

# General Linear Model (GLM)

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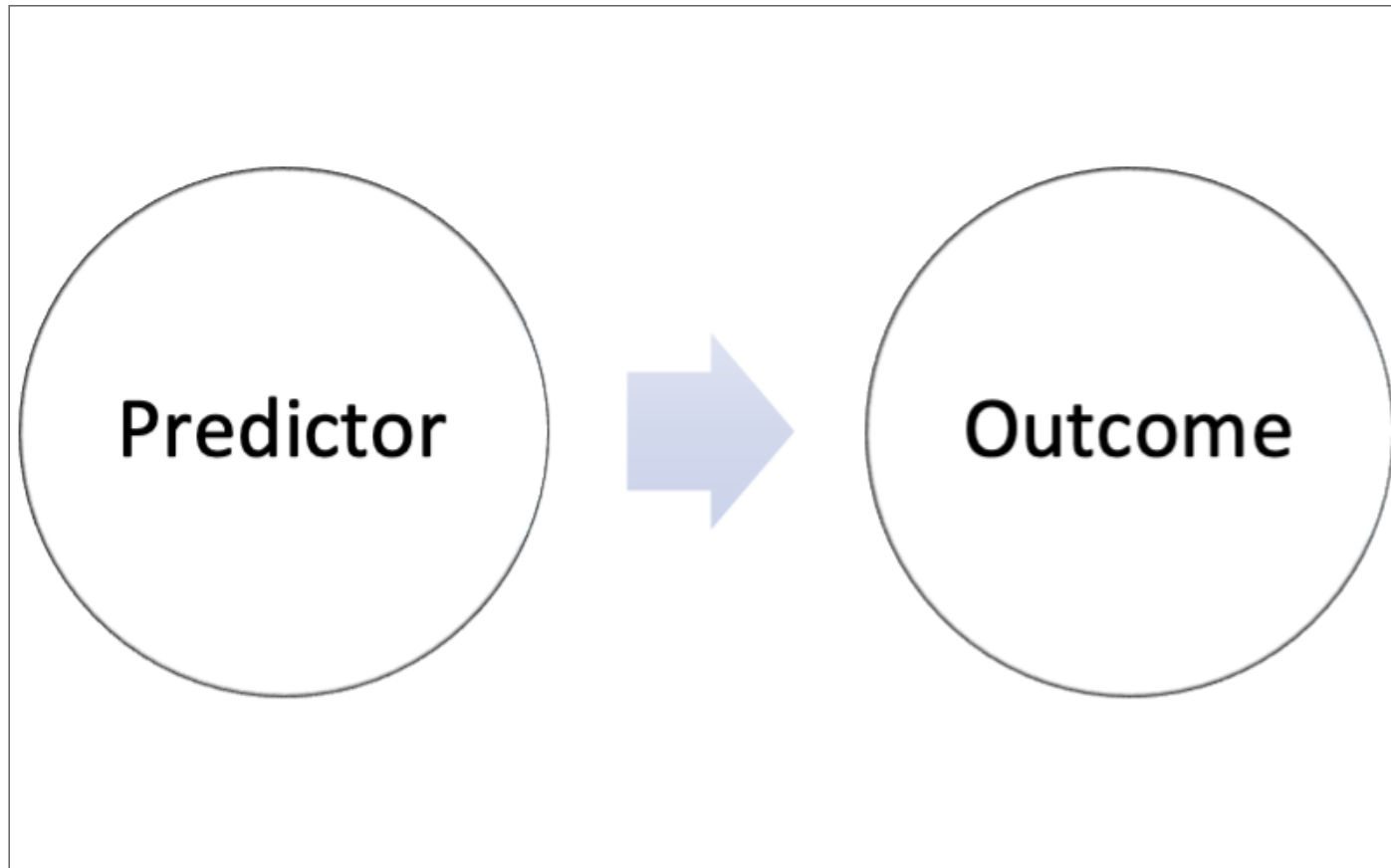


## Objectives

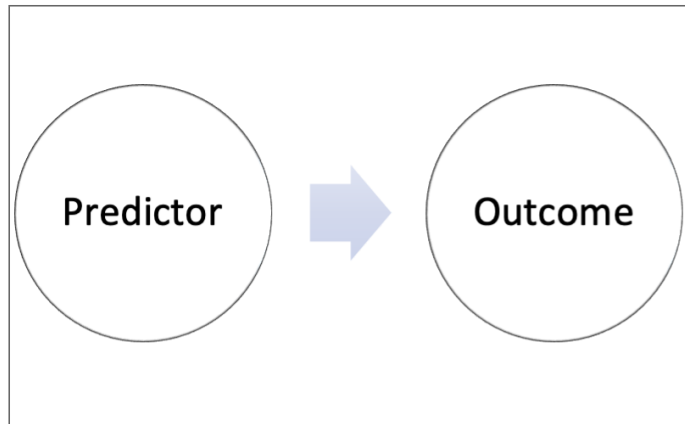
- Understand the basic concepts of the GLM
- Understand the usefulness of the GLM
- Understand how the GLM underlies most stats methods
- Understand the basic process of applying the GLM

# GLM basics

## Form

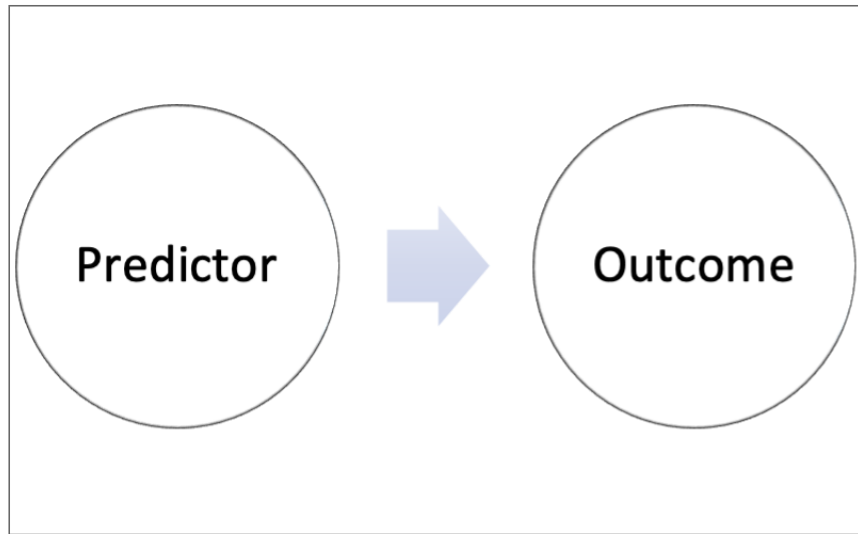


## Examples



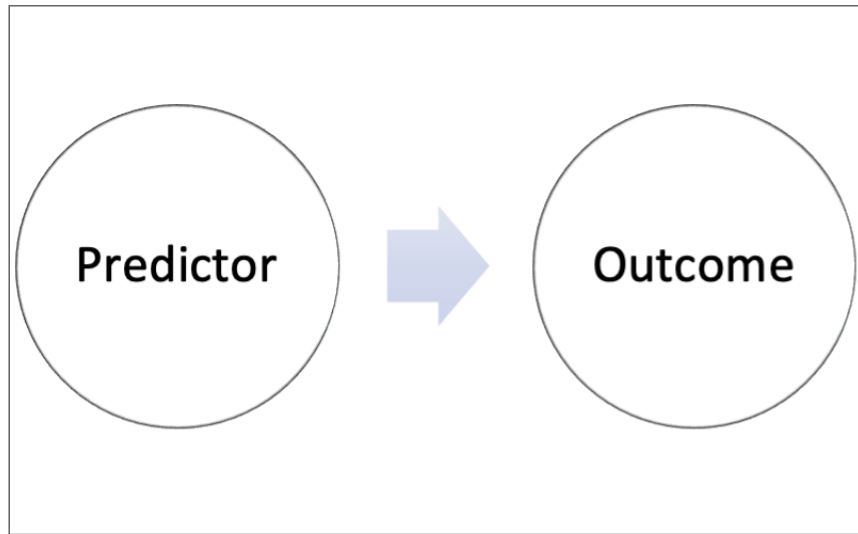
- Attention -> WM
- Art -> Sustained attention
- ADHD -> Inattention
- Celiac disease -> Processing speed
- Intervention -> Selective attention
- Musical training -> EF

## Form



- $Outcome = (Predictor)$
- $Outcome = (Predictor) + \text{error}$
- $Y = (\beta) + \epsilon$
- $Y = (\beta_0 + \beta_1) + \epsilon$
- $Y = (\beta_0 + \beta_1 + \beta_2) + \epsilon$

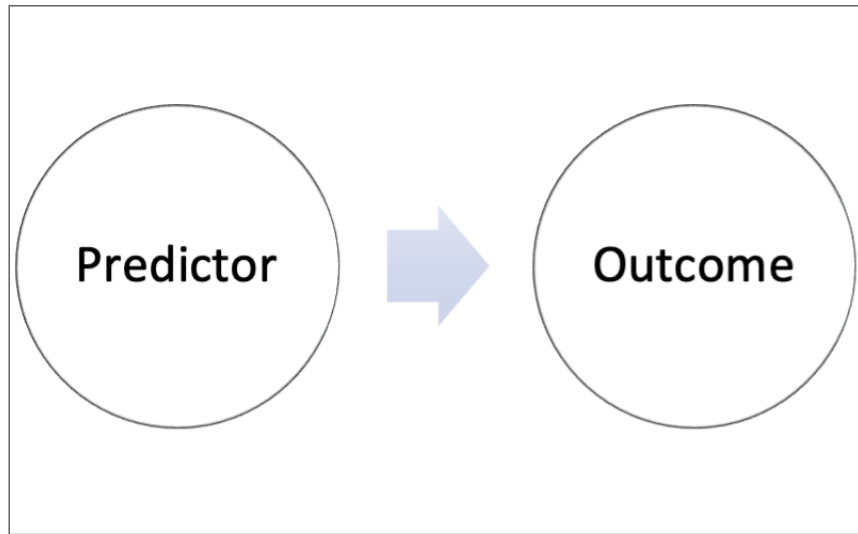
## Study effects



- Relationship
- Difference between groups



## Usefulness



- Existence: statistical significance
- Size: effect size, parameter

## GLM with different variables

**Lets see the variables**

## Phonological loop span

Ready?





0











Numbers?

## Selective attention

Ready?

Go

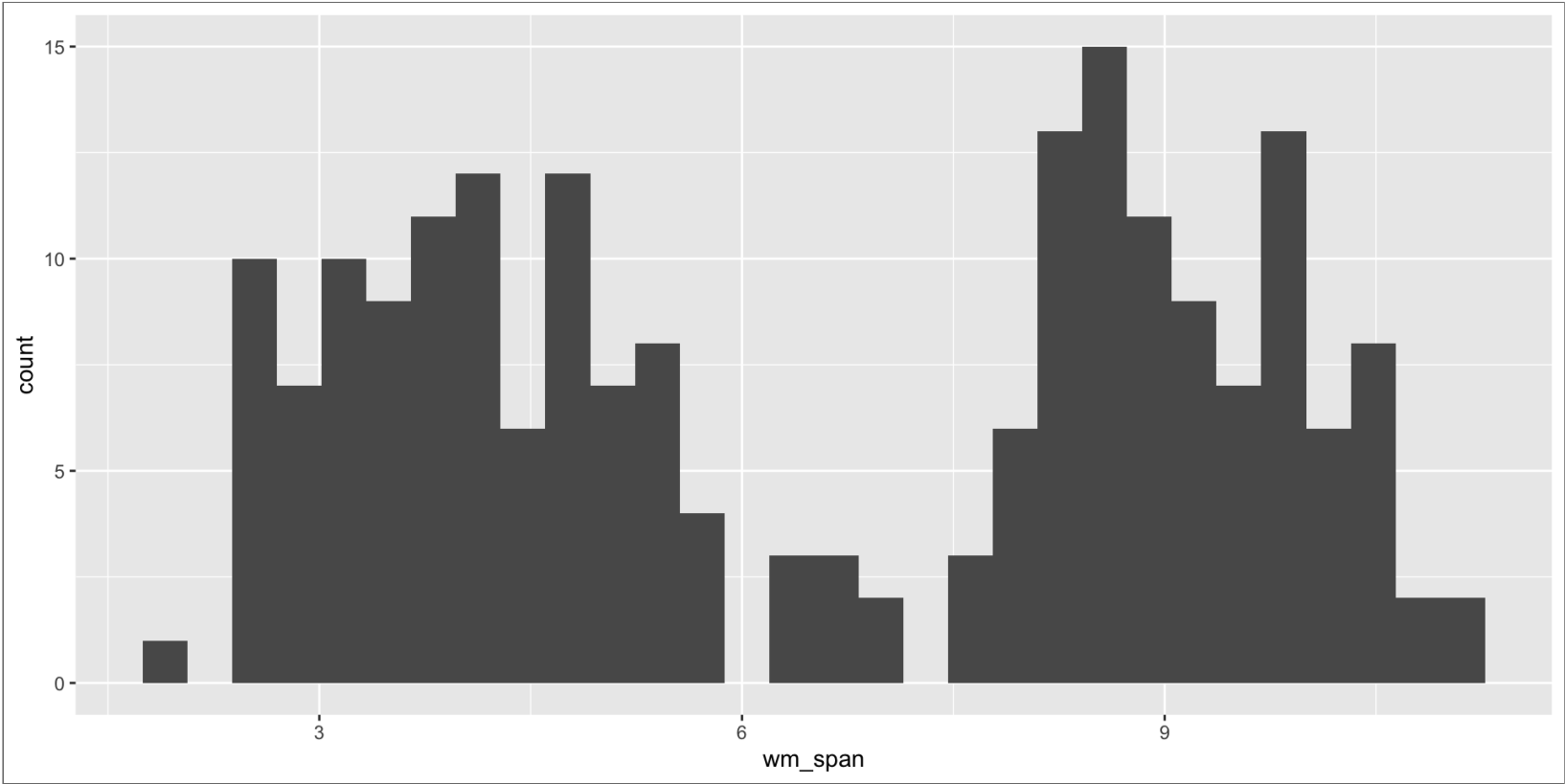


azul amarillo rojo  
marrón azul naranja  
rojo blanco verde  
azul rojo blanco



**Back to GLM with different variables**

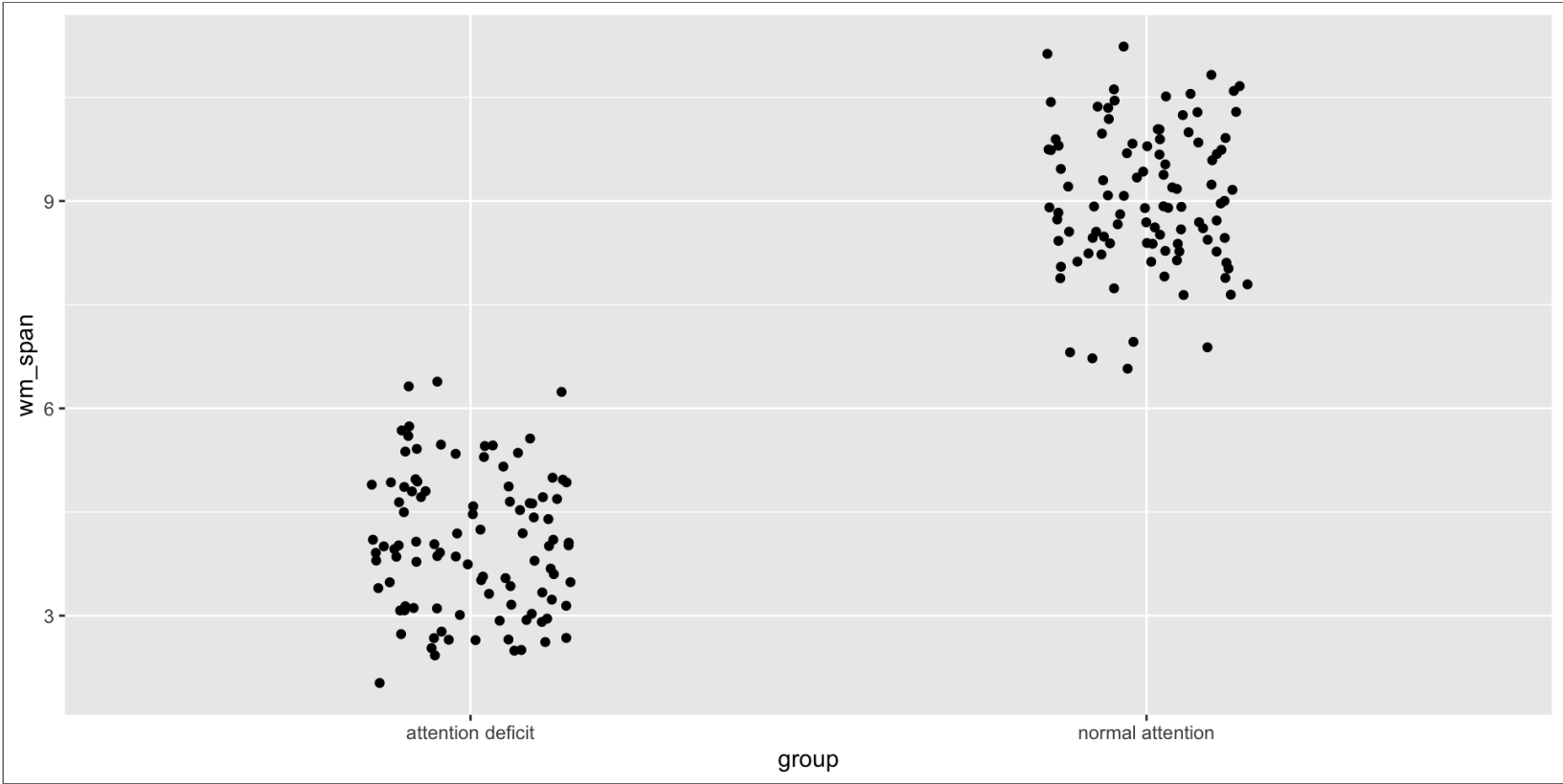
First, there were data



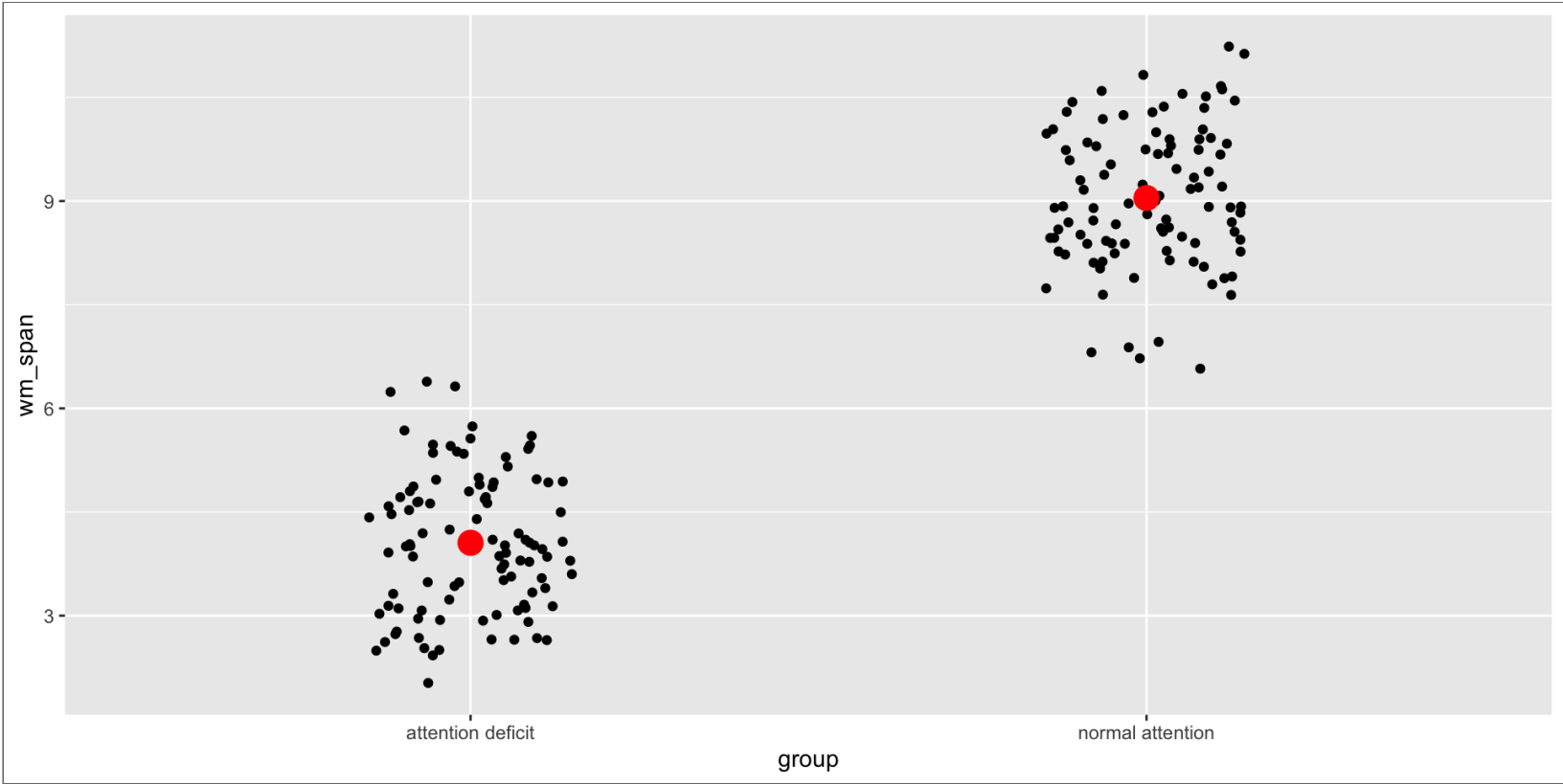
**Differences between 2 groups?**

**Process**

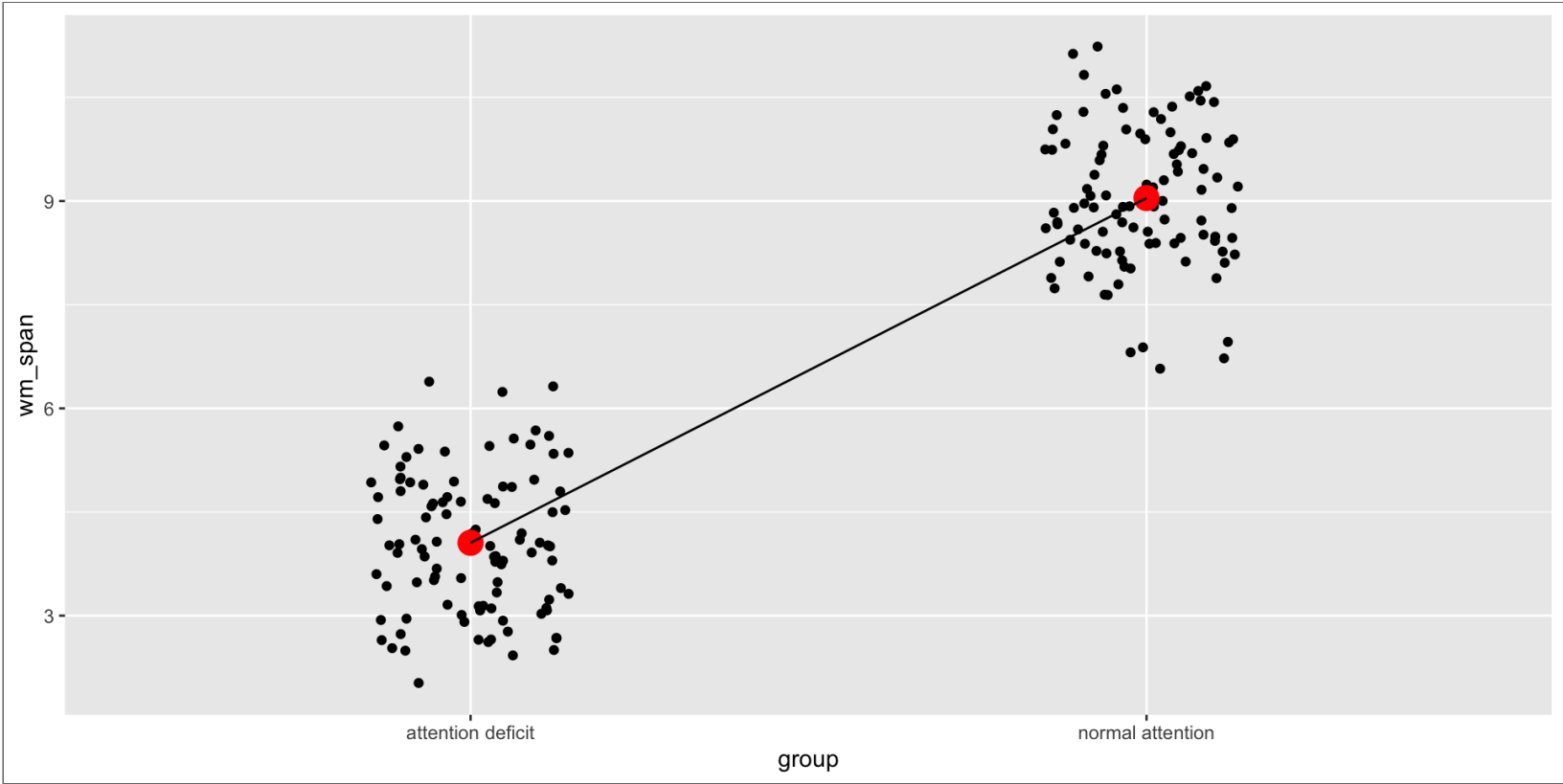
Group by attentional level



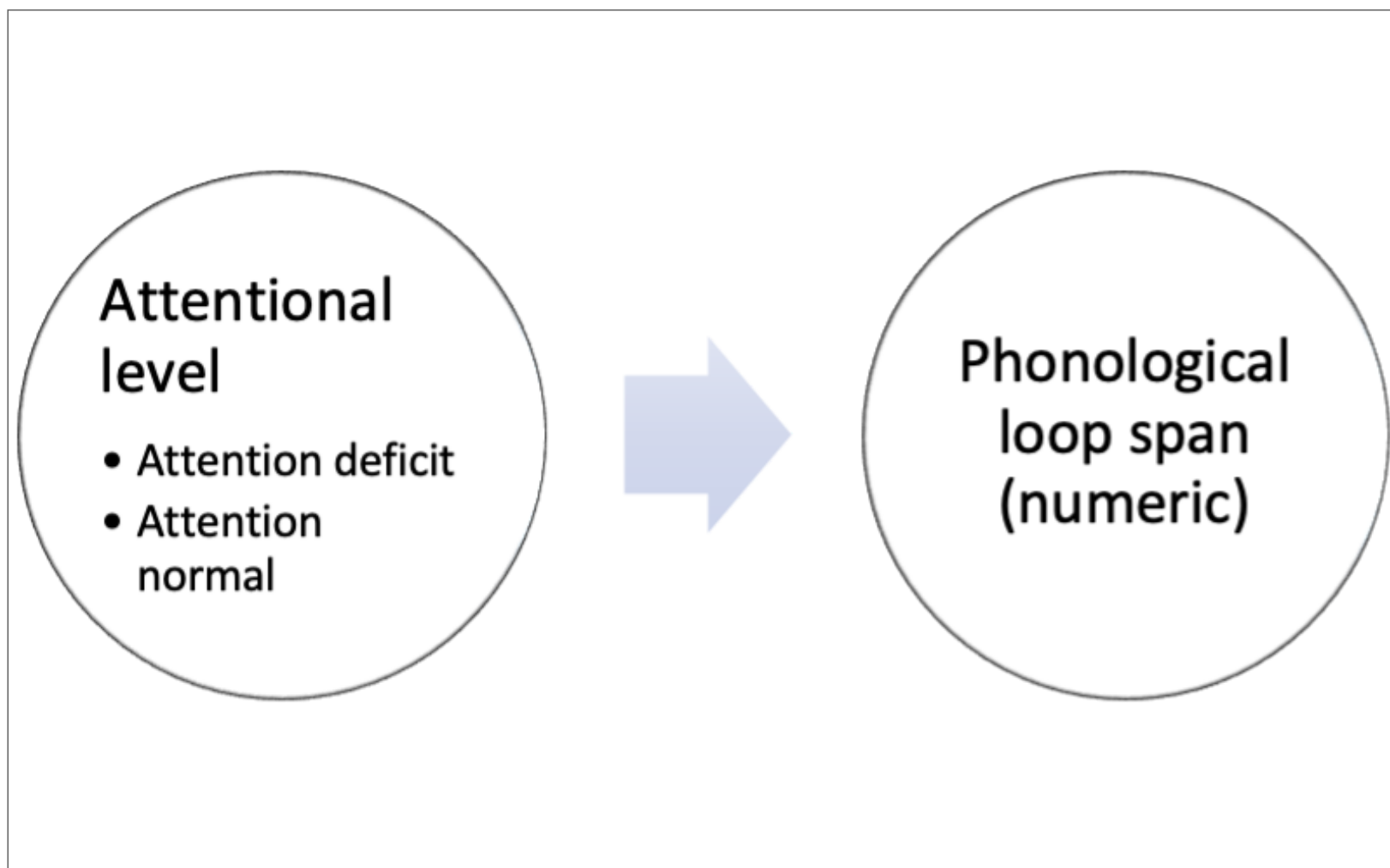
Estimate mean



Estimate relationship (difference)



## GLM form





## GLM analysis

```
Call:
lm(formula = wm_span ~ group, data = data_2_groups)

Residuals:
    Min       1Q   Median       3Q      Max
-2.4693 -0.7450 -0.0649  0.7651  2.3321

Coefficients:
              Estimate Std. Error t value Pr(>|t|)
(Intercept)      4.0547    0.1007   40.26  <2e-16 ***
groupnormal attention  4.9890    0.1424   35.03  <2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

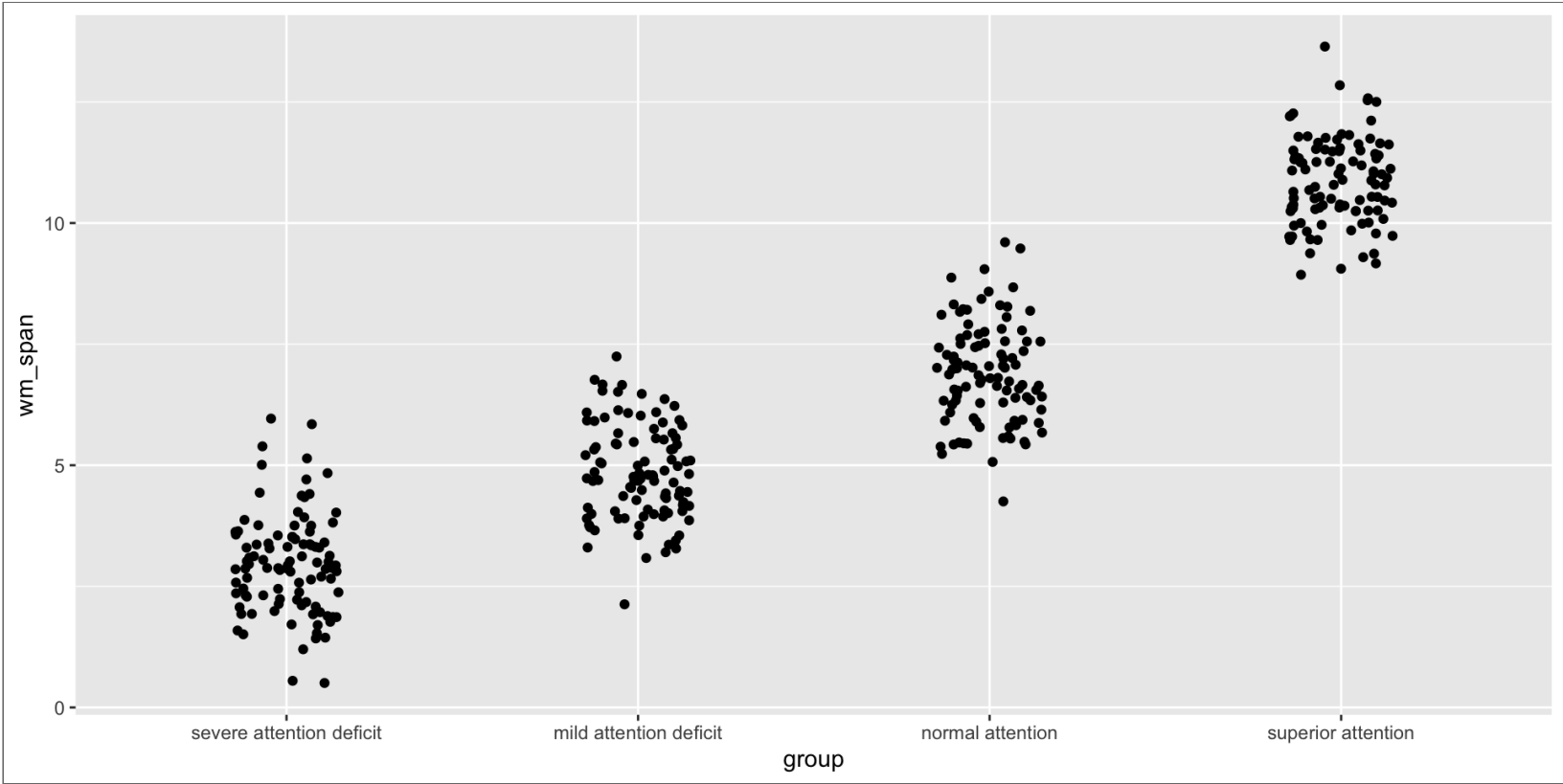
Residual standard error: 1.007 on 198 degrees of freedom
```

**A step further...**

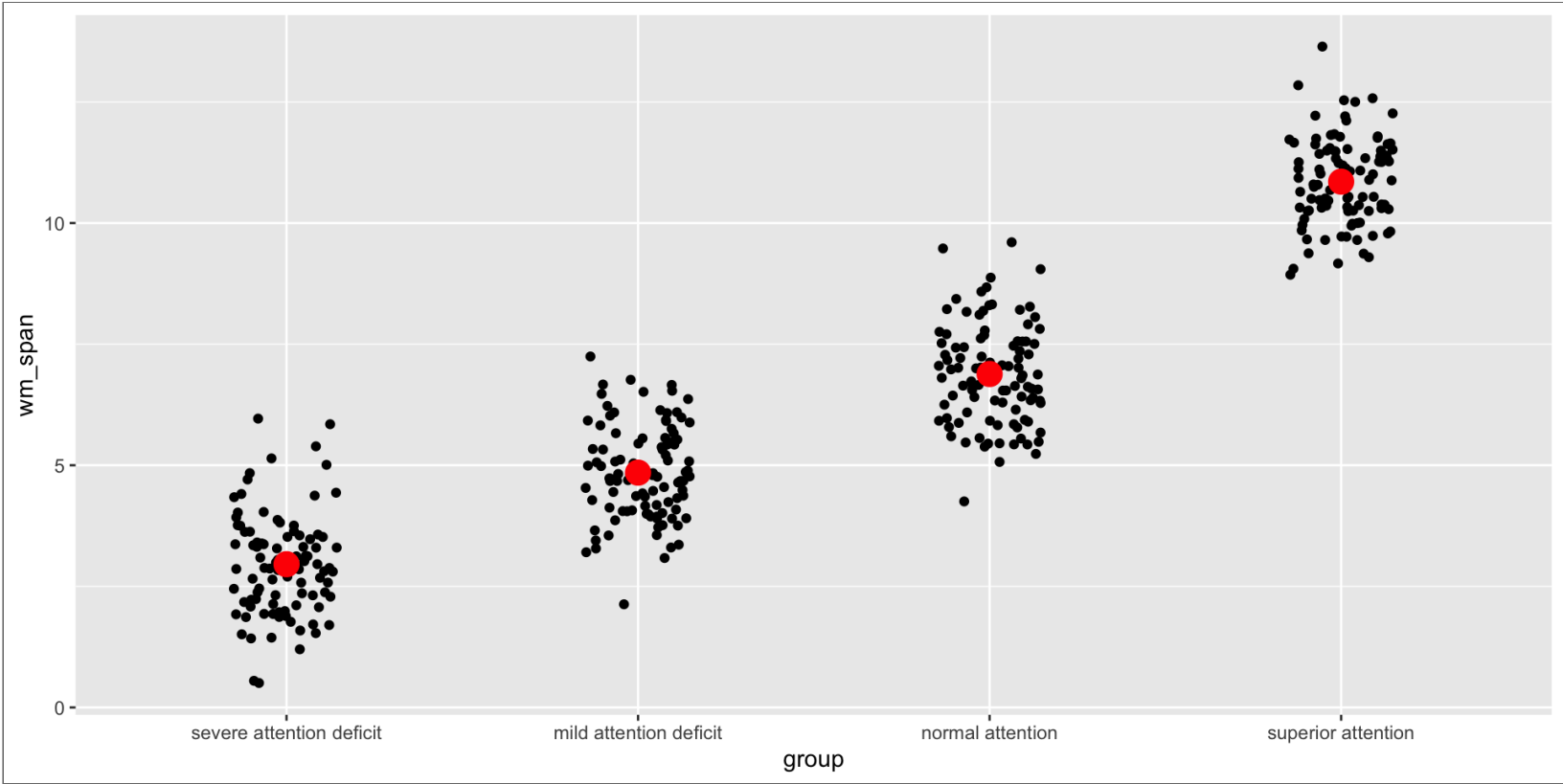
## Differences between 4 groups

**Process**

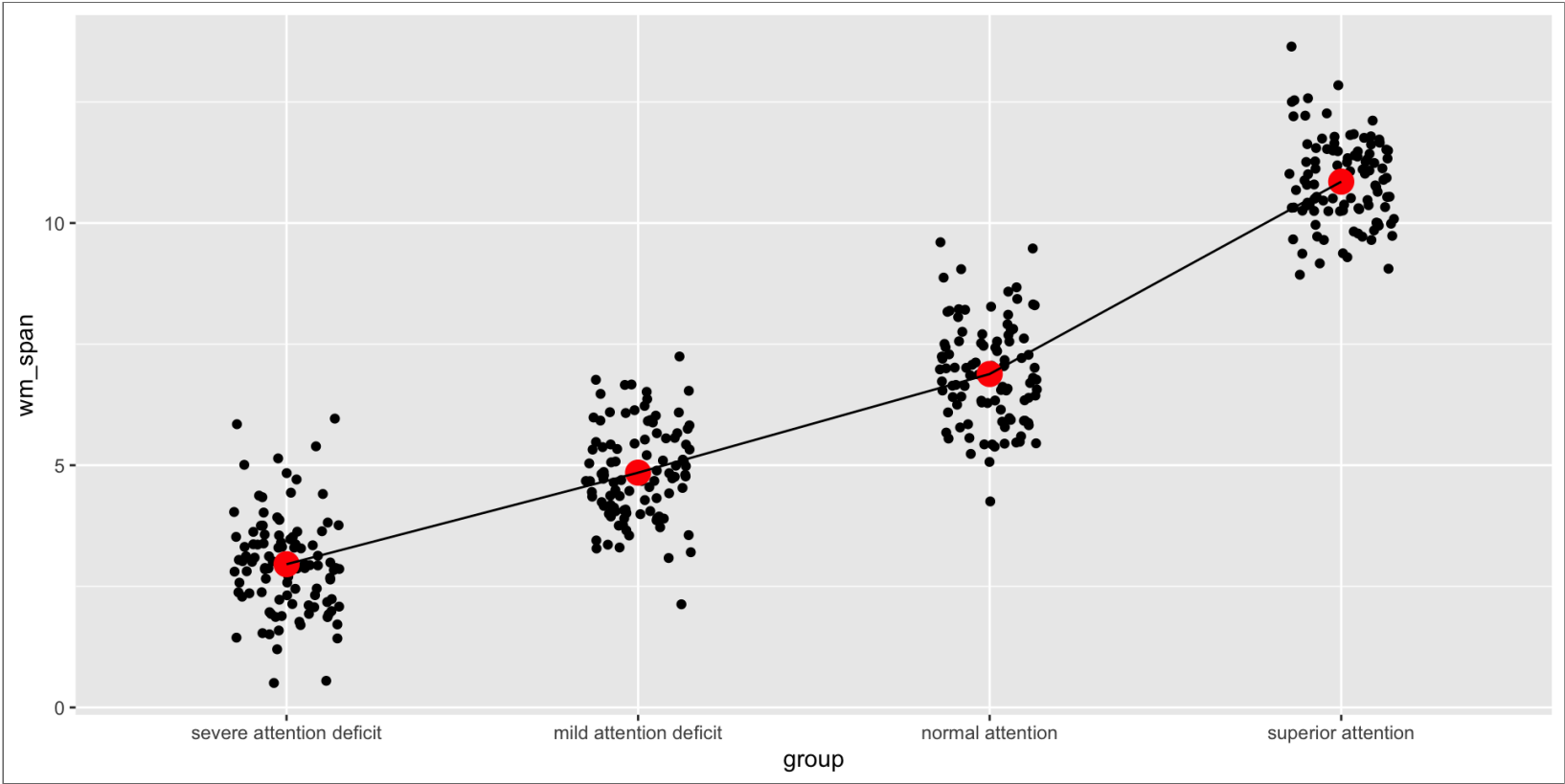
Group by attentional level



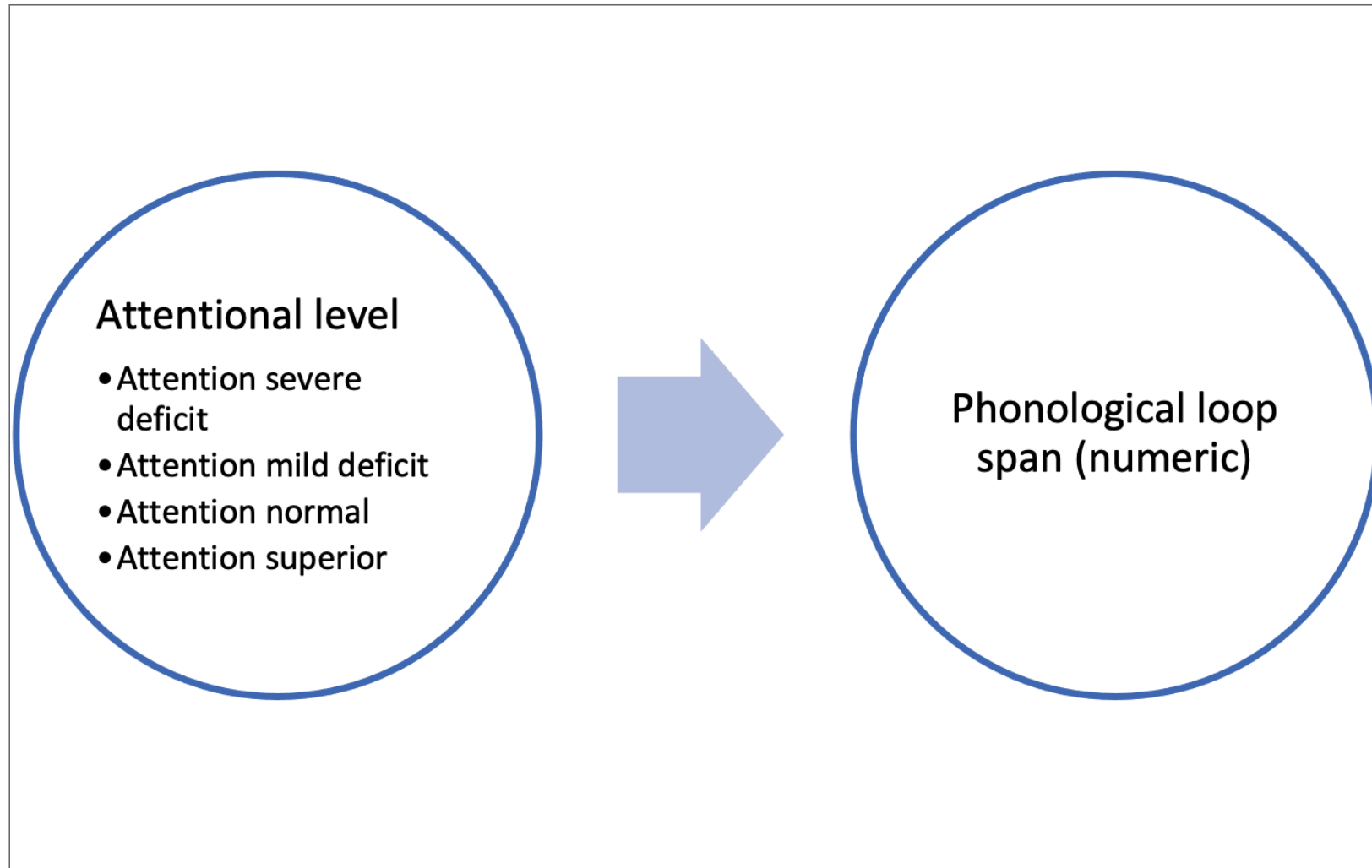
Estimate mean



Estimate relationship (difference)

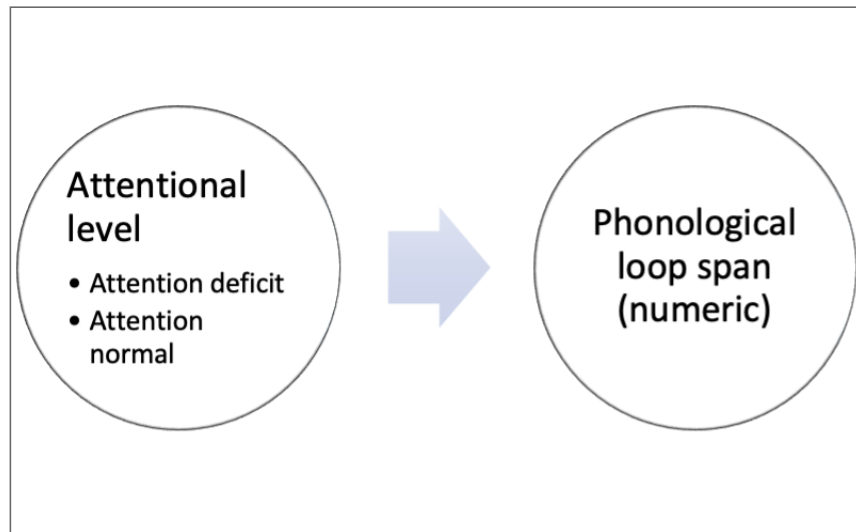
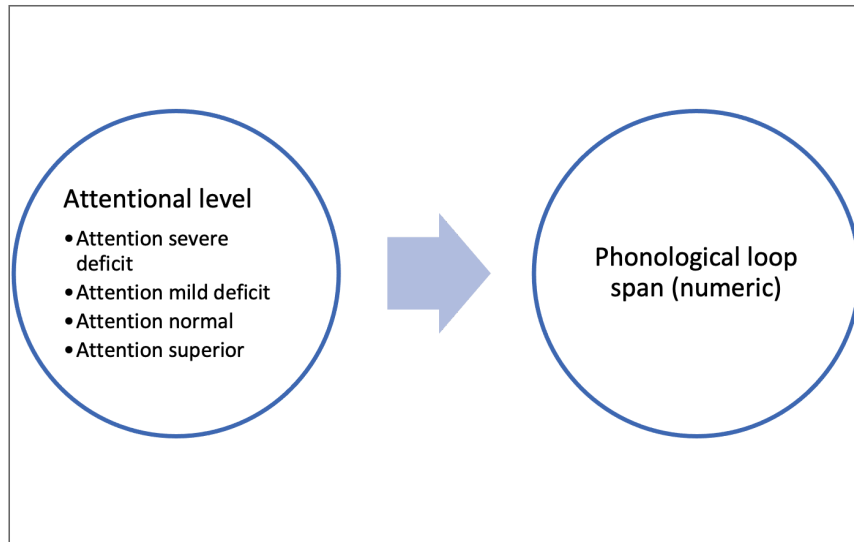


## GLM form





## GLM form 2 vs 4 groups



## GLM analysis

Call:

```
lm(formula = wm_span ~ group, data = data_4_groups)
```

Residuals:

Min	1Q	Median	3Q	Max
-2.71741	-0.69447	-0.06123	0.63348	3.00594

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	2.95692	0.09861	29.98	<2e-16	***
groupmild attention deficit	1.89012	0.13946	13.55	<2e-16	***
groupnormal attention	3.92689	0.13946	28.16	<2e-16	***
groupsuperior attention	7.89711	0.13946	56.63	<2e-16	***

---

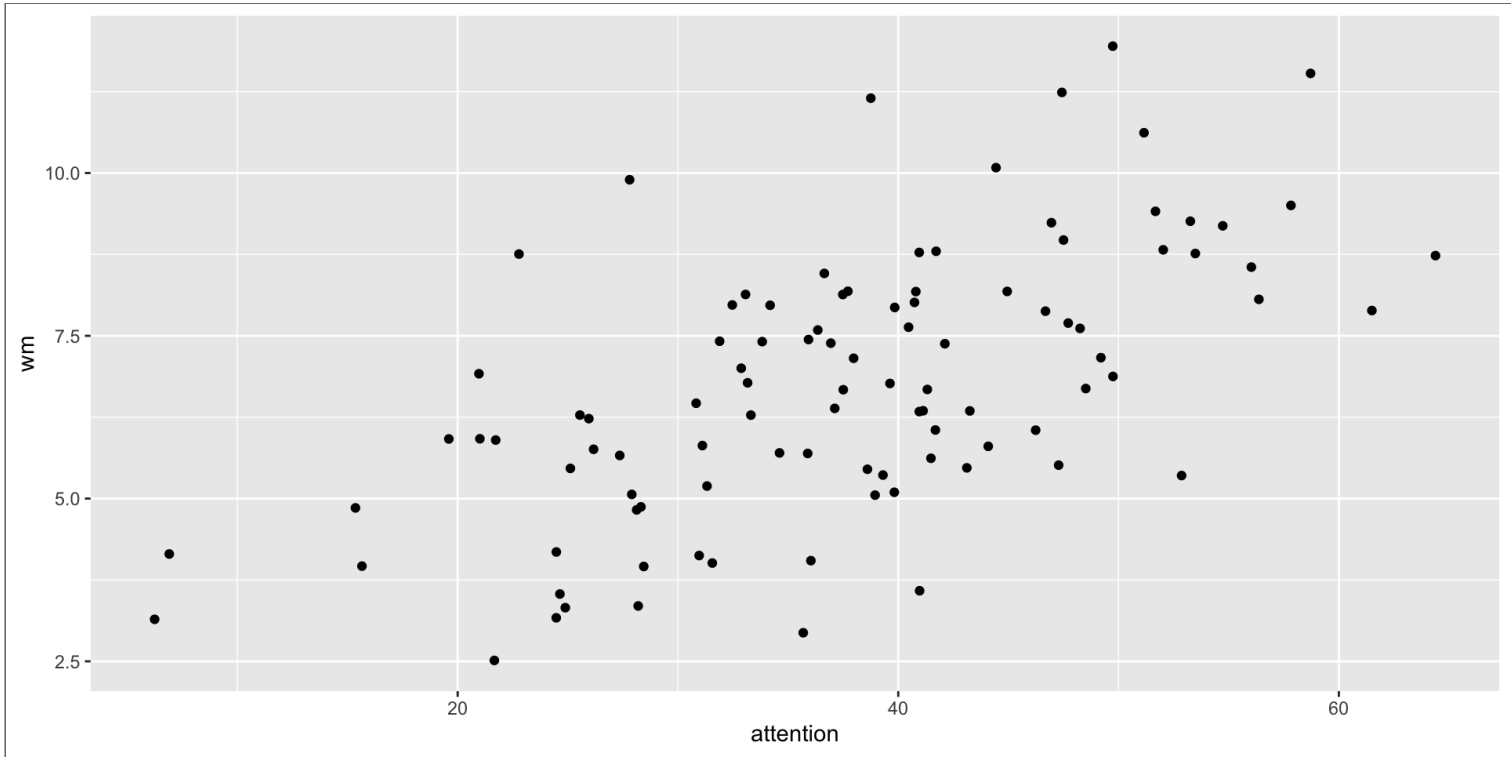
Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

**A step further...**

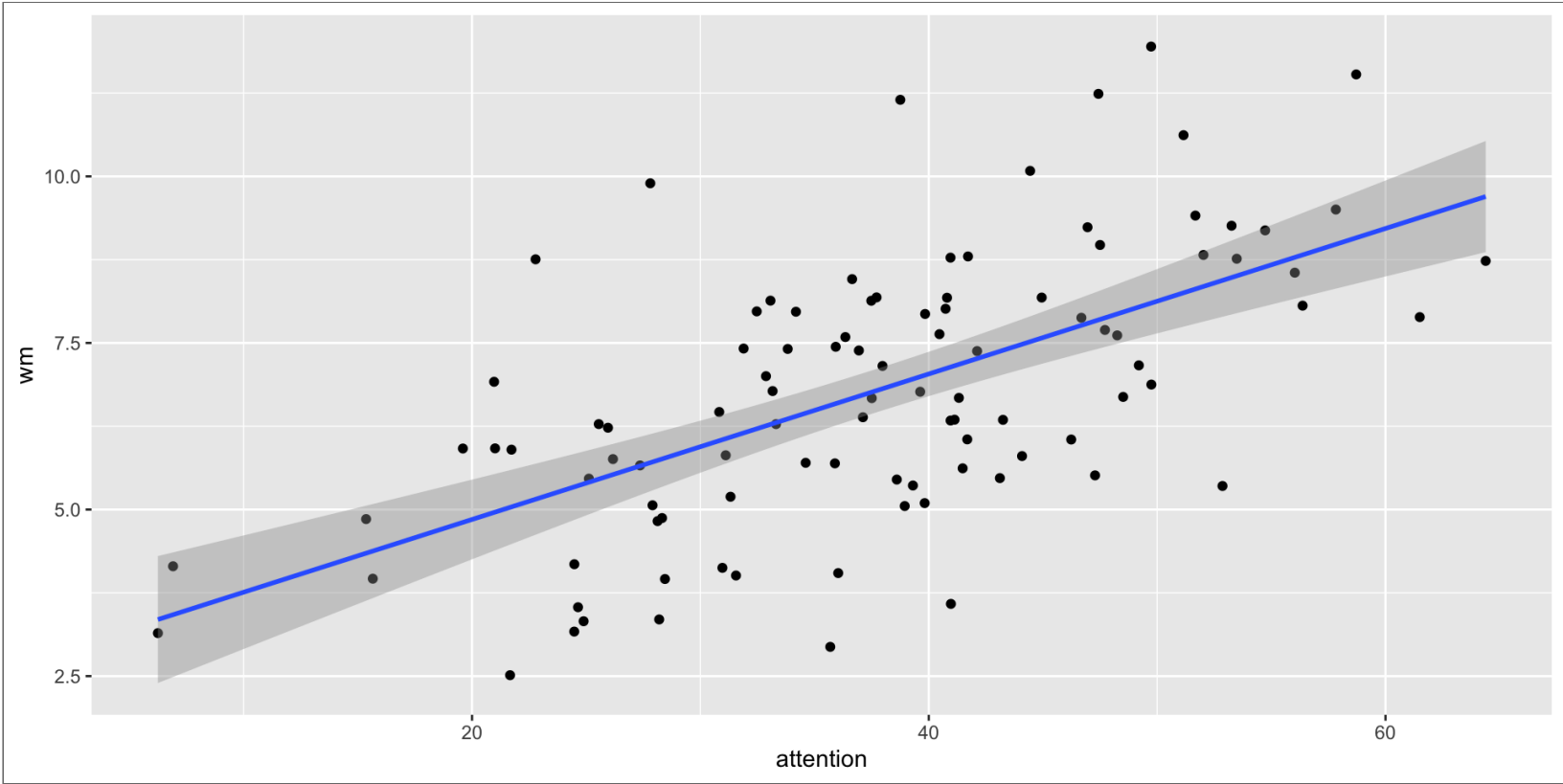
## Numeric predictors

# Attention and WM

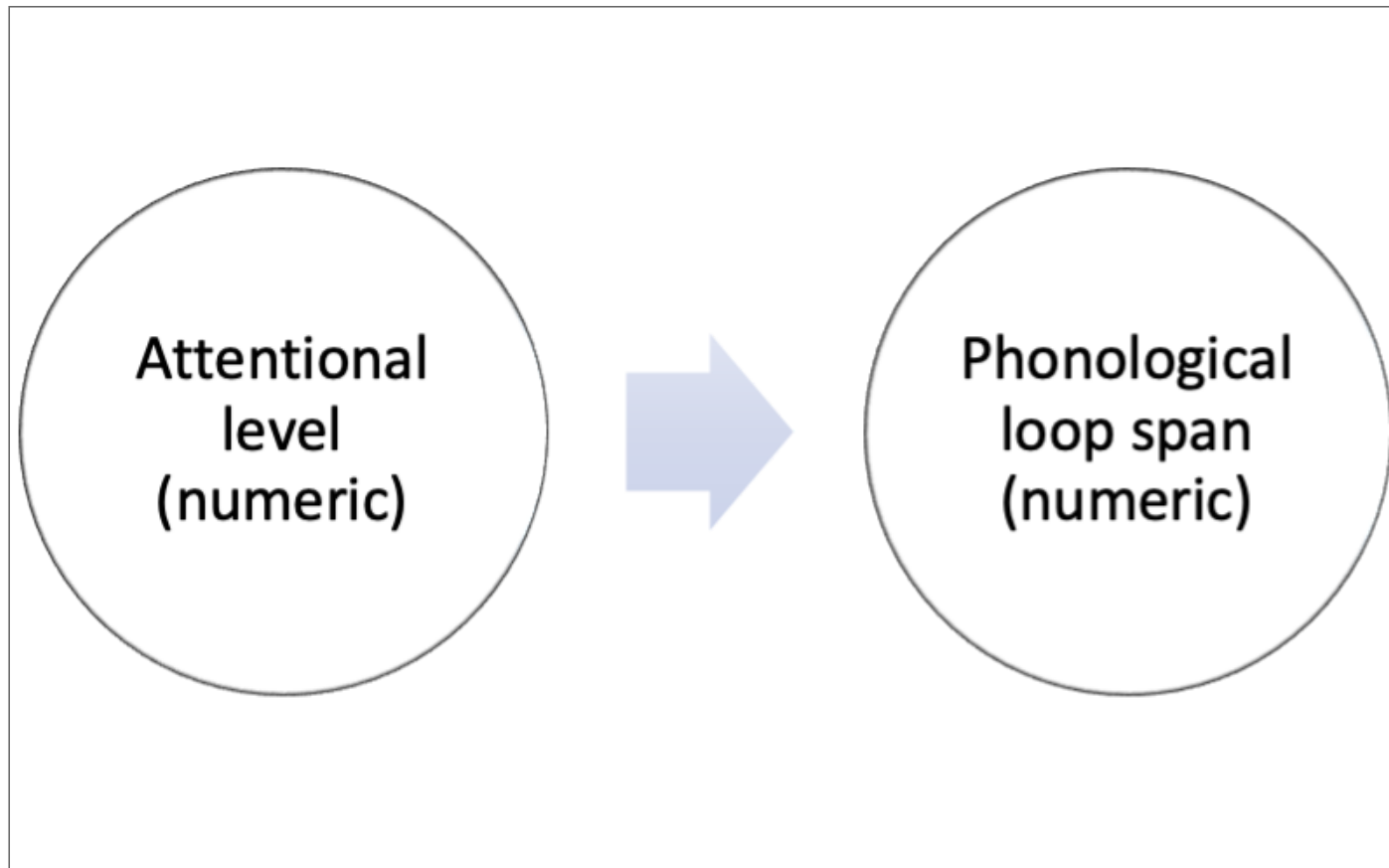
Estimate relationship



Estimate relationship line



## GLM form



## GLM analysis

```
Call:
lm(formula = wm ~ attention, data = data_cont_vars)

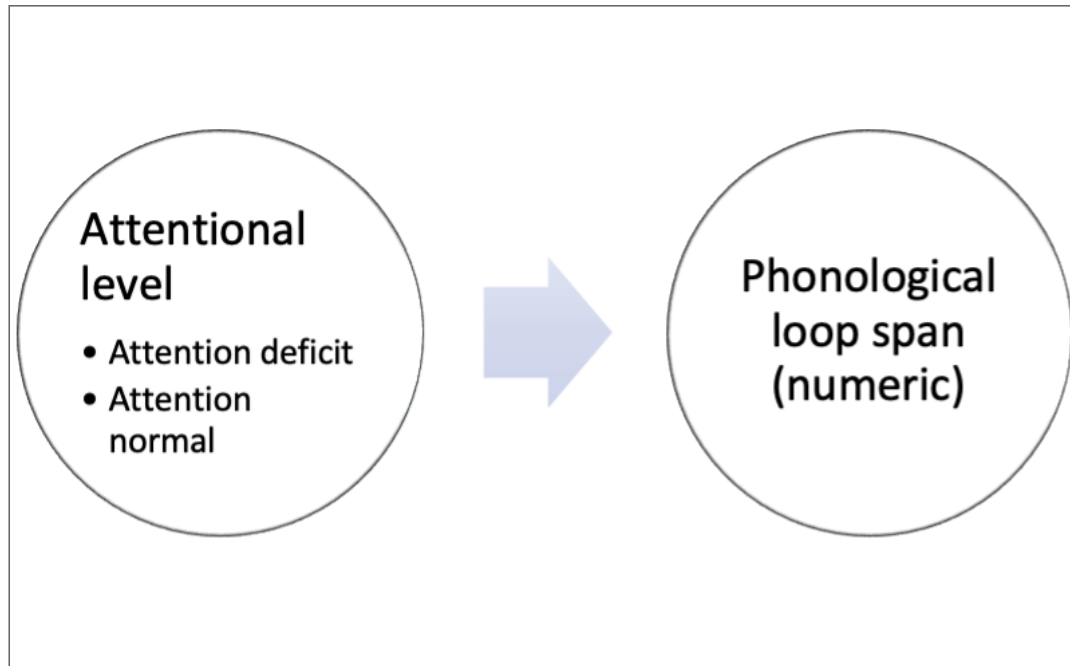
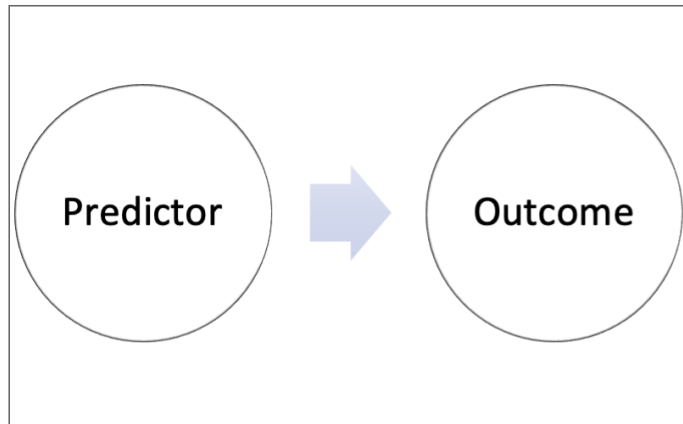
Residuals:
    Min       1Q   Median       3Q      Max
-3.6246 -1.1619  0.0310  0.9539  4.2508

Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept)  2.66748    0.56520   4.720 7.86e-06 ***
attention    0.10918    0.01441   7.577 2.01e-11 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 1.638 on 98 degrees of freedom
```



## Summary of models



### Attentional level

- Attention severe deficit
- Attention mild deficit
- Attention normal
- Attention superior



Phonological loop span (numeric)

Attentional level  
(numeric)



Phonological loop span  
(numeric)

## Closing

## Conclusions

- GLM underlies most stats methods
- Simple but powerful idea
- Use variables to predict variables
- Effects = relationships, differences

## Questions or Comments

## Further resources

- Andy Field Lectures - YouTube
- Field, A. (2017). Discovering Statistics Using IBM SPSS Statistics (5th ed.). London: Sage Publications. Chapter 2.

## Bonus

- Always GLM



# GLM subtypes

