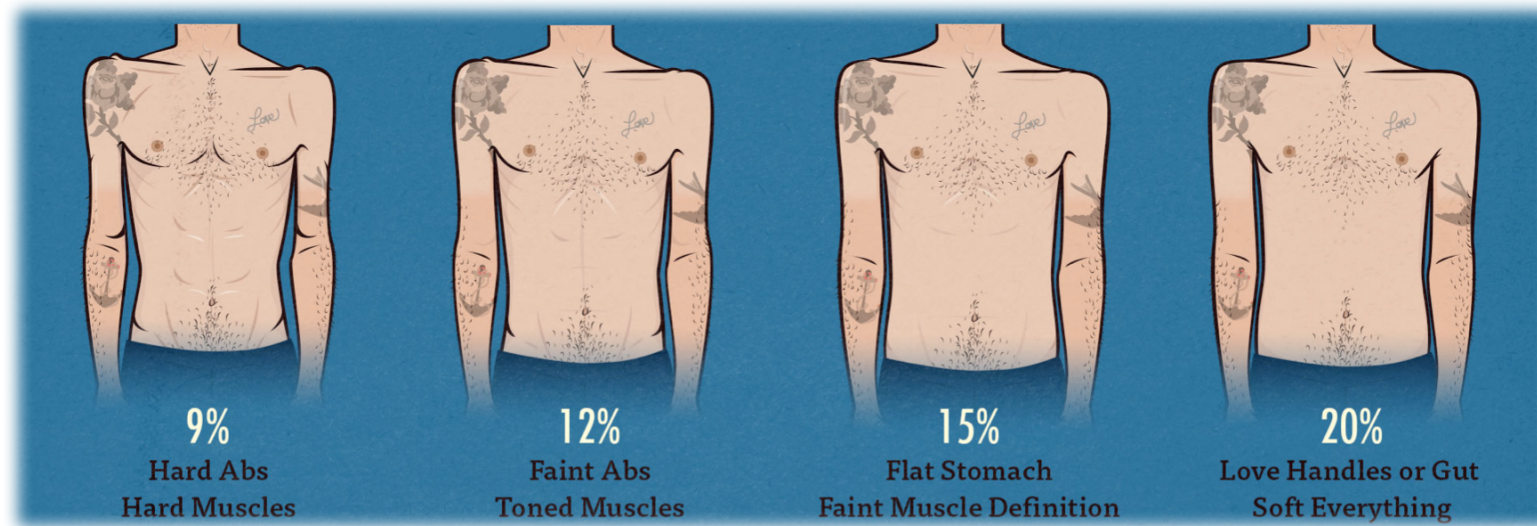


Goup4 Presentation

Body Fat Project



Summary of Data Cleaning

- We transformed the units of **NECK, CHEST, ABDOMEN, HIP, THIGH, KNEE, ANKLE, BICEPS, FOREARM** and **WRIST** from centimeters to inches to unify with the unit of **HEIGHT** and the custom of the US.
- We imputed **individual IDNO 42's HEIGHT**

Reasons:

The individual IDNO 42 has a height of 29.5 inches (about 75 cm) which is abnormal for an adult.

IDNO	Original HEIGHT	Imputed HEIGHT	Imputation Method
42	29.5 inches	69.5 inches	Using ABDOMEN(BMI) and WEIGHT to calculate to true value for HEIGHT

Summary of Data Cleaning

- We deleted **individual IDNO 172** and **individual IDNO 182**

Reasons:

The individual IDNO 172 has a **BODYFAT** of 0 and individual IDNO 182 has a **BODYFAT** of 1.9, which is below the normal range of human being. We tried using the formula $bodyfat = \frac{495}{Density} - 450$ to calculate the imputed value of **BODYFAT** for these two individuals but got smaller even negative values. So we just deleted them.

IDNO	Original BODYFAT	Imputed BODYFAT
172	1.9	0.697
182	0	-3.61

Summary of Data Cleaning

- Final Cleaned Data: **n=250** (from n=252) with p = 11 predictors

Predictor	Unit
WEIGHT	pound
HEIGHT	inch
ADIPOSIY	kg/m ²
NECK	inch
CHEST	inch
ABDOMEN	inch
HIP	inch
THIGH	inch
KNEE	inch
BICEPS	inch
WRIST	inch

Metric for Model Performance

- We define the desired model based on the following criteria:
 1. Prefer easy to measure predictors for users
 2. Prefer not confusing/clearly specified predictors for users
 3. Prefer using less predictors than more predictors
 4. Using *R-Squared* and *adjusted R-Squared* for evaluating models

Finding Desired Model for Bodyfat

- Decision making:

1. Try linear models with different dependent variables:

BODYFAT vs DENSITY

2. After decided using BODYFAT as the dependent variable, try different simple linear regression models:

BODYFAT ~ ABDOMEN

BODYFAT ~ ADIPOSITY

BODYFAT ~ CHEST

BODYFAT ~ HIP

BODYFAT ~ WEIGHT

... ..

Finding Desired Model for Bodyfat

- Decision making:

3. Starting from $BODYFAT \sim ABDOMEN$ we select a new variable to be added to the SLR model:

$BODYFAT \sim ABDOMEN + ADIPOSITY$

$BODYFAT \sim ABDOMEN + CHEST$

$BODYFAT \sim ABDOMEN + HIP$

$BODYFAT \sim ABDOMEN + WEIGHT$

$BODYFAT \sim ABDOMEN + HEIGHT$

... ..

Finding Desired Model for Bodyfat

- Decision making:

4. Then we decide to add the interaction terms considering the high correlation between variables:

BODYFAT ~ ABDOMEN + ADIPOSITIVITY + ABDOMEN ADIPOSITIVITY*

*BODYFAT ~ ABDOMEN + CHEST + ABDOMEN*CHEST*

*BODYFAT ~ ABDOMEN + HIP + ABDOMEN*HIP*

*BODYFAT ~ ABDOMEN + WEIGHT + ABDOMEN*WEIGHT*

*BODYFAT ~ ABDOMEN + HEIGHT + ABDOMEN*HEIGHT*

... ..

5. Ended using **BODYFAT ~ ABDOMEN + WEIGHT + ABDOMEN*WEIGHT**

Results

Model	R^2	Adjusted R^2
<i>BODYFAT ~ ABDOMEN</i>	0.6522	0.6508
<i>BODYFAT ~ ADIPOSITY</i>	0.5193	0.5174
<i>BODYFAT ~ CHEST</i>	0.4808	0.4787
<i>BODYFAT ~ HIP</i>	0.3765	0.3739
<i>BODYFAT ~ WEIGHT</i>	0.3588	0.3563
... ..		

Results

Model	R^2	Adjusted R^2	Correlations between predictors
<i>BODYFAT ~ ABDOMEN + ADIPOSITY</i>	0.656	0.6532	0.922
<i>BODYFAT ~ ABDOMEN + CHEST</i>	0.6643	0.6616	0.914
<i>BODYFAT ~ ABDOMEN + HIP</i>	0.6853	0.6827	0.871
<i>BODYFAT ~ ABDOMEN + WEIGHT</i>	0.7135	0.7111	0.885
<i>BODYFAT ~ ABDOMEN + HEIGHT</i>	0.69	0.6875	0.174
... ..			

Results

Model	R^2	Adjusted R^2
<i>BODYFAT ~ ABDOMEN + ADIPOSITY + ABDOMEN* ADIPOSITY</i>	0.671	0.667
<i>BODYFAT ~ ABDOMEN + CHEST + ABDOMEN*CHEST</i>	0.6795	0.6756
<i>BODYFAT ~ ABDOMEN + HIP + ABDOMEN*HIP</i>	0.6948	0.6911
<i>BODYFAT ~ ABDOMEN + WEIGHT + ABDOMEN*WEIGHT</i>	0.7235	0.7201
<i>BODYFAT ~ ABDOMEN + HEIGHT + ABDOMEN*HEIGHT</i>	0.6949	0.6912
... ..		

Discussion of Results

- Reasons for building model starting from ABDOMEN:
 1. In univariate models, BODAYFAT ~ ABDOMEN model gives the highest R-Squared and Adjusted R-Squared.
 2. For models using two predictors(with or without the interaction term), the formulas using ABDOMEN always perform better under the criteria of R-Squared and Adjusted R-Squared.
- Reasons for using interaction terms:
 1. ABDOMEN and WEIGHT are two highly correlated variables.
 2. This interaction term is significantly important at significance level 0.01.

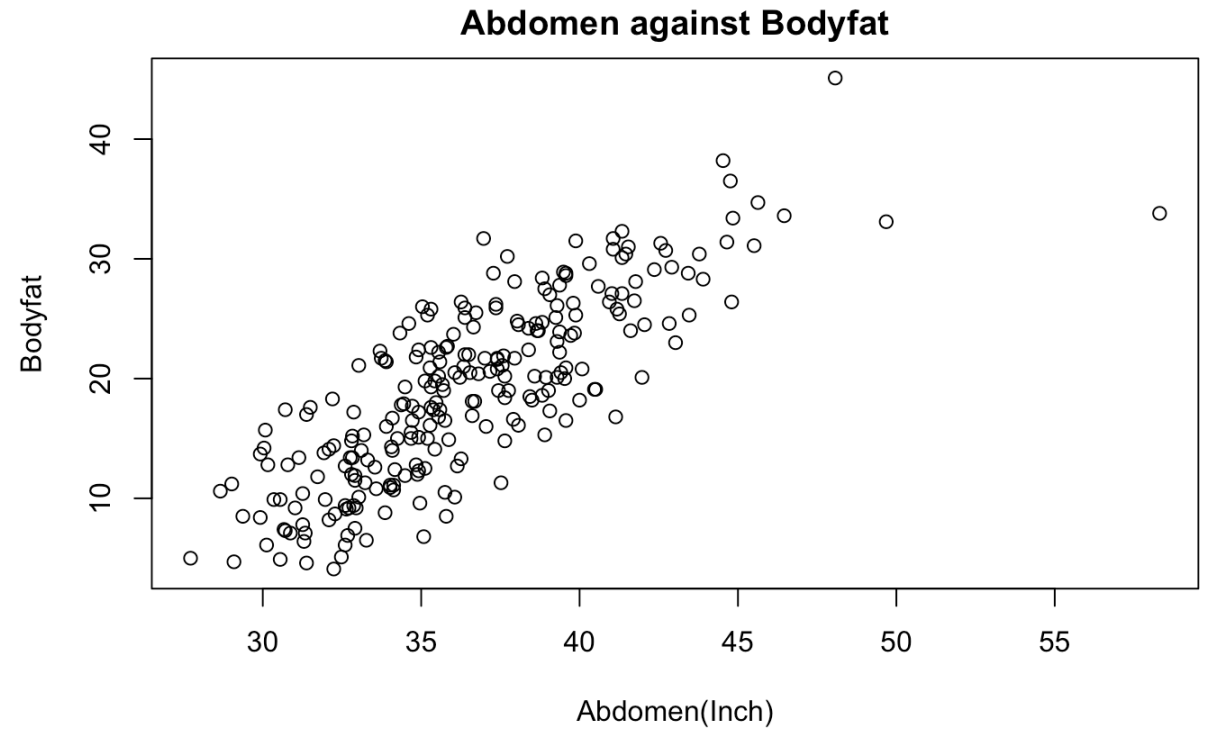
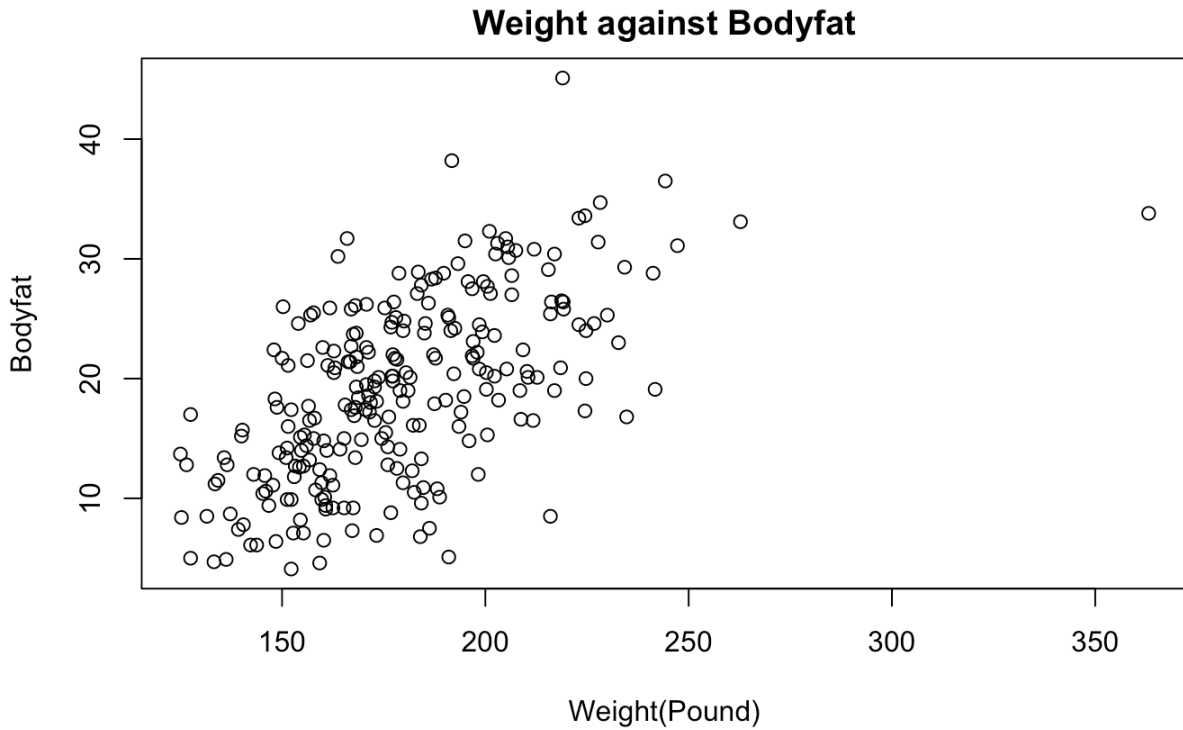
Final Model:

$$\text{BODYFAT} = -62.51 + 2.86 * \text{ABDOMEN} - 0.014 * \text{WEIGHT} - 0.003 * \text{ABDOMEN} * \text{WEIGHT}$$

Explanation:

- As men's abdomen circumference get bigger by 1 inch, he is expected to gain about 2.9% in body fat while the actual increase depends on his weight.
- The increase of a man's weight does not tell us whether he is gaining muscle or fat. This explains why the coefficient for weight is negative and not significant.
- If a man gets heavier with the unchanged abdomen circumference, he is more likely to gaining muscle because muscle has higher density than fat. This is why the coefficient for the interaction term is negative.

Visualization of Final Model



Visualization of Final Model



Statistical Properties of Final Model

1. Coefficient for ABDOMEN is significant at significance level 0.001 based on two-sided t-test with p-value less than 2×10^{-16} .
2. Coefficient for the interaction term of ABDOMEN and WEIGHT is significant at significance level 0.001 based on two-sided t-test with p-value equal to 0.00313.
3. Coefficient of WEIGHT is negative and NOT significant at 0.1.
4. Overall model is significant at 0.05 based on F-test with p-value less than 2.2×10^{-16} .
5. The final model has an adjusted R-Squared equal to 0.72, making this model reliable and trustworthy.

Strengths and Weaknesses

Final Model:

$$\text{BODYFAT} = -62.51 + 2.86 * \text{ABDOMEN} - 0.014 * \text{WEIGHT} - 0.003 * \text{ABDOMEN} * \text{WEIGHT}$$

- **Strengths**

1. Simple: Easy to use for users, easy to explain for statisticians.
2. Explains 72% of variation in body fat.

- **Weaknesses**

1. Our model does not treat special points(leverage points or outliers) specifically. For extremely obese people or skinny people, we may not be able to estimate their body fat percentage accurately.

Thank you!