

Logistic Regression Introduction

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Cartwheel Data

Random sample of 25 adults attempted cartwheels

Primary Variable of interest: Cartwheel completion



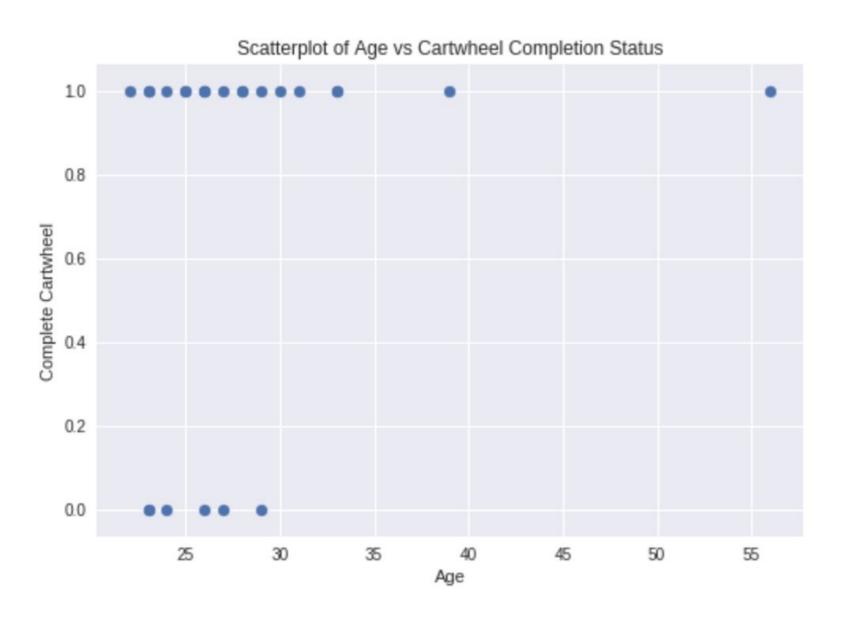


Research Question

Based on age, can we predict whether a cartwheel is completed?

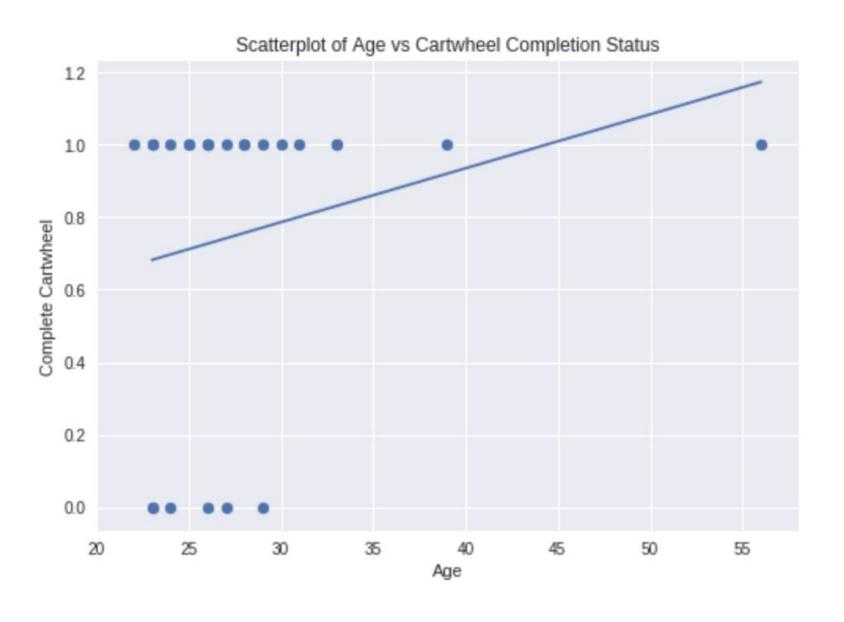


Let's Look at the Data



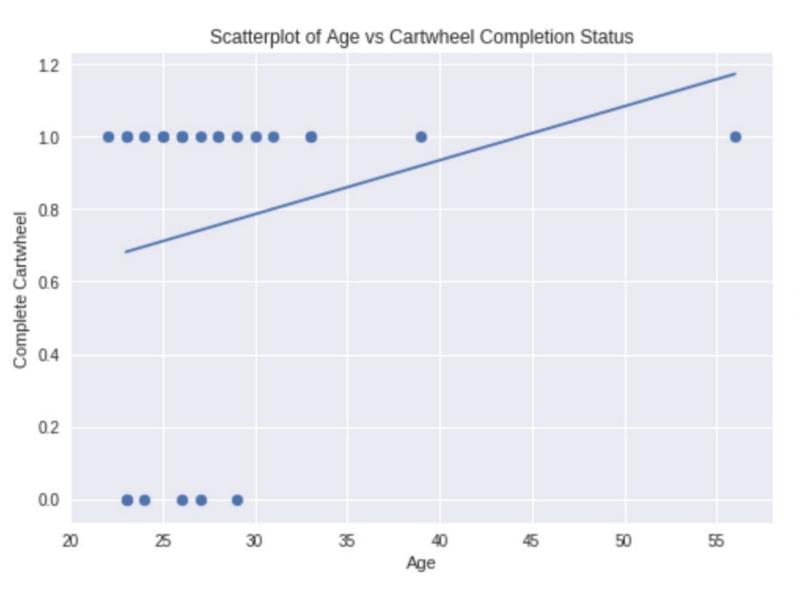


Linear Model





Linear Model



$$\hat{y} = 0.34 + 0.015$$
 age



Logit Transformation

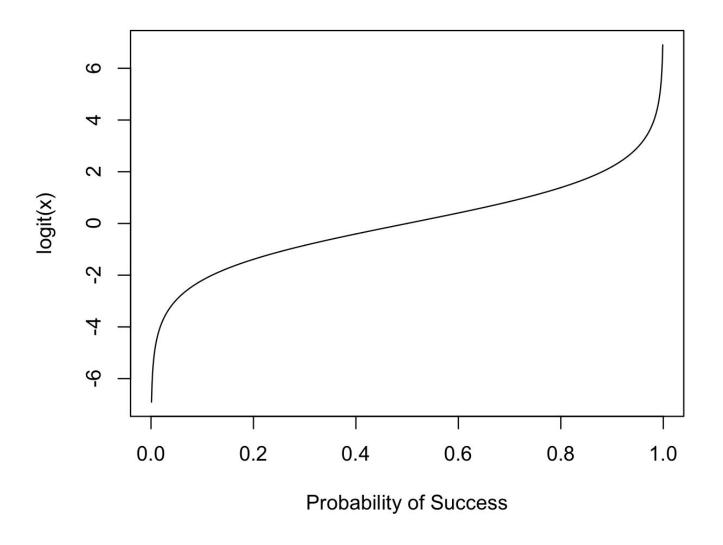
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Logit Transformation

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- Uses the logit function $\ln(\frac{P}{1-P})$

Logistic Function of Probability

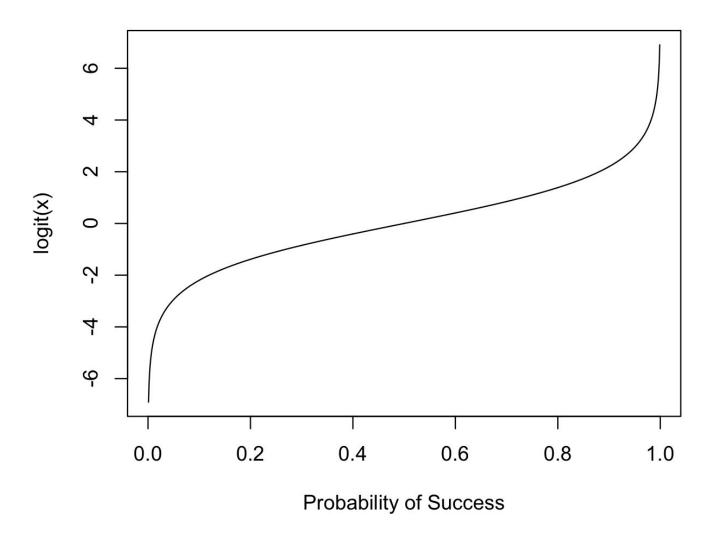




Logit Transformation

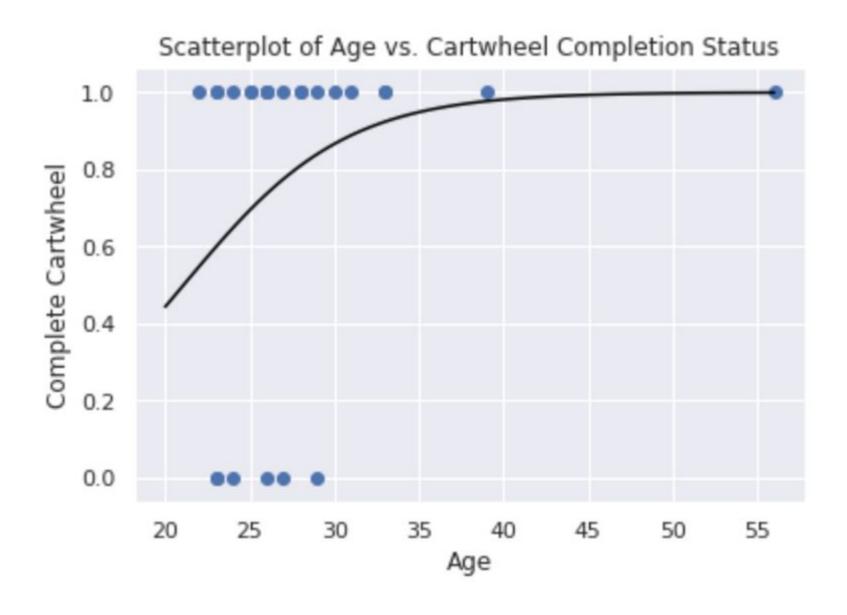
- Instead of predicting completion status, we predict a *transformed version* of the probability of a success
- Uses the logit function: $\ln(\frac{P}{1-D})$
- logit(\hat{y}) = $b_0 + b_1 x$





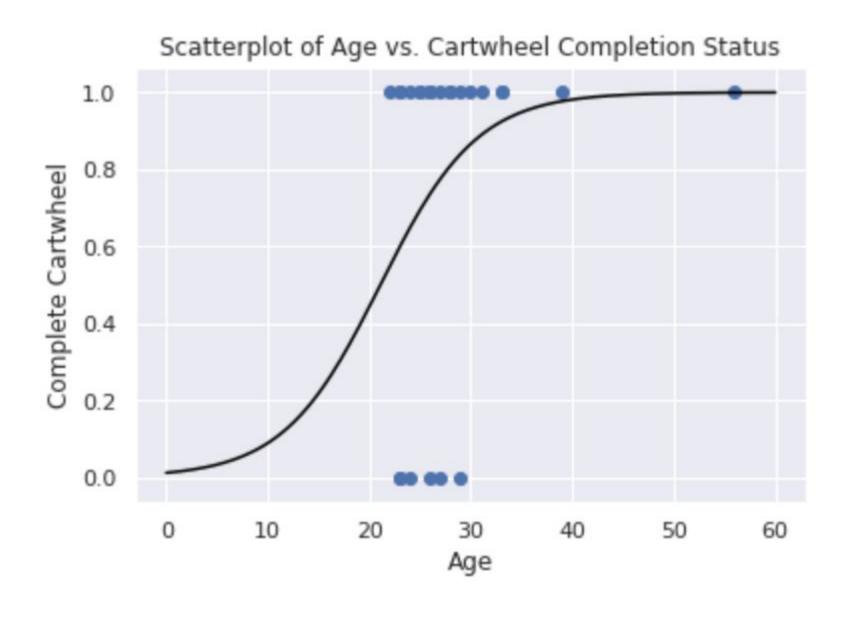


Logistic Regression Line





Logistic Regression Line





Extrapolation IVQ

Would you feel comfortable using this model to estimate the probability that a teenager who is 15 can complete a cartwheel?



Generalized Linear Model Regression Results

Dep. Variable: CompleteGroup No. Observations: 25

Model: GLM Df Residuals: 23

Model Family: Binomial Df Model: 1

Link Function: logit Scale: 1.0

coef std err z P>|z| [0.025 0.975]

Intercept -4.4213 4.429 -0.998 0.318 -13.101 4.259



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 P>lzl
 [0.025
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 Intercept -4.4213
 4.429
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 Age
 0.2096
 0.171
 1.225
 0.221
 -0.126
 0.545



 $logit(\hat{y}) = -4.42 + 0.2096$ age

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Slope interpretation:

For each increase in age by I year, the log odds of a successful cartwheel increases by about 0.2096, on average.

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 $logit(\hat{y}) = -4.42 + 0.2096$ age

Slope interpretation: For each year increase in age, the odds of a successful cartwheel increases by about 1.23 (e^{0.2096}) times that of the younger age, on average.

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Predicted Probability of Success

 For someone who is 36, what is their predicted log odds of a successful cartwheel using the model?



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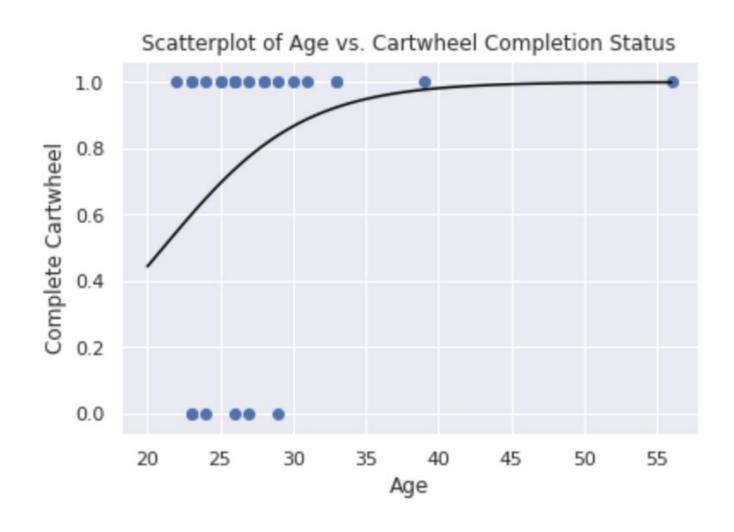
 For someone who is 36, what is their predicted log odds of a successful cartwheel using the model?

```
logit(\hat{y}) = -4.42 + 0.2096 age
= -4.42 + 0.2096 (36)
= 3.13
```



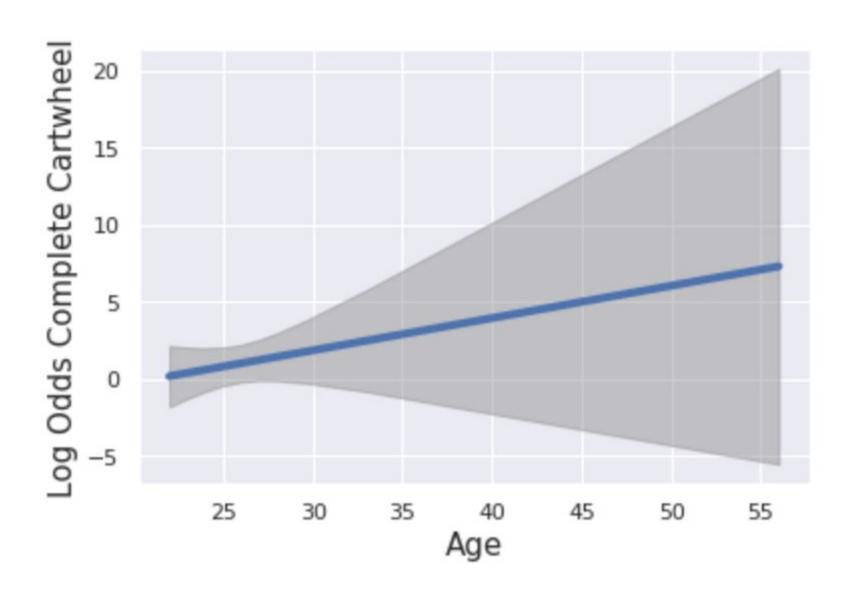
Predicted Probability of Success

- For someone who is 36, what is their predicted log odds of a successful cartwheel using the model?
- Using the graph on the right, estimate what the probability of success might be?



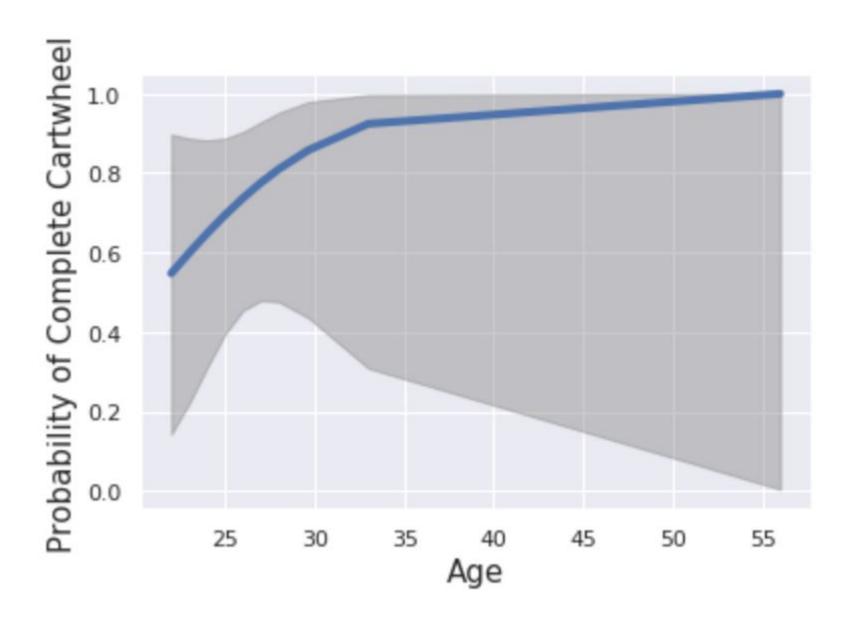


Prediction Uncertainty





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We need to assume that our model logit(y) = $\beta_0 + \beta_1 x_1$ is appropriate



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- ~with a large enough sample size, you can identify discrepancies with residual plots
- ~y only takes two values, so residuals can be limited
- ~to create informative residual plots, it helps if x takes a wide range of values and to have additional covariates