np
import math
import pandas as pd
import first
import thinkplot
import thinkstats2
from scipy import stats
import statsmodels.formula.api as smf
#read data and build dataframe
live, firsts, others = first.MakeFrames()
live = live.dropna(subset=['agepreg', 'totalwgt lb'])
#read data and build dataframe
live, firsts, others = first.MakeFrames()
#live = live.dropna(subset=['agepreg', 'totalwgt lb'])
C:\Users\qyanr\qyan-python-workspace\DSC-530\nsfq.py:68: FutureWarning: A value is trying
to be set on a copy of a DataFrame or Series through chained assignment using an inplace
method.
The behavior will change in pandas 3.0. This inplace method will never work because the i
ntermediate object on which we are setting values always behaves as a copy.
For example, when doing 'df[col].method(value, inplace=True)', try using 'df.method({col:
value}, inplace=True)' or df[col] = df[col].method(value) instead, to perform the operati
on inplace on the original object.
  df.birthwgt lb.replace(na vals, np.nan, inplace=True)
C:\Users\gyanr\gyan-python-workspace\DSC-530\nsfg.py:69: FutureWarning: A value is trying
to be set on a copy of a DataFrame or Series through chained assignment using an inplace
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C:\Users\gyanr\gyan-python-workspace\DSC-530\nsfg.py:70: FutureWarning: A value is trying
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```

C:\Users\gyanr\gyan-python-workspace\DSC-530\nsfq.py:72: FutureWarning: A value is trying

df.hpagelb.replace(na vals, np.nan, inplace=True)

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For example, when doing 'df[col].method(value, inplace=True)', try using 'df.method({col: value}, inplace=True)' or df[col] = df[col].method(value) instead, to perform the operati on inplace on the original object.

df.babysex.replace([7, 9], np.nan, inplace=True)

C:\Users\gyanr\gyan-python-workspace\DSC-530\nsfg.py:73: FutureWarning: A value is trying to be set on a copy of a DataFrame or Series through chained assignment using an inplace method.

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df.nbrnaliv.replace([9], np.nan, inplace=True)

C:\Users\gyanr\gyan-python-workspace\DSC-530\nsfg.py:68: FutureWarning: A value is trying to be set on a copy of a DataFrame or Series through chained assignment using an inplace method.

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df.birthwgt_lb.replace(na_vals, np.nan, inplace=True)

C:\Users\gyanr\gyan-python-workspace\DSC-530\nsfg.py:69: FutureWarning: A value is trying to be set on a copy of a DataFrame or Series through chained assignment using an inplace method.

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```
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on inplace on the original object.
 df.nbrnaliv.replace([9], np.nan, inplace=True)
In [2]:
```

```
#data mining
import nsfg
live = live[live.prglngth>30]
resp = nsfg.ReadFemResp()
resp.index = resp.caseid
join = live.join(resp, on='caseid', rsuffix=' r')
join.shape
Out[2]:
(8884, 3331)
In [3]:
#this function will try out all variables and find coresponding R2 for each variables.
# the variable for baby birth date is cmbabdob (Century Month for baby's or babies' date
of birth (delivery date))
import patsy
def GoMining(df):
    variables = []
    for name in df.columns:
        try:
            if df[name].var() < 1e-7:</pre>
                continue
            formula = 'cmbabdob ~ agepreg + ' + name
            model = smf.ols(formula, data=df)
            if model.nobs < len(df)/2:</pre>
                continue
            results = model.fit()
        except (ValueError, TypeError, patsy.PatsyError) as e:
            continue
        variables.append((results.rsquared, name))
    return variables
variables = GoMining(join)
In [4]:
```

```
# this is the function to rank and pick out the top 20 variables with largest R2 values
import re
def ReadVariables():
    """Reads Stata dictionary files for NSFG data.
   returns: DataFrame that maps variables names to descriptions
   vars1 = thinkstats2.ReadStataDct('2002FemPreg.dct').variables
   vars2 = thinkstats2.ReadStataDct('2002FemResp.dct').variables
   all vars = pd.concat([vars1, vars2])
   all vars.index = all vars.name
   return all vars
```

```
def MiningReport(variables, n=20):
    """Prints variables with the highest R^2.

    t: list of (R^2, variable name) pairs
    n: number of pairs to print
    """
    all_vars = ReadVariables()

    variables.sort(reverse=True)
    for r2, name in variables[:n]:
        key = re.sub('_r$', '', name)
        try:
        desc = all_vars.loc[key].desc
        if isinstance(desc, pd.Series):
            desc = desc[0]
        print(name, r2, desc)
        except (KeyError, IndexError):
            print(name, r2)
```

In [5]:

```
MiningReport (variables)

cmprgend 1.0 CM FOR PREGNANCY END DATE (REGARDLESS OF OUTCOME)

cmbabdob 1.0 CM FOR BABY'S OR BABIES' DATE OF BIRTH (DELIVERY DATE)
```

```
cmbabdob 1.0 CM FOR BABY'S OR BABIES' DATE OF BIRTH (DELIVERY DATE)
maternly i 0.9644611042675953 MATERNLY IMPUTATION FLAG
cmintfin 0.9345564585496872 CM FOR DATE OF END OF PREGNANCY INTERVAL
datend i 0.9143157249464409 DATEND IMPUTATION FLAG
datecon i 0.8666324562960489 DATECON IMPUTATION FLAG
frsteatd 0.8126056954505722 AGE (IN MOS) WHEN 1ST SUPPLEMENTED - 1ST FROM THIS PREG
pncarewk i 0.6920031959762878 PNCAREWK IMPUTATION FLAG
cmprqbeq 0.6331523928948615 CM FOR PREGNANCY START DATE
cmfstprg r 0.5292820118493895 CM FOR R'S FIRST COMPLETED PREGNANCY
cmfstprg 0.5292820118493895 CM FOR R'S FIRST COMPLETED PREGNANCY
cmlastlb r 0.45688384405815197 CM FOR R'S MOST RECENT LIVE BIRTH
cmlastlb 0.45688384405815197 CM FOR R'S MOST RECENT LIVE BIRTH
agepreg i 0.4014168719339821 AGEPREG IMPUTATION FLAG
\verb|agecon_i 0.3911090825296043| AGECON IMPUTATION FLAG|
bfeedwks i 0.2702522130679593 BFEEDWKS IMPUTATION FLAG
cmhsgrad 0.2585764653626096 CENTURY MONTH OF HIGH SCHOOL GRADUATION
kidage 0.16737748339080993 CURRENT AGE (IN MOS) OF R'S CHILD(REN) FROM THIS PREGNANCY
cmlstprg r 0.16299057868186884 CM FOR R'S MOST RECENT COMPLETED PREGNANCY
cmlstprg 0.16299057868186884 CM FOR R'S MOST RECENT COMPLETED PREGNANCY
```

```
C:\Users\gyanr\AppData\Local\Temp\ipykernel_10044\2441321277.py:30: FutureWarning: Series
.__getitem__ treating keys as positions is deprecated. In a future version, integer keys
will always be treated as labels (consistent with DataFrame behavior). To access a value
by position, use `ser.iloc[pos]`
   desc = desc[0]
```

Conclusion:

From the list above, need to keep those variables only know before brith and known to that coworkers (assume this could be first baby, the couples already know the sex of baby ahead, do not know if the pregency will finish).

Besides age of pregnency, The following variables will be picked to make date of birth prediction.

- 1. agecon 0.10203149928156052 AGE AT TIME OF CONCEPTION
- 2. race r 0.016199503586252995 RACE
- 3. race 0.016199503586252995 RACE
- 4. paydu 0.014003795578114597 IB-10 CURRENT LIVING QUARTERS OWNED/RENTED, ETC
- 5. totincr 0.011870069031173602 TOTAL INCOME OF R'S FAMILY
- 6. marcon03 0.011752599354395654 FORMAL MARITAL STATUS WHEN PREGNANCY BEGAN 3RD
- 7. cebow 0.011437770919637158 NUMBER OF CHILDREN BORN OUT OF WEDLOCK

-

Example 11-3.

If the quantity you want to predict is a count, you can use Poisson regression, which is implemented in StatsModels with a function called poisson. It works the same way as ols and logit. As an exercise, let's use it to predict how many children a woman has born; in the NSFG dataset, this variable is called numbabes.

Suppose you meet a woman who is 35 years old, black, and a college graduate whose annual household income exceeds \$75,000. How many children would you predict she has born?

based on the condition, we need to select independent variables as age and square age (age_r + age2), race(race), income (totincr) and eductaion(educat) to make prediction.

In [6]:

```
join.numbabes.replace([97], np.nan, inplace=True)
join['age2'] = join.age_r**2

formula = 'numbabes ~ age_r + age2 + C(race) + totincr + educat'
model = smf.poisson(formula, data=join)
results = model.fit()
results.summary()
```

Optimization terminated successfully.

Current function value: 1.677002

Iterations 7

C:\Users\gyanr\AppData\Local\Temp\ipykernel_10044\2591200501.py:1: FutureWarning: A value is trying to be set on a copy of a DataFrame or Series through chained assignment using a n inplace method.

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For example, when doing 'df[col].method(value, inplace=True)', try using 'df.method({col: value}, inplace=True)' or df[col] = df[col].method(value) instead, to perform the operati on inplace on the original object.

join.numbabes.replace([97], np.nan, inplace=True)

Out[6]:

Poisson Regression Results

Dep. Vari	able:	numbabes N		No. Obs	ervation	s:	8884	
Model:		Poisson		Df I	Residua	ls:	8877	
Method:		MLE			Df Mod	el:	6	
	Date: Th	u, 28 No	v 2024	Pseud	do R-sq	u.: 0	.03686	
Time:		22:00:47		Log-L	ikelihoo	d: -	14898.	
converged:		True			LL-Nu	ili: -	15469.	
Covariance Type:		nonrobust		LLI	R p-valu	e: 3.68	1e-243	
	coef	std err	2	z P>izi	[0.025	0.975]		
Intercept	-1.0324	0.169	-6.098	0.000	-1.364	-0.701		
C(race)[T.2]	-0.1401	0.015	-9.479	0.000	-0.169	-0.111		
C(race)[T.3]	-0.0991	0.025	-4.029	0.000	-0.147	-0.051		
age_r	0.1556	0.010	15.006	0.000	0.135	0.176		
age2	-0.0020	0.000	-13.102	2 0.000	-0.002	-0.002		
totincr		0.002	-9.830	0.000		-0.015		

educat -0.0471 0.003 -16.076 0.000 -0.053 -0.041

In [7]:

columns - [laco rl | laco2] | Index | Itatinani | Laducatii

```
new = pd.DataFrame([[35, 35**2, 1, 14, 16]], columns=columns)
results.predict(new)

Out[7]:
```

0 2.496802 dtype: float64

Conclusion:

From the model, it predict this woman will give birth to 2-3 babies.

Example 11-4.

f the quantity you want to predict is categorical, you can use multinomial logistic regression, which is implemented in StatsModels with a function called mnlogit. As an exercise, let's use it to guess whether a woman is married, cohabitating, widowed, divorced, separated, or never married; in the NSFG dataset, marital status is encoded in a variable called rmarital.

Suppose you meet a woman who is 25 years old, white, and a high school graduate whose annual household income is about \$45,000. What is the probability that she is married, cohabitating, etc?

based on the condition, we need to select independent variables as age and square age (age $_r$ + age2), race(race), income (totincr) and eductaion(educat) to make prediction.

```
In [8]:
```

```
formula='rmarital ~ age_r + age2 + C(race) + totincr + educat'
model = smf.mnlogit(formula, data=join)
results = model.fit()
results.summary()
```

Optimization terminated successfully.

Current function value: 1.084053

Iterations 8

Out[8]:

MNLogit Regression Results

Dep. Variable:		rn	narital I	No. Obs	s: 88	384	
Model:		MNLogit		Df I	ls: 88	349	
Method:		MLE			el:	30	
Date: Th		u, 28 Nov 2024		Pseud	u.: 0.16	82	
Time:		22:00:49		Log-L	d: -963	0.7	
converged:		True			LL-Nu	ili: -115	79.
Covariance Type:		noni	robust	LLI	e: 0.0	000	
rmarital=2	coef	std err	z	P> z	[0.025	0.975]	
Intercept	9.0156	0.805	11.199	0.000	7.438	10.593	
C(race)[T.2]	-0.9237	0.089	-10.418	0.000	-1.097	-0.750	
C(race)[T.3]	-0.6179	0.136	-4.536	0.000	-0.885	-0.351	
age_r	-0.3635	0.051	-7.150	0.000	-0.463	-0.264	
age2	0.0048	0.001	6.103	0.000	0.003	0.006	
totincr	-0.1310	0.012	-11.337	0.000	-0.154	-0.108	
educat	-0.1953	0.019	-10.424	0.000	-0.232	-0.159	
rmarital=3	coef	std err	z	P>IzI	[0.025	0.975]	
Intercept	2.9570	3.020	0.979	0.328	-2.963	8.877	

```
C(race)[T.2] -0.4411
                      0.237
                             -1.863 0.062 -0.905
                                                   0.023
C(race)[T.3]
            0.0591
                      0.336
                              0.176 0.860 -0.600
                                                   0.718
     age_r -0.3177
                      0.177
                             -1.798 0.072 -0.664
                                                   0.029
     age2 0.0064
                      0.003
                              2.528 0.011 0.001
                                                   0.011
    totincr -0.3258
                     0.032 -10.175 0.000 -0.389
                                                  -0.263
    educat -0.0991
                      0.048
                             -2.050 0.040 -0.194
                                                  -0.004
 rmarital=4
               coef
                    std err
                                  z P>|z| [0.025
                                                  0.975]
  Intercept -3.5238
                      1.205
                             -2.924 0.003 -5.886 -1.162
C(race)[T.2] -0.3213
                      0.093
                             -3.445 0.001 -0.504 -0.139
C(race)[T.3] -0.7706
                             -4.509 0.000 -1.106 -0.436
                     0.171
            0.1155
                      0.071
                              1.626 0.104 -0.024
     age r
      age2 -0.0007
                      0.001
                             -0.701 0.483 -0.003
                                                   0.001
    totincr -0.2276
                     0.012 -19.621 0.000 -0.250 -0.205
    educat 0.0667
                     0.017
                              3.995 0.000 0.034
                                                   0.099
 rmarital=5
              coef std err
                                 z P>|z| [0.025
                                                  0.975]
  Intercept -2.8963
                      1.305
                             -2.220 0.026 -5.453
                                                  -0.339
C(race)[T.2] -1.0407
                      0.104 -10.038 0.000 -1.244
                                                 -0.837
                             -3.635 0.000 -0.871 -0.261
C(race)[T.3] -0.5661
                      0.156
     age_r
            0.2411
                      0.079
                              3.038 0.002 0.086
                                                   0.397
                             -2.977 0.003 -0.006 -0.001
      age2 -0.0035
                     0.001
    totincr -0.2932
                      0.015 -20.159 0.000 -0.322 -0.265
    educat -0.0174
                     0.021
                             -0.813 0.416 -0.059
                                                   0.025
 rmarital=6
                    std err
                                 z P>|z| [0.025
              coef
                                                  0.975]
  Intercept 8.0533
                     0.814
                              9.890 0.000 6.457
                                                   9.649
C(race)[T.2] -2.1871
                      0.080 -27.211 0.000 -2.345
                     0.138 -14.188 0.000 -2.232 -1.690
C(race)[T.3] -1.9611
     age_r -0.2127
                     0.052
                            -4.122 0.000 -0.314 -0.112
      age2 0.0019
                     0.001
                              2.321 0.020 0.000
                                                   0.003
    totincr -0.2945
                     0.012 -25.320 0.000 -0.317 -0.272
    educat -0.0742
                     0.018
                            -4.169 0.000 -0.109 -0.039
```

In [9]:

```
#make prediction
columns = ['age_r', 'age2', 'race', 'totincr', 'educat']
new = pd.DataFrame([[25, 25**2, 2, 11, 12]], columns=columns)
results.predict(new)
```

Out[9]:

```
        0
        1
        2
        3
        4
        5

        0
        0.750028
        0.126397
        0.001564
        0.033403
        0.021485
        0.067122
```

Conclusion:

From the model, it predict this woman will have 75% chance to be married, some chance (13%) to be cohabitating. little chance to be widowed or separated)