BodyFat Analysis

Shiwei Cao, Shurong Gu, Yuwei Sun

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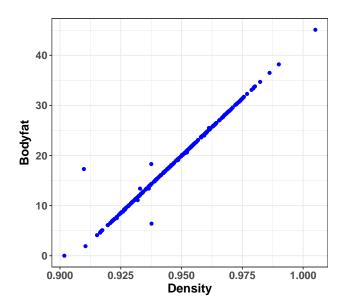
Procedure

- Data preprocessing: EDA, data cleaning
- Statistical Modeling:
 - Multiple Linear Regression
 - Variable Selection: stepwise method
 - Model Diagnosis
 - Lasso
- Model selection:
 - Make predictions on validation set and choose the best model (the one with smallest mse)
- ► Model Interpretation, "rule of thumb"
- ► Strengths and weaknesses

Explore the data

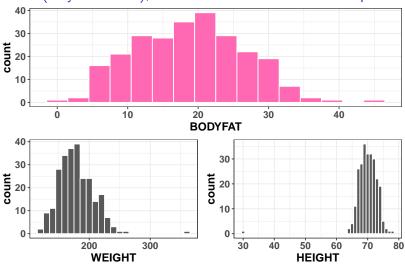
IDNO	BODY	FAT I	DENSITY	AGE	WEIGH	T HEIGH	T ADIPOSITY	NECK	CHEST	ABDO	MEN	HIP	THIGH	KNEE	ANKLE	BICEP	S FOR	EARM	WRIS	т
1	1	12.6	1.0708	23	154.2	5 67.7	5 23.	36.2	93.1		85.2	94.5	59.0	37.3	21.9	32	.0	27.4	17.	1
2		6.9	1.0853	22	173.2	5 72.2	5 23.4	38.5	93.6		83.0	98.7	58.7	37.3	23.4	30	.5	28.9	18.	2
3	2	24.6	1.0414	22	154.0	0 66.2	5 24.	34.0	95.8		87.9	99.2	59.6	38.9	24.0	28	.8	25.2	16.	6
4	1	10.9	1.0751	26	184.7	5 72.2	5 24.5	37.4	101.8		86.4	101.2	60.1	37.3	22.8	32	.4	29.4	18.	2
5	2	27.8	1.0340	24	184.2	5 71.2	5 25.0	34.4	97.3		100.0	101.9	63.2	42.2	24.0	32	.2	27.7	17.	7
6	2	20.6	1.0502	24	210.2	5 74.7	5 26.5	39.0	104.5		94.4	107.8	66.0	42.0	25.6	35	.7	30.6	18.	8
	DNO I	BODY	FAT DEN	SITY	AGE V	EIGHT F	EIGHT ADIP	OSITY I	NECK C	HEST	ABDO	MEN	HIP T	HIGH I	KNEE A	NKLE	BICEPS	FORE	ARM	WR
247	247			.0308	AGE V	215.50	70.50	30.5		113.7			HIP TI	63.3	44.0	22.6	37.5	FORE	32.6	WR 1
		2	29.1 1.								1	07.6						FORE		
247	247	2	29.1 1. 11.5 1.	.0308	69	215.50	70.50	30.5	40.8 34.9	113.7	1	07.6	110.0	63.3	44.0	22.6	37.5	FORE	32.6	1
247 248	247 248	1 3	29.1 1. 11.5 1. 32.3 1.	0308	69 70	215.50 134.25	70.50 67.00	30.5 21.1	40.8 34.9 40.9	113.7 89.2	1	07.6 83.6 05.0	110.0	63.3 49.6	44.0 34.8	22.6 21.5	37.5 25.6	FORE	32.6 25.7	1
247 248 249	247 248 249	2 1 3	29.1 1. 11.5 1. 32.3 1. 28.3 1.	0308 0736 0236	69 70 72	215.50 134.25 201.00	70.50 67.00 69.75	30.5 21.1 29.1	40.8 34.9 40.9 38.9	113.7 89.2 108.5	1 1	07.6 83.6 05.0	110.0 88.8 104.5	63.3 49.6 59.6	44.0 34.8 40.8	22.6 21.5 23.2	37.5 25.6 35.2		32.6 25.7 28.6	1

Data Cleaning



We decide to remove 3 points for further analysis:

The one with BODYFAT=0; the one with HEIGHT=29.5 inches (only 75cm tall); the one with WEIGHT=363.15 pounds



Variable Selection

- Divide the whole data into train (80%) and validation (20%) set
- Train set: Use several different methods to choose the best subset of variables
- 1. Mallow's Cp: leaps() in R (do an exhaustive search)
- 2. Stepwise regression based on AIC
- 3. Stepwise regression based on BIC
- ▶ Validation set: See if the models generalize well on unseen data

```
## Model MSE
## 1 Mallow's Cp 15.069
## 2 AIC 14.702
## 3 BIC 13.989
```

BIC results

Start:

```
> model.BIC <- step(model, k=log(249))
Start: AIC=624.5
BODYFAT ~ AGE + WEIGHT + HEIGHT + ADIPOSITY + NECK + CHEST +
ABDOMEN + HIP + THIGH + KNEE + ANKLE + BICEPS + FOREARM +
WRIST</pre>
```

```
Df Sum of Sa
                            RSS
- KNEE
                    0.16 3002.1 618.99
                    6.13 3008.1 619.39
- BICEPS
- NECK
                    6.38 3008.3 619.41
- CHEST
                   11.80 3013.7 619.77
                   15.46 3017.4 620.01
- HIP
- ANKLE
                   20.35 3022.3 620.33
- FOREARM
                   20.53 3022.4 620.35
                   25.75 3027.7 620.69
- THIGH
- HEIGHT
                   32.21 3034.1 621.12
                   38.57 3040.5 621.54
- AGE
- ADIPOSITY
                   42.18 3044.1 621.77
                   47.07 3049.0 622.09
- WEIGHT
<none>
                          3001.9 624.50
- WRIST
                  121.38 3123.3 626.91
- ABDOMEN
                 1117.80 4119.7 682.29
```

End:

```
Step: AIC=576.93

BODYFAT ~ WEIGHT + ABDOMEN + WRIST

Df Sum of Sq RSS AIC

- ONDES 3205.4 576.93

- WRIST 1 121.7 3327.1 578.86

- WEIGHT 1 138.9 3344.3 579.89

- ABDOMEN 1 3348.7 6554.1 714.46
```

Variables selected by BIC:

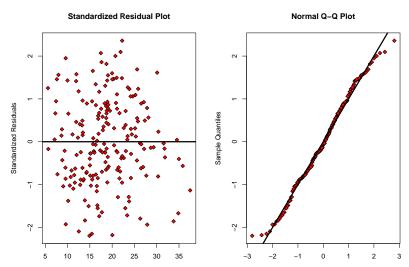
At this point, we come up with a multiple linear regression:

► Bodyfat ~ Abdomen + Wrist + Weight



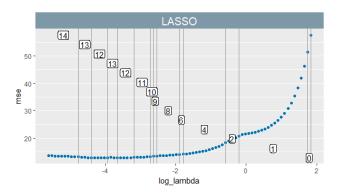
Model Diagnosis Adjusted R-squared for the final model: 0.7175

Linear regression assumptions?



Another approach: Lasso

$$\min_{\beta} \sum_{i=1}^{N} (y_i - \beta_0 - \beta_1 x_i - \dots - \beta_p x_p)^2 + \lambda \left| \sum_{j=1}^{p} \beta_j \right|$$



Conclusion

Our proposed linear model to predict body fat %:

```
(BodyFat %)= -23.794 + 0.852 \times Abdomen - 1.258 \times Wrist - 0.073 \times Weight
```

- Possible rule of thumb:
 - ► Your % Bodyfat =
 - ► Your *abdomen* circumference (cm) ×0.9
 - ightharpoonup minus *wrist* circumference (cm) imes 1.3
 - ightharpoonup minus weight (lbs) imes 0.1
 - minus 24
- ► For a normal graduate male student, with circumferences: Abdomen=85cm, Wrist=18cm, Weight=130lbs, his predicted body fat percentage would be around 16.43%. There is a 95% probability that his body fat is between 8.26% and 24.59%.

Strengths and Weaknesses

- Strengths
- 1. Use a separate validation set to avoid overfitting
- 2. Simple, easy to interpret
- Weaknesses
- 1. May lose information using only 200 data points
- 2. Trade off between simplicity and precision