NHANES_FPED_EDAipynb

April 4, 2021

Objectives

Perform more exploratory analysis on the input data for the logistic regression model. The goal is to come up with data filters and transformations that can be used to address assumptions about the research question. The model can then be re-evaulated and compared to the intial prediction rate results.

Re-evaluate model with different techniques for constraining the observation space:

Use subsets of meals within calorie distribution categories. For example: is there any difference between low calories and high calorie meals?

Explore the possibility of transforming the input data from continuous to categorical.The component variables contain a high amount of 0's with a right tail heavy distribution.

Use only meals that are consumed at home, since eating out is generally more unhealthy

Constrain the non-seafood class to meals containing meat only. This will exclude vegetarian meals and compare seafood consumers to meat consumers.

Use a participant age filter, to look for adult participants only.

Attempt a classification of meals that contain both meat and seafood.

Convert input variables to standard units (some are in grams, some are in cups, etc.)

Input from research collaborators:

```
Obtain a more educated selection of input variables.
```

```
Attempt to weight the observations, to adjust for the survey design methods.
```

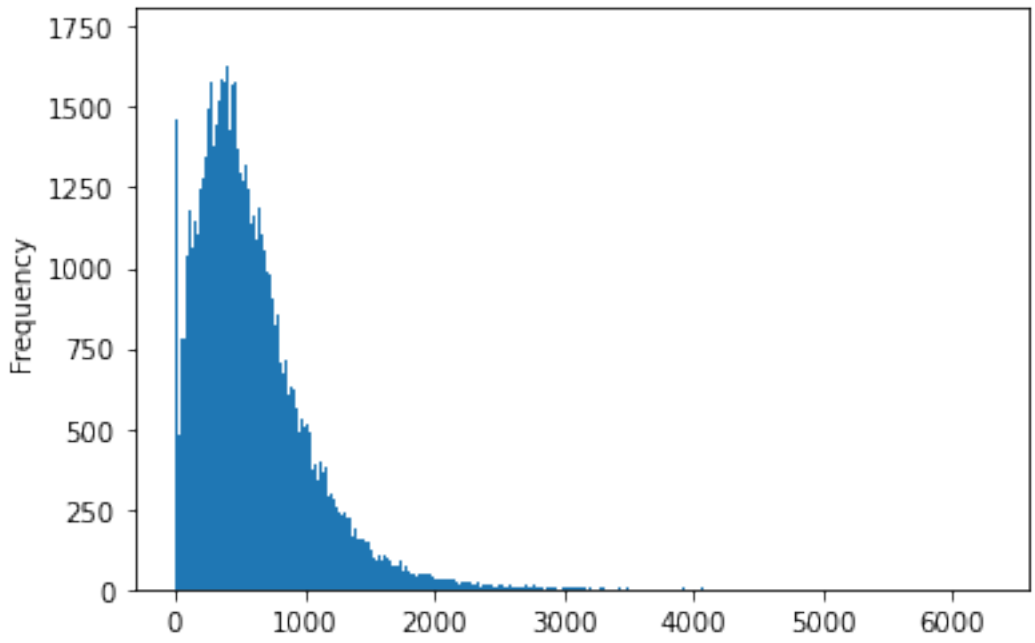
Section 1: Meal Energy Statistics

[19]:
```
import pandas as pd

#Read data frame and add plant pf total variable
df = pd.read_csv('../../Data/nhanes_full_pre_proc.csv')
df['PF_PLANT_D_TOTAL'] =␣
 ↪df['PF_EGGS']+df['PF_SOY']+df['PF_NUTSDS']+df['PF_LEGUMES']

df['DR1IKCAL'].plot.hist(bins=500)
```

[19]:
```
981
```



The plot above is displaying the distribution of the KCAL variable. One apparent potential issue is the left most bar, indicating meals where this variable is equal to 0.

[24]:
```
zero_kcal_meals = df[df['DR1IKCAL']==0]
len(zero_kcal_meals)
```

[24]:
```
981
```

Indeed, there are 981 meals where KCAL = 0.

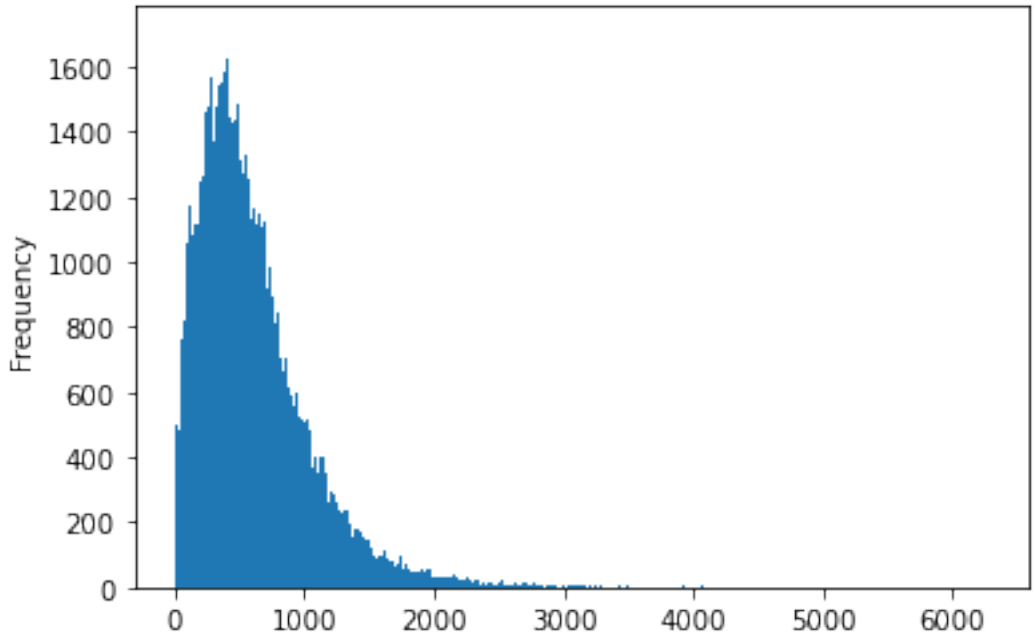
[96]:
```
zero_kcal_meals[['SEQN', 'DR1.030Z', 'DR1.020', 'DR1IKCAL']].head(5)
```

[96]:

	SEQN	DR1.030Z	DR1.020	DR1IKCAL
361	31330	3	90000	0.0
413	31364	14	86400	0.0
489	31411	15	54000	0.0
490	31411	15	70200	0.0
491	31411	15	75600	0.0

[72]:
```
#Filter out 0 KCAL meals
df = df[df['DR1IKCAL'] > 0]
df['DR1IKCAL'].plot.hist(bins=500)
```

[72]:
```
<matplotlib.axes._subplots.AxesSubplot at 0x7feaa52b32b0>
```



[58]:
```
#Obtain statistics for KCAL
df['DR1IKCAL'].describe()
```

[58]:
```
count    101731.000000
mean        605.018942
std         445.749113
min           1.000000
25%         297.000000
50%         508.000000
75%         798.000000
max        6264.000000
Name: DR1IKCAL, dtype: float64
```

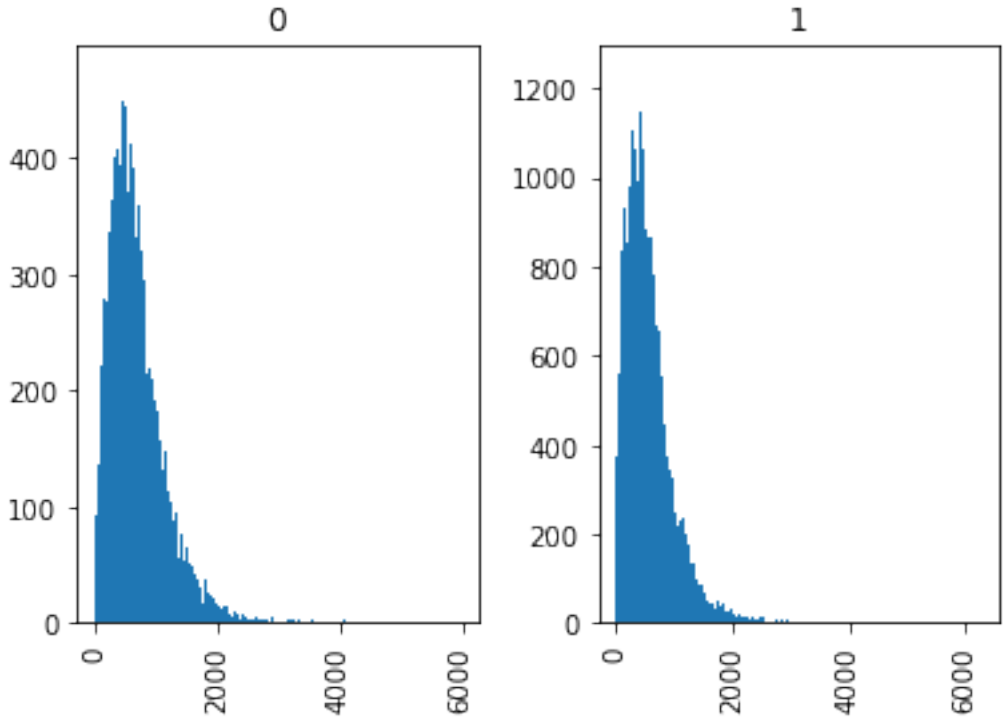
[69]:
```
#Home vs Out Meals
df.groupby('eathome')['DR1IKCAL'].describe()
```

[69]:

eathome	count	mean	std	min	25%	50%	75%	max
0	32830.0	687.545111	476.684325	1.0	360.0	587.0	899.0	5957.0
1	68901.0	565.696811	424.626035	1.0	271.0	471.0	751.0	6264.0

[77]:
```
#Home vs Out Meals
df['DR1IKCAL'].hist(bins=500, by=df['eathome'])
```

[77]:
```
array([<matplotlib.axes._subplots.AxesSubplot object at 0x7feacb564ee0>,
       <matplotlib.axes._subplots.AxesSubplot object at 0x7feaa686e6d0>],
      dtype=object)
```



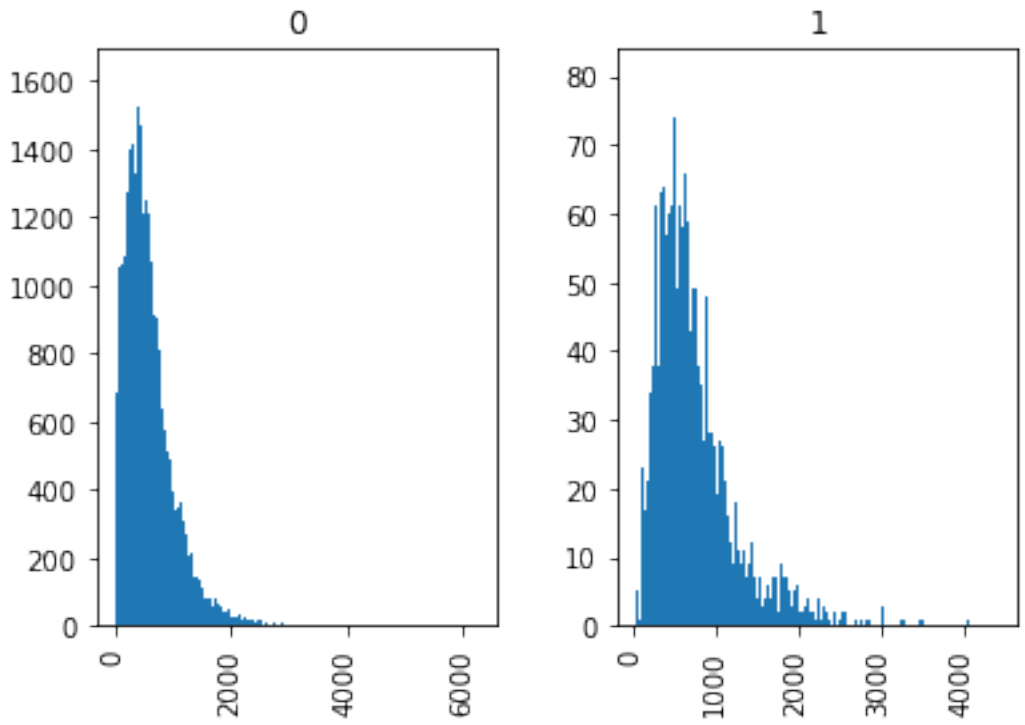
[78]:
```
#Seafood vs non seafood meals
df.groupby('seafood_meal')['DR1IKCAL'].describe()
```

[78]:

seafood_meal	count	mean	std	min	25%	50%	75%	max
0	95782.0	595.879560	442.124805	1.0	289.0	499.0	789.0	6264.0
1	5949.0	752.167759	477.014340	40.0	431.0	643.0	943.0	4451.0

[79]:
```
#Seafood vs non seafood meals
df['DR1IKCAL'].hist(bins=500, by=df['seafood_meal'])
```

[79]:
```
array([<matplotlib.axes._subplots.AxesSubplot object at 0x7feacebd8eb0>,
       <matplotlib.axes._subplots.AxesSubplot object at 0x7feacdc885b0>],
      dtype=object)
```



[81]:
```
#Seafood vs non seafood home and out
df.groupby(['seafood_meal','eathome'])['DR1IKCAL'].describe()
```

[81]:

seafood_meal	eathome	count	mean	std	min	25%	50%	75%	max
0	0	30638.0	675.534695	470.570282	1.0	352.00	576.0	885.00	5957.0
	1	65144.0	558.416800	422.941684	1.0	265.00	463.5	743.00	6264.0
1	0	2192.0	855.416971	527.356298	50.0	495.75	736.0	1073.25	4451.0
	1	3757.0	691.927602	433.886534	40.0	403.00	599.0	860.00	3509.0

Section 1: Meal Energy Grouping

Create meal energy grouping based on quantiles from the 'KCAL' variable.

[99]:
```
#Create meal energy category based on quantiles from KCAL
df.loc[df['DR1IKCAL'] < df['DR1IKCAL'].describe()['25%'], 'meal_energy'] = "Low"
df.loc[(df['DR1IKCAL'] > df['DR1IKCAL'].describe()['25%'])
```

```
       & (df['DR1IKCAL'] < df['DR1IKCAL'].describe()['50%']), 'meal_energy'] =␣
 ↪"Medium-Low"
df.loc[(df['DR1IKCAL'] > df['DR1IKCAL'].describe()['50%'])
       & (df['DR1IKCAL'] < df['DR1IKCAL'].describe()['75%']), 'meal_energy'] =␣
 ↪"Medium-High"
df.loc[df['DR1IKCAL'] > df['DR1IKCAL'].describe()['75%'], 'meal_energy'] =␣
 ↪"High"

#Display top of new meal energy category
df[['SEQN', 'DR1IKCAL','meal_energy']].head(20)
```

[99]:

	SEQN	DR1IKCAL	meal_energy
0	31127	447.0	Medium-Low
1	31127	264.0	Low
2	31128	861.0	High
3	31129	867.0	High
4	31129	1150.0	High
5	31131	253.0	Low
6	31131	88.0	Low
7	31132	130.0	Low
8	31132	836.0	High
9	31132	491.0	Medium-Low
10	31133	843.0	High
11	31133	448.0	Medium-Low
12	31134	963.0	High
13	31134	1181.0	High
14	31135	332.0	Medium-Low
15	31135	501.0	Medium-Low
16	31137	720.0	Medium-High
17	31137	813.0	High
18	31138	314.0	Medium-Low
19	31138	204.0	Low

Section 1: Questions

1. Why are there so many meals with 0 KCAL?
2. Why are the KCAL values so large? Looking at the statistics, these seem more like calories instead of kilo-calories.
3. Is the grouping method for this variable adequate?

[ ]: