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# Analyzing time series data using GAMs

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# Day 1 • July 7

- 1) Short summary of **linear regression**
  - 2) Intro **GAMMs**
  - 3) Example of GAMM analysis: **articulography data**
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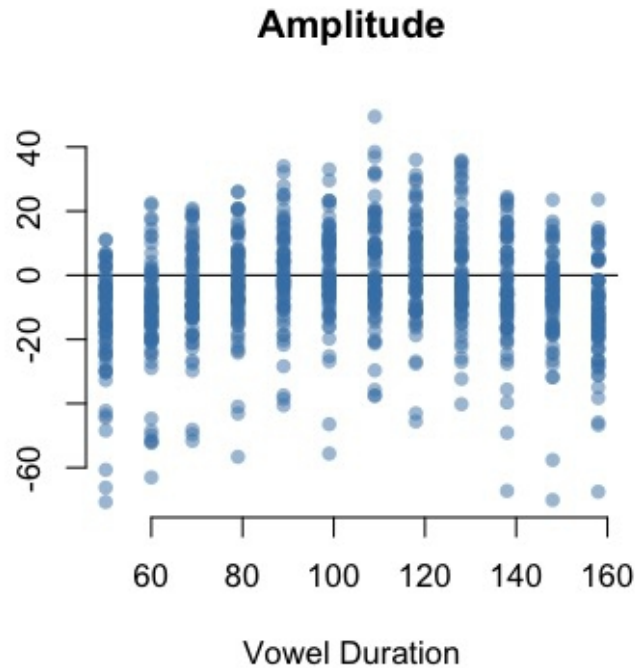
# 1. Linear regression

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# Example data

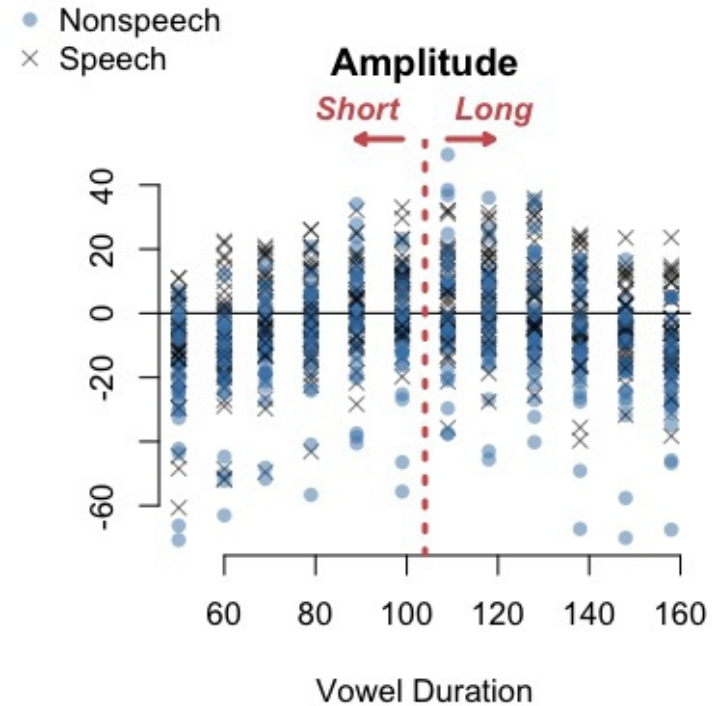
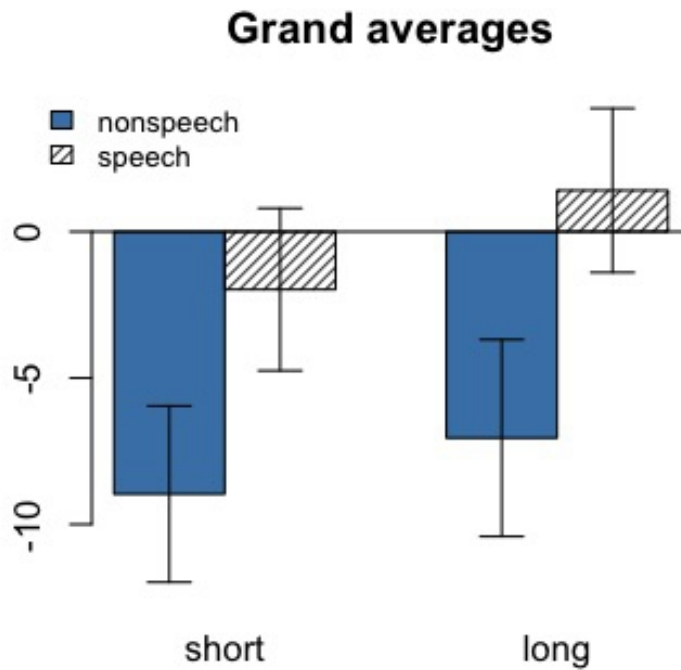
- MEG data of Tomaschek et al. (2013)
  - *Vowel duration and mode*





# ANOVA

## ■ Categorical data





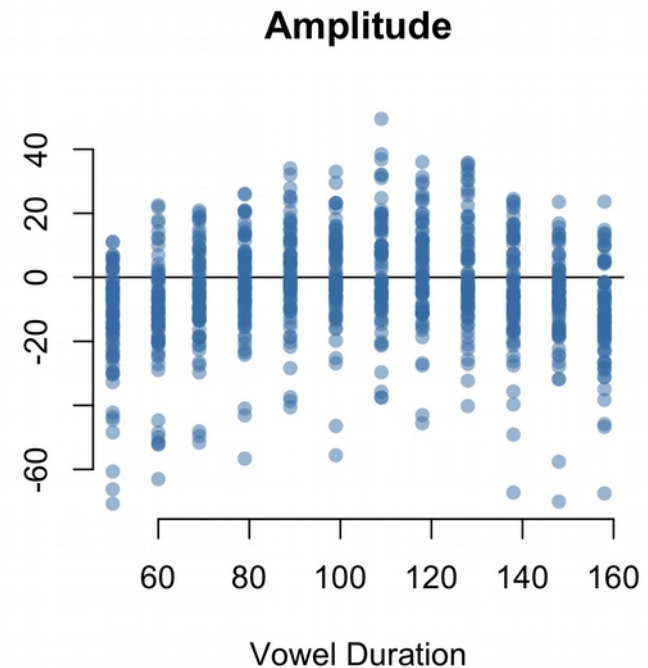
# Linear regression

- Continuous and categorical predictors

- Example 1:

Could the value of *Vowel Duration* explain the measured *Amplitude*?

- Vowel duration = numeric



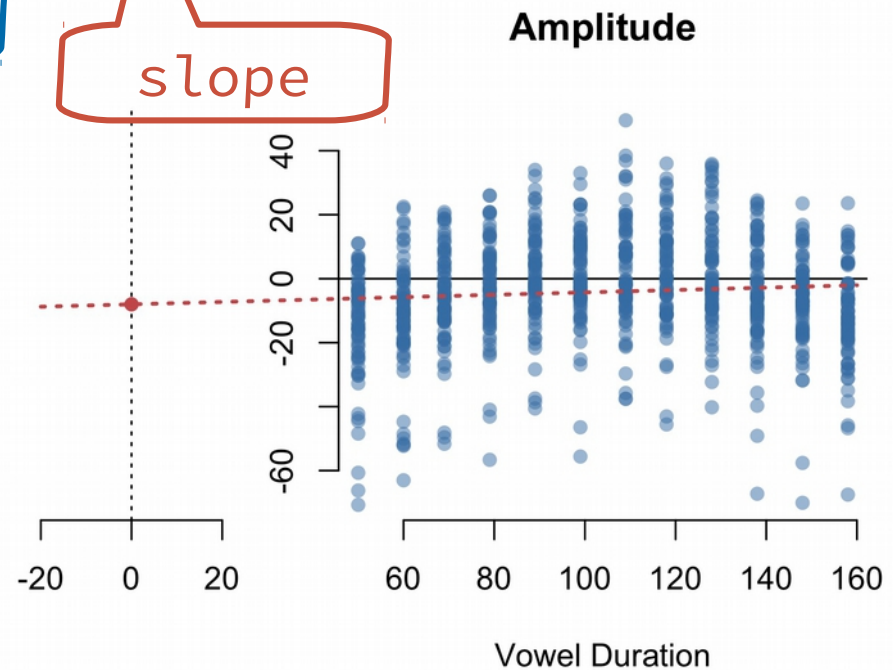


# Continuous predictors

- Regression formula:  $y = \beta_0 + \beta_{VD} * x + \epsilon$

intercept

slope





# Continuous predictors

- Regression formula:  $y = \beta_0 + \beta_{VD} * x + \epsilon$

```
lm1 <- lm(Amp ~ VD, data=dat)
summary(lm1)
```

Coefficients:

|  | Estimate | Std. Error | t value | Pr(> t ) |
|--|----------|------------|---------|----------|
|--|----------|------------|---------|----------|

|             |          |  |  |  |
|-------------|----------|--|--|--|
| (Intercept) | -8.01569 |  |  |  |
|-------------|----------|--|--|--|

|    |         |  |  |  |
|----|---------|--|--|--|
| VD | 0.03730 |  |  |  |
|----|---------|--|--|--|

1.78739

-4.485

8.24e-06

0.01638

2.277

0.023



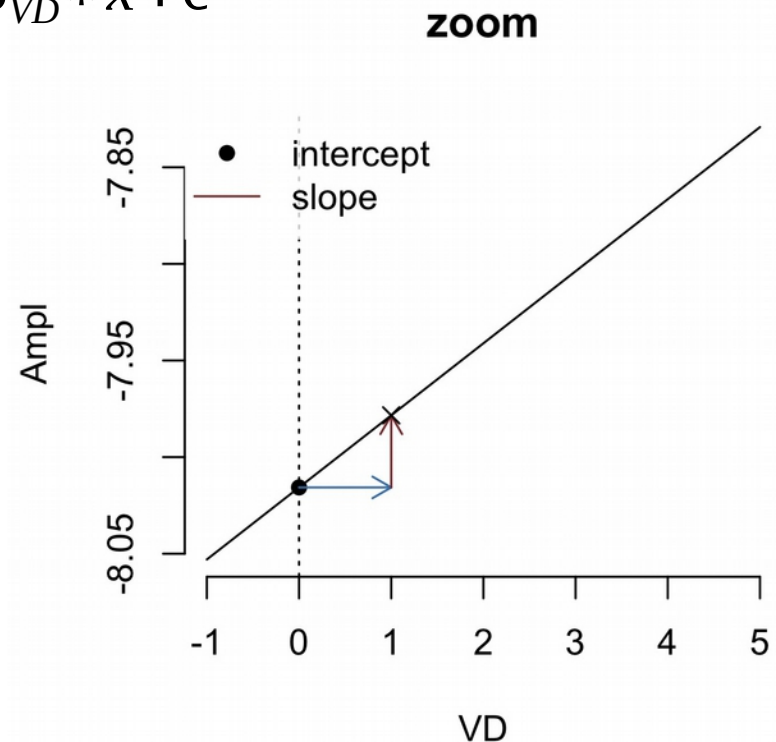
# Intercept and slope

■ Regression formula:  $y = \beta_0 + \beta_{VD} * x + \epsilon$

■  $\beta_0$  = Intercept: value of y when  $x=0$  (crossing with y-axis)

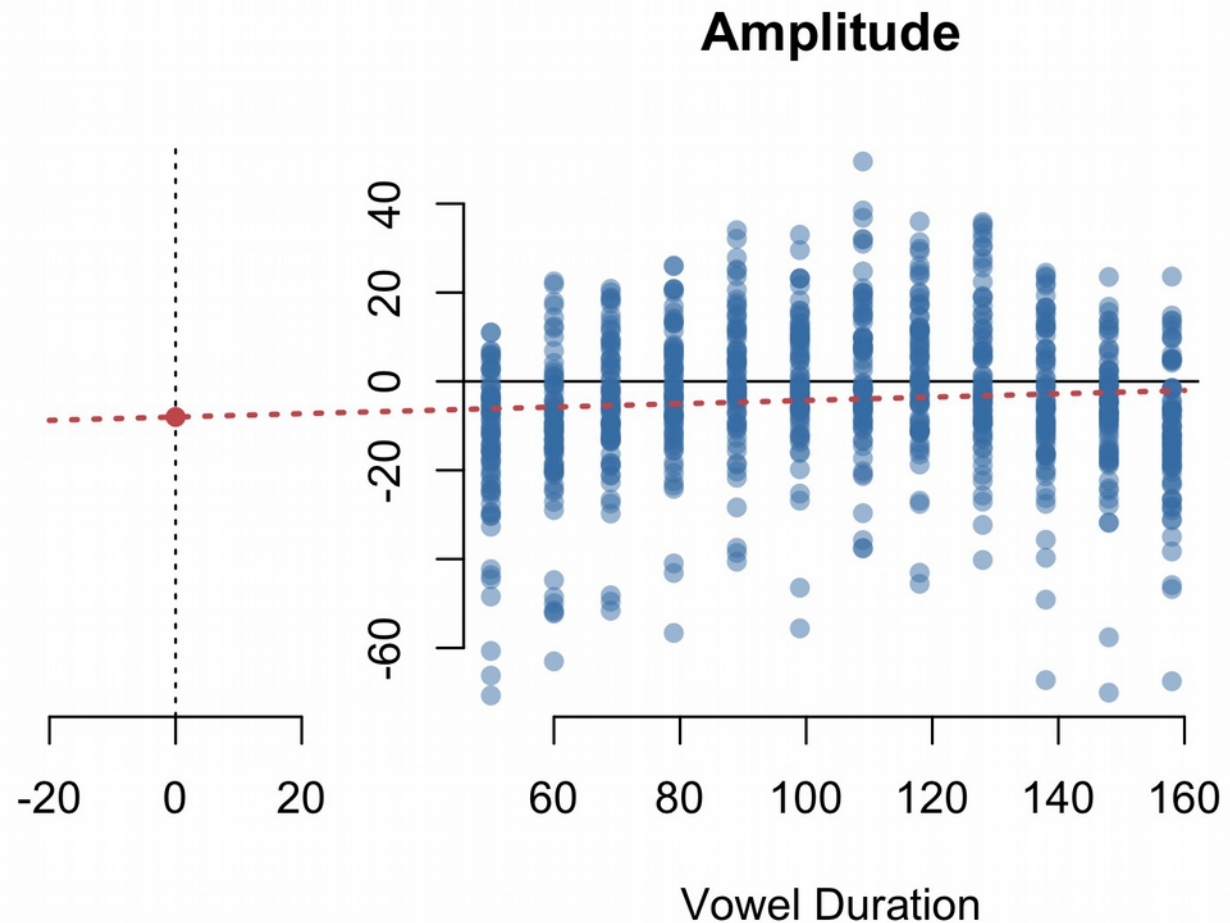
■  $\beta_1$  = Slope: change in y-value with increasing 1 unit on x-axis

■  $Y = -8.02 + 0.03730 * VD$





# Intercept and slope





# Categorical predictors

- *Mode*: “speech”, “nonspeech”
- Regression formula:  $y = \beta_0 + \beta_{VD} * x_{VD} + \beta_{Mode} * x_{Mode} + \epsilon$
- Contrast coding  

```
contrasts(dat$Mode)
      sp
ns    0
sp    1
```



# Categorical predictors

- Regression formula:  $y = \beta_0 + \beta_{VD} * x_{VD} + \beta_{Mode} * x_{Mode} + \epsilon$

```
lm2 <- lm(Amp ~ VD+Mode, data=dat)
summary(lm2)
```

Coefficients:

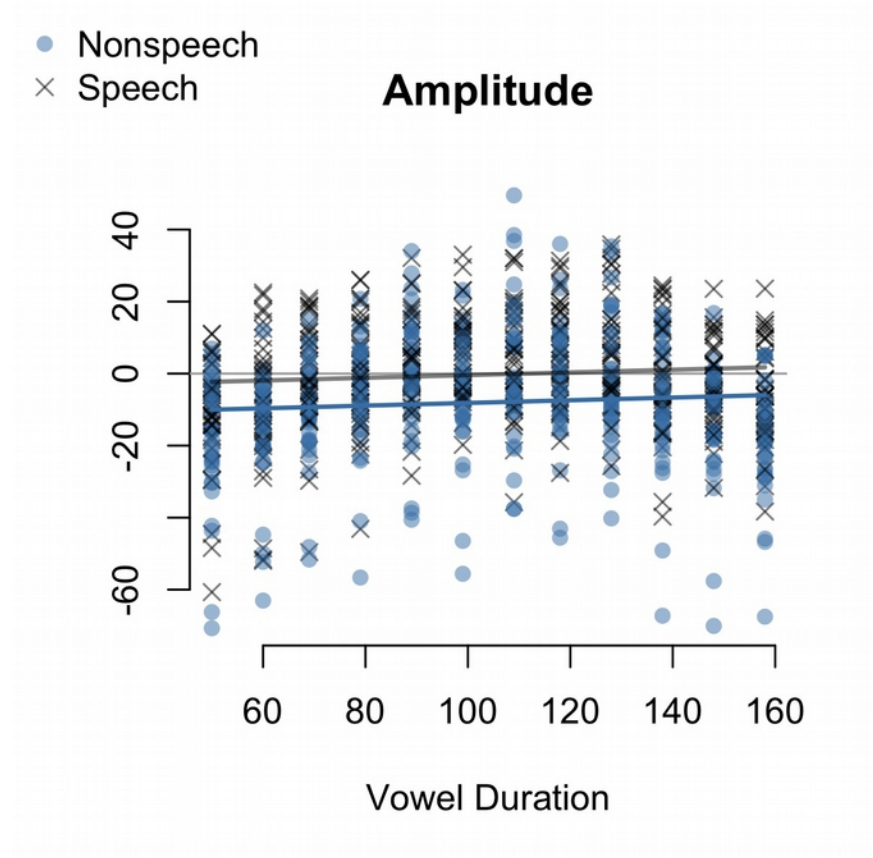
|             | <b>Estimate</b>  | <b>Std. Error</b> | <b>t</b> |
|-------------|------------------|-------------------|----------|
| (Intercept) | <b>-11.87865</b> | 1.82184           |          |
| VD          | <b>0.03730</b>   | 0.01594           |          |
| Modesp      | <b>7.72592</b>   | 1.07965           |          |

```
if Mode == "ns":
    x_Mode = 0
    + 0
if Mode s=="sp":
    x_Mode = 1
    + 7.72
```



# Categorical predictor

- Two regression lines:
  - Same effect of duration,
  - but difference in average amplitude between the two conditions:





# Interaction

```
lm3 <- lm(Ampl ~ VD*Mode, data=dat)
#      == lm(Ampl ~ VD + Mode + VD:Mode, dat)
```

```
summary(lm3)
```

Coefficients:

|             | <b>Estimate</b> | Std. Error | t value | Pr(> t ) |
|-------------|-----------------|------------|---------|----------|
| (Intercept) | <b>-9.87585</b> | 2.46014    | -4.014  | 6.45e-05 |
| VD          | <b>0.01799</b>  | 0.02254    | 0.798   | 0.425    |
| Modesp      | <b>3.72032</b>  | 3.47916    | 1.069   | 0.285    |
| VD:Modesp   | <b>0.03861</b>  | 0.03188    | 1.211   | 0.226    |



# Interaction

- Testing for significance:

**`anova(lm2, lm3)`**

Analysis of Variance Table

Model 1: `Ampl ~ VD + Mode`

Model 2: `Ampl ~ VD * Mode`

|   | Res.Df | RSS    | Df | Sum of Sq | F      | Pr(>F) |
|---|--------|--------|----|-----------|--------|--------|
| 1 | 909    | 241581 |    |           |        |        |
| 2 | 908    | 241192 | 1  | 389.6     | 1.4667 | 0.2262 |



# Fitted values and residuals

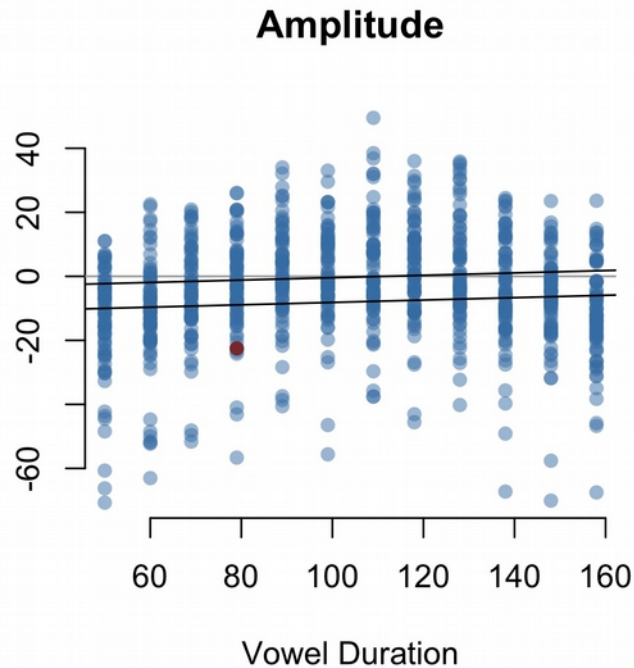
- Regression formula:  $y = \beta_0 + \beta_{VD} * x_{VD} + \beta_{Mode} * x_{Mode} + \epsilon$
- Regression model tries to predict how the predictors influence the values of response variable
  - Fitted values: explained part of data
  - Residuals: unexplained part of data
    - ▶ Should be gaussian distributed
    - ▶ No structure in residuals

$$\epsilon \sim N(0, \sigma)$$



# Model criticism

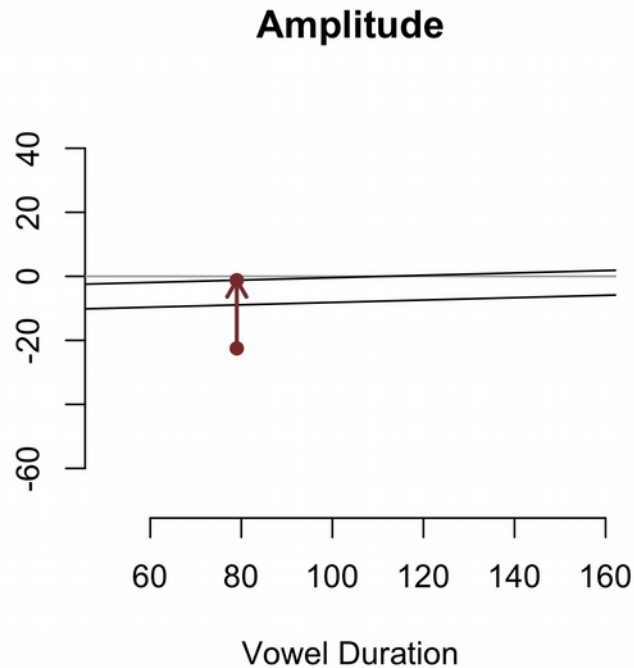
- Inspection of residuals: Comparing the fitted values with the actual data





# Model criticism

- Inspection of residuals: Comparing the fitted values with the actual data



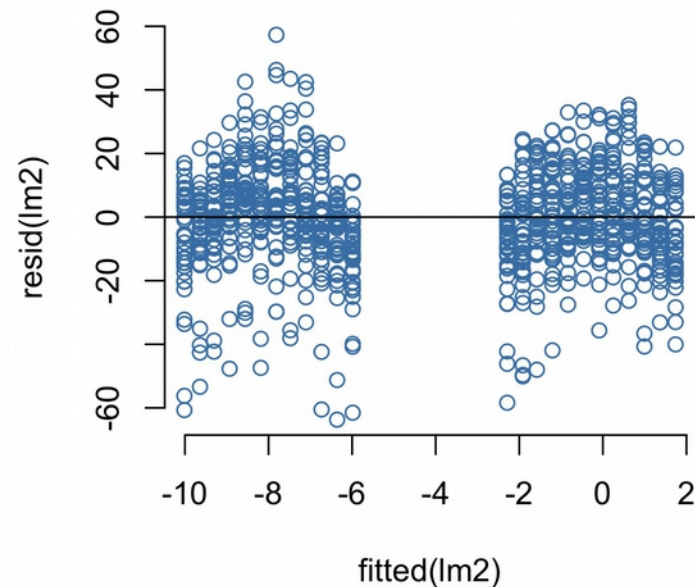
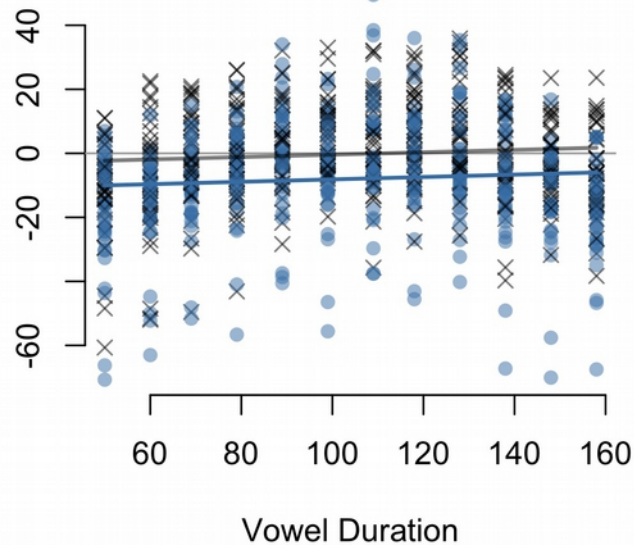


# Model criticism

- Inspection of residuals: Comparing the fitted values with the actual data

● Nonspeech  
× Speech

## Amplitude



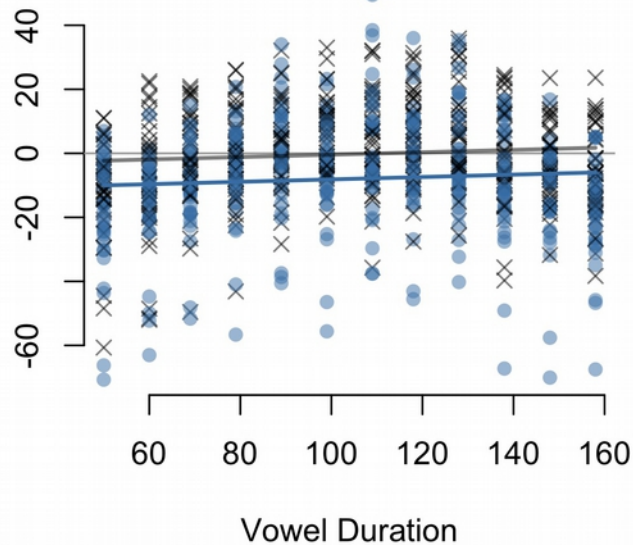


# Model criticism

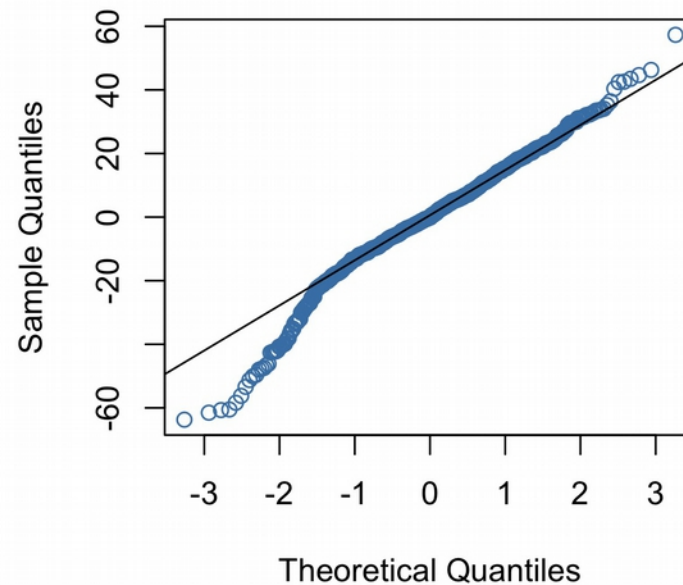
- Inspection of residuals: Comparing the fitted values with the actual data

● Nonspeech  
× Speech

**Amplitude**



**Normal Q-Q Plot**







## 2. GAMMs



# Generalized Additive Models (GAM)

- Relaxing assumption of linear relation between dependent variable and predictor

- Regression formula:  $y = \beta_0 + \beta_{VD} * x_{VD} + \beta_{Mode} * x_{Mode} + \epsilon$

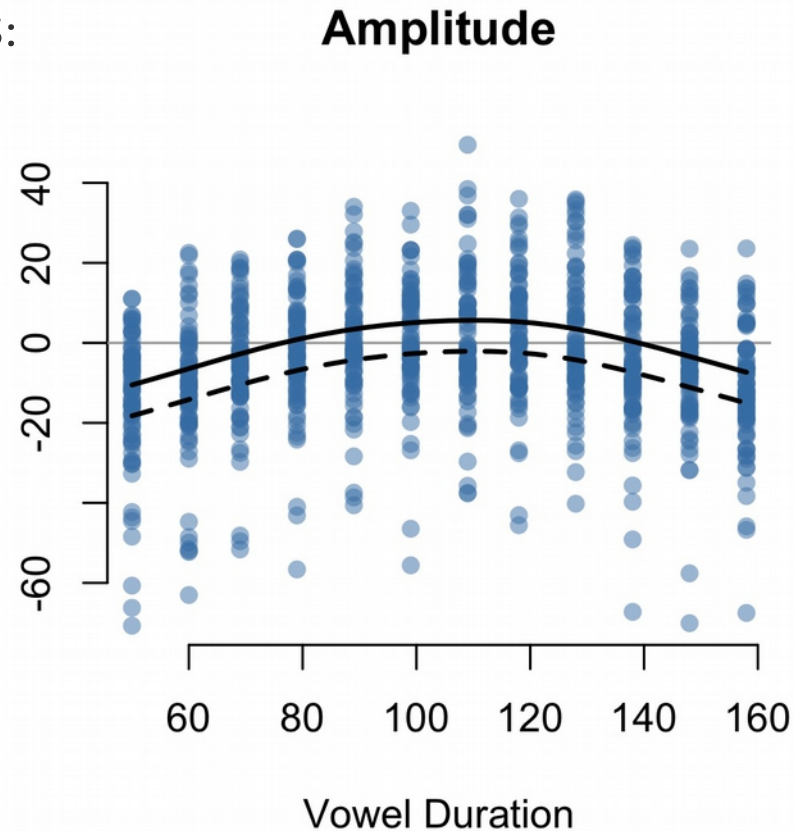
becomes:  $y = \beta_0 + f(x_{VD}) + f(x_{Mode}) + \epsilon$

- References:
  - GAMs (Lin & Zao, 1999)
  - R package mgcv (Wood, 2006; 2011)



# Generalized Additive Models (GAM)

- Nonlinear regression lines:





### 3. Articulography data

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