Physical Activities and BMI

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Hypothesis

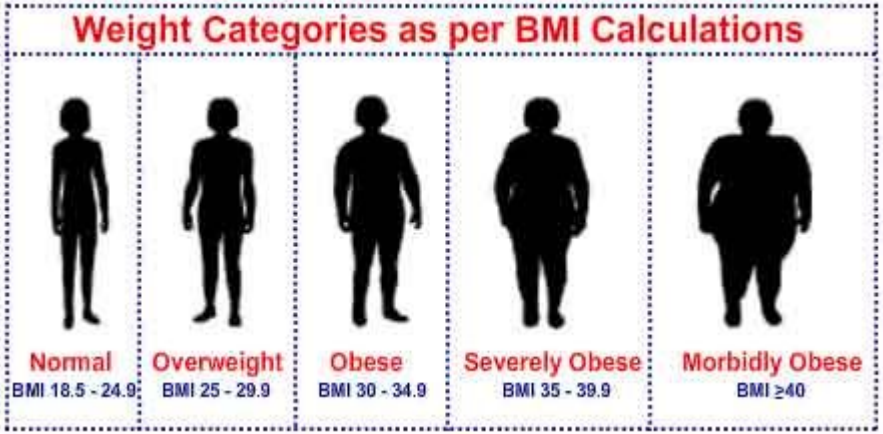
- Low physical activities level is correlated to higher BMI

National Health and Nutrition Examination Survey (NHANES)

- Identification Code 1 – 6482 ID
- Gender 0 = Male, 1 = Female GENDER
- Age at Screening Years AGE
- Marital Status 1 = Married, 2 = Widowed, 3 = Divorced, 4= Separated, 5 = Never Married, 6 = Living Together MARSTAT
- Statistical Weight 4084.478 – 153810.3 SAMPLEWT
- Pseudo-PSU 1, 2 PSU
- Pseudo-Stratum 1 – 15 STRATA
- Total Cholesterol mg/dL TCHOL
- HDL-Cholesterol mg/dL HDL
- Systolic Blood Pressure mm Hg SYSBP
- Diastolic Blood Pressure mm Hg DBP
- Weight kg WT
- Standing Height cm HT
- Body mass Index Kg/m^2 BMI
- Vigorous Work Activity 0 = Yes, 1 = No VIGWRK
- Moderate Work Activity 0 = Yes, 1 = No MODWRK
- Walk or Bicycle 0 = Yes, 1 = No WLKBIK
- Vigorous Recreational Activities 0 = Yes, 1 = No VIGRECEXR
- Moderate Recreational Activities 0 = Yes, 1 = No MODRECEXR
- Minutes of Sedentary Activity per Week Minutes SEDMIN
- BMI>35 0 = No, 1 = Yes OBESE

6,482 X 21

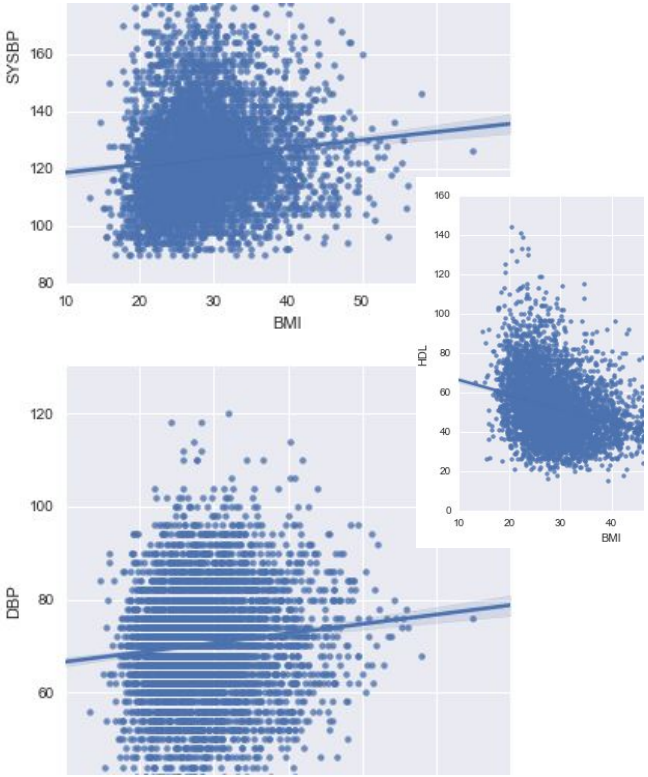
Background

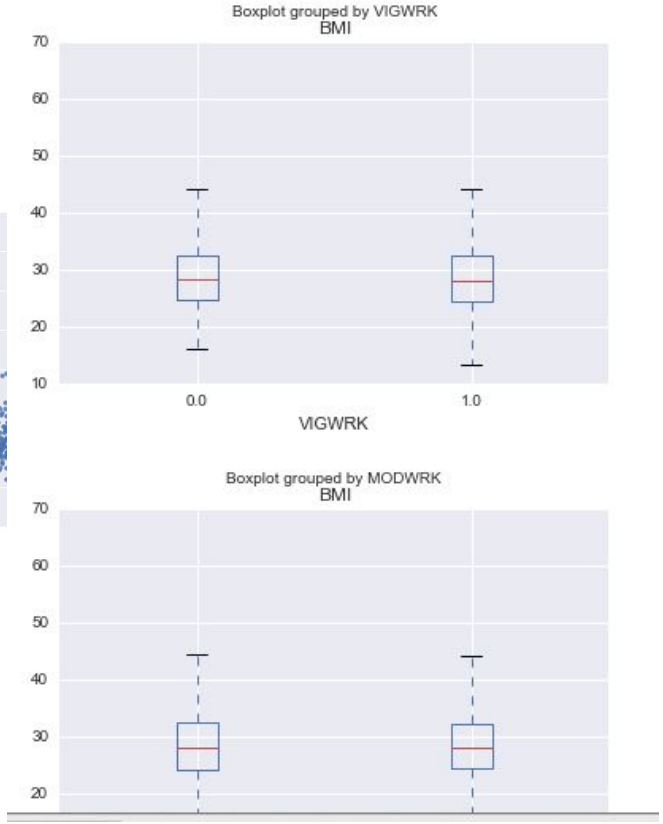


National Cholesterol Education Program Cholesterol Guidelines			
	Desirable	Borderline High	High
Total Cholesterol	Less than 200	200 - 239	240 and higher
LDL Cholesterol (the "bad" cholesterol)	Less than 130	130 - 159	160 and higher
HDL Cholesterol (the "good" cholesterol)	50 and higher	40 - 49	Less than 40
		200 - 399	400 and higher

Exploratory Data Analysis

ID	3252.092777
AGE	49.272431
TCHOL	195.740793
HDL	52.754222
SYSBP	123.982909
DBP	70.440285
WT	81.192228
HT	167.461994
BMI	28.876973
VIGWRK	0.811394
MODWRK	0.641302
WLKBIK	0.739980
VIGRECEXR	0.811801
MODRECEXR	0.609969
SEDMIN	311.312716
Male_0	0.493591
Married_1.0	0.522686
Widowed_2.0	0.081180
Divorced_3.0	0.109257
Separated_4.0	0.032553
Cohab_6.0	0.083215
Obese_1.0	0.156256
dtype: float64	





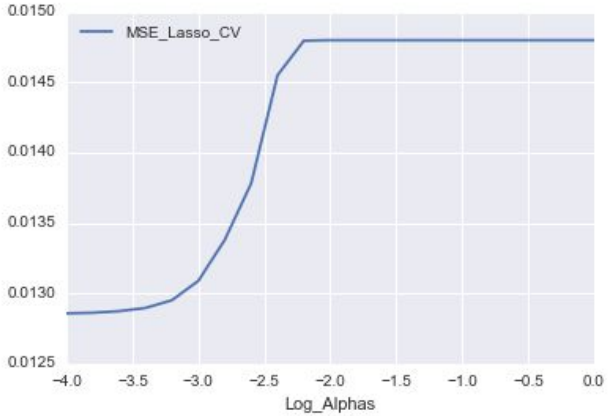
Models

- Lasso Regression
- Logistic Regression
- Voting Classifier

Lasso Regression

```
Alpha: 0.0158489319246
Alpha: 0.0251188643151
Alpha: 0.0398107170553
Alpha: 0.063095734448
Alpha: 0.1
Alpha: 0.158489319246
Alpha: 0.251188643151
Alpha: 0.398107170553
Alpha: 0.63095734448
Alpha: 1.0

: <matplotlib.axes._subplots.AxesSubplot at 0x1146e50
```



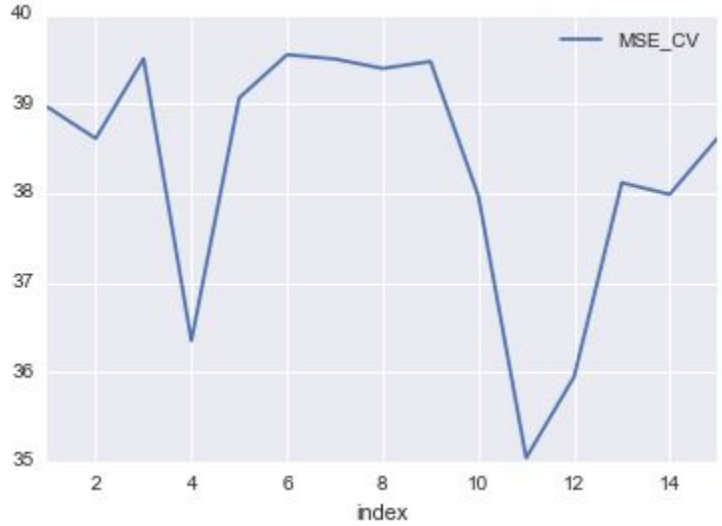

```
[(-0.2200296031416358, 'HDL'),
 (-0.026167275204777921, 'Male_0'),
 (-0.019315969322357344, 'VIGRECEXR_0.0'),
 (-0.010879095196041985, 'WLKBIK_0.0'),
 (-0.0088854602022295139, 'MODRECEXR_0.0'),
 (-0.0041773552881292572, 'Cohab_6.0'),
 (0.0, 'Divorced_3.0'),
 (0.0, 'MODWRK_0.0'),
 (0.0, 'Separated_4.0'),
 (0.0, 'TCHOL'),
 (-0.0, 'Widowed_2.0'),
 (0.0014050089281379338, 'Married_1.0'),
 (0.0072284598477105637, 'VIGWRK_0.0'),
 (0.011218753397490447, 'AGE'),
 (0.011483633726529113, 'SYSBP'),
 (0.018146112597906335, 'SEDMIN'),
 (0.039590879454649383, 'DBP')]
```

```
y_pred = lm.predict(Xr_test)
lm.score(Xr_train, yr_train)
```

```
0.12080539995504114
```

Lasso Regression

X11 = NewNew[['TCHOL','HDL','SYSBP','DBP','Male_0']]



Logistic Regression


```
[(-4.7644018771035732, 'WT'),
 (-1.046330520788491, 'Male_0'),
 (-0.43771431064072774, 'VIGWRK_0.0'),
 (-0.35646819311427141, 'MODWRK_0.0'),
 (-0.33074783258766827, 'Widowed_2.0'),
 (-0.240631432765681O8, 'Cohab_6.0'),
 (-0.13016768120913585, 'AGE'),
 (-0.041599806743366906, 'DBP'),
 (-0.035806070236703194, 'BMI'),
 (-0.003438941490913716, 'SYSBP'),
 (0.0, 'SEDMIN'),
 (0.071729842861943824, 'Separated_4.0'),
 (0.24295920652833647, 'Married_1.0'),
 (0.3107381604924988, 'Divorced_3.0'),
 (0.67975372666205502, 'HT'),
 (0.86262779862585004, 'TCHOL'),
 (1.227408433097136, 'HDL')]
```

```
from sklearn.metrics import confusion_matrix
y_hat = lm.predict(Xc_train)
confusion_matrix(y_hat,yc_train)
```

```
array([[3723,  674],
       [  13,   13]])
```

```
lm.score(Xc_train, yc_tr
```

```
0.84467555957494911
```

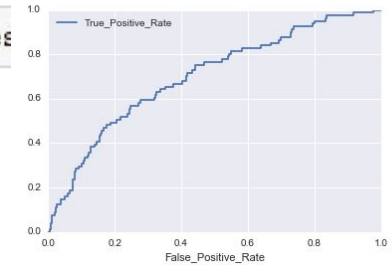
```
lm.score(Xc_test, yc_tes
```

```
0.82723577235772361
```

```
0.699978973296
```

```
Roc_DataFrame = pd.DataFrame({'False_Positive_Rate'
Roc_DataFrame.plot(x = 'False_Positive_Rate' , y =
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x11648a8
```




```
from sklearn.cross_validation import cross_val_score
print(1 - cross_val_score(lm,Xc_train,yc_train,cv=10).mean())

# misclassification error around 15.6%
```

```
0.156228580452
```

Ensemble Method

74]:
```
from sklearn.grid_search import GridSearchCV
clf1 = LogisticRegression()
clf2 = RandomForestClassifier()
clf3 = BernoulliNB()
eclf = VotingClassifier(estimators=[('lr', clf1), ('rf', clf2), ('bnb', clf3)],voting='hard')
params = {'lr__C': [.01,.1,1,10],
          'rf__n_estimators':[1000],
          'rf__max_depth':[2,5,10],
          'bnb__alpha':[0.1,0.5,1]}
grid = GridSearchCV(estimator=eclf, param_grid=params, cv=2)
gridfit = grid.fit(Xc_train, yc_train)
```

76]:
```
print gridfit.best_params_
```

```
{'bnb__alpha': 0.1, 'rf__max_depth': 2, 'rf__n_estimators': 1000, 'lr__C': 0.01}
```

Top 10 Reasons Why The BMI Is Bogus

July 4, 2009 · 8:00 AM ET
Heard on Weekend Edition Saturday

KEITH DEVLIN

Americans keep putting on the pounds — at least according to a report released this week from the Trust for America's Health. The study found that nearly two-thirds of states now have adult obesity rates above 25 percent.

But you may want to take those findings — and your next meal — with a grain of salt, because they're based on a calculation called the body mass index, or BMI.

As the *Weekend Edition* math guy, I spoke to Scott Simon and told him the body mass index fails on 10 grounds:

The BMI Formula

BMI = weight in pounds/(height in inches x height in inches) x 703

The 703 is to convert the index fro the original metric version of the formula.

Next Steps

- Add additional features/ data
- Find different measures of obesity and compare models

Thank you!