



STAT199

Tuesday

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5:15 PM

# MARCH MADNESS

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Blue Devils



Basketball!

# AGENDA

## INTRODUCTION



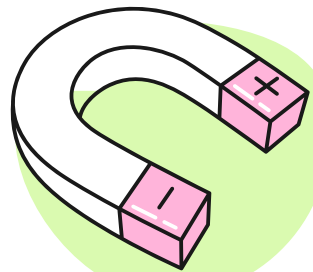
What are we accomplishing?

## EDA



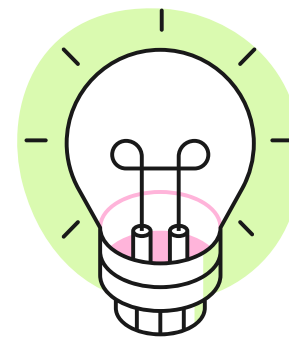
What did we do?

## RESULTS



What did we find?

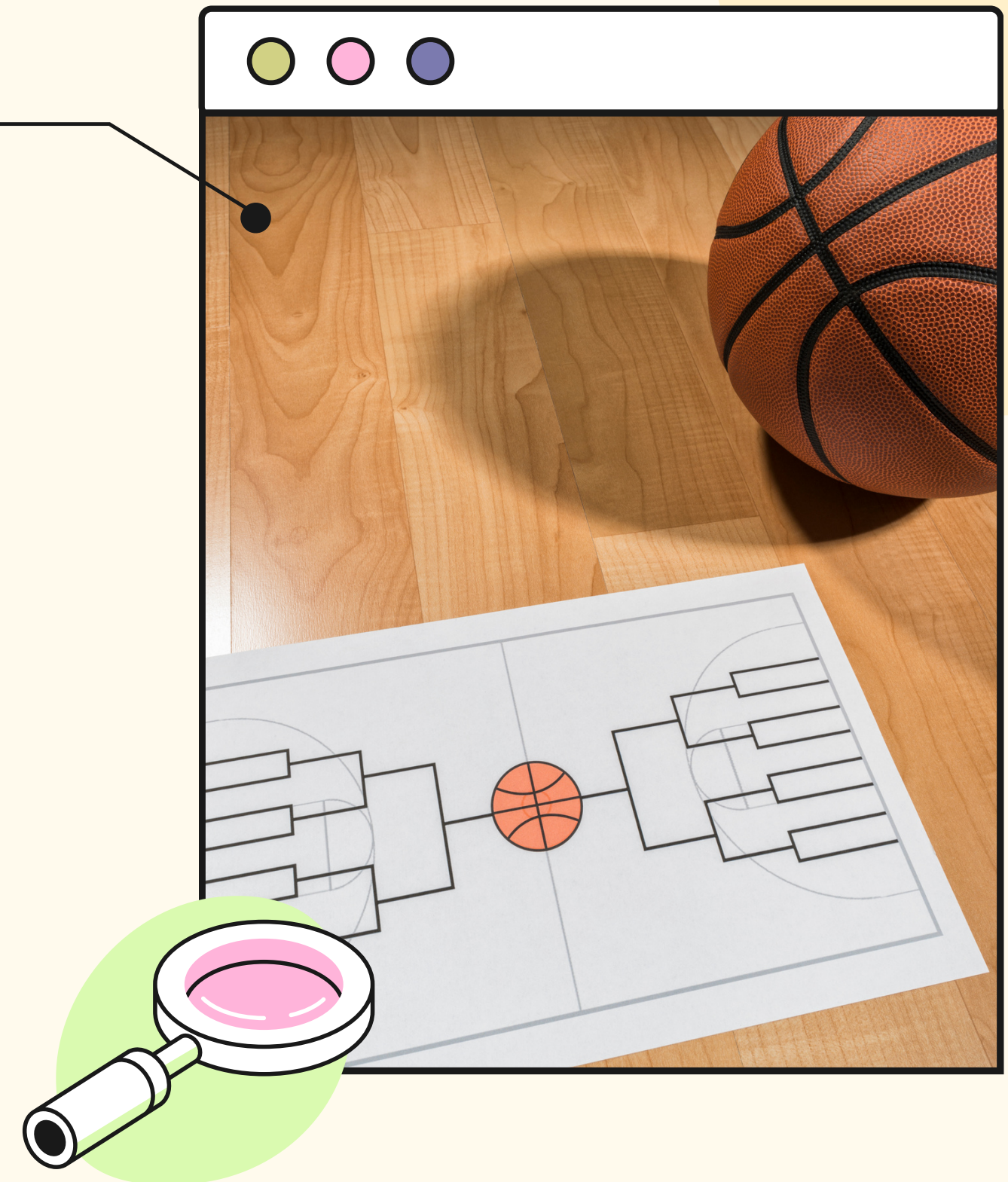
## CONCLUSION



Now what?

# INTRODUCTION

The March Madness tournament is a highly anticipated event for college basketball fans, who try to fill out the perfect bracket by predicting which teams will make it to each round. This project aims to identify the strongest predictors of success in the tournament to answer our question "What factors are the strongest predictors of success in March Madness?"



# INTRODUCTION

## **BARTHAG**

The team's chance of winning against the average DI team

## **BARTTORVIK ADJUSTED EFFICIENCY**

Bart Torvik's calculation of how efficient a team is offensively and defensively

## **WINS ABOVE BUBBLE**

How many more or less wins the average bubble team would have against the team's schedule

## **Hypothesis:**

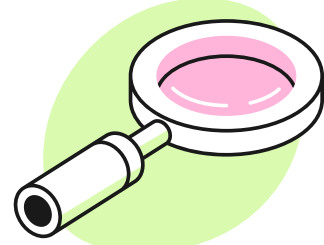
Teams with a high BARTHAG value will most be likely to find success in the tournament.

`using set.seed(101)`

**EDA**

# METHODOLOGY

## BOXPLOT



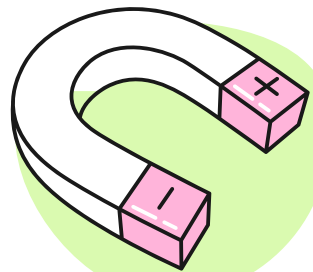
Is there a general trend?

## FIT MODELS\*\*



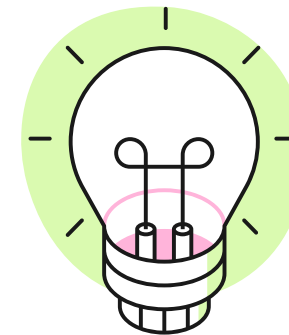
What's the equation?  
*\*\*Start with single variables!*

## ROC



How does the ROC look?

## AUC

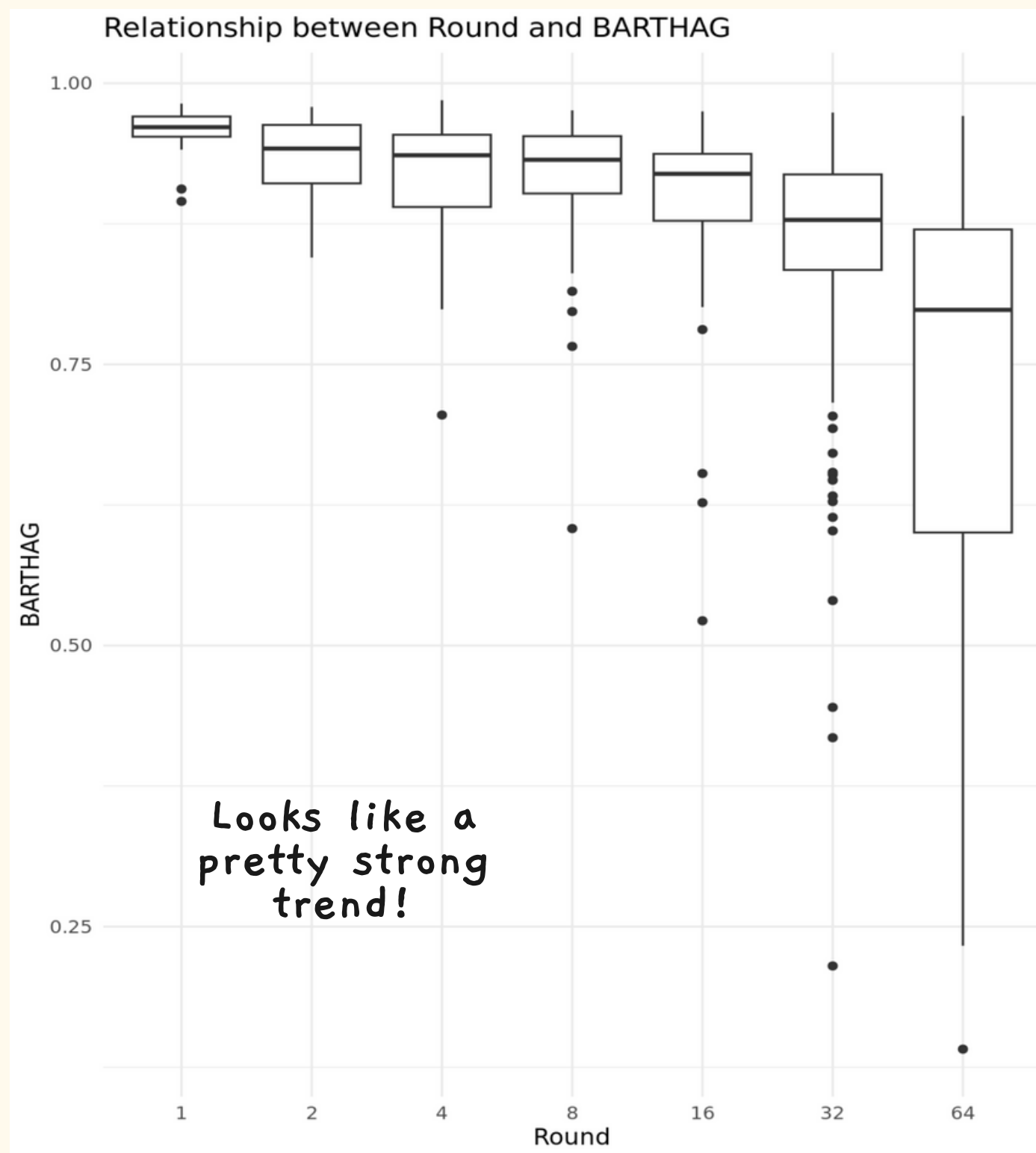


What's the area under  
the ROC curve?

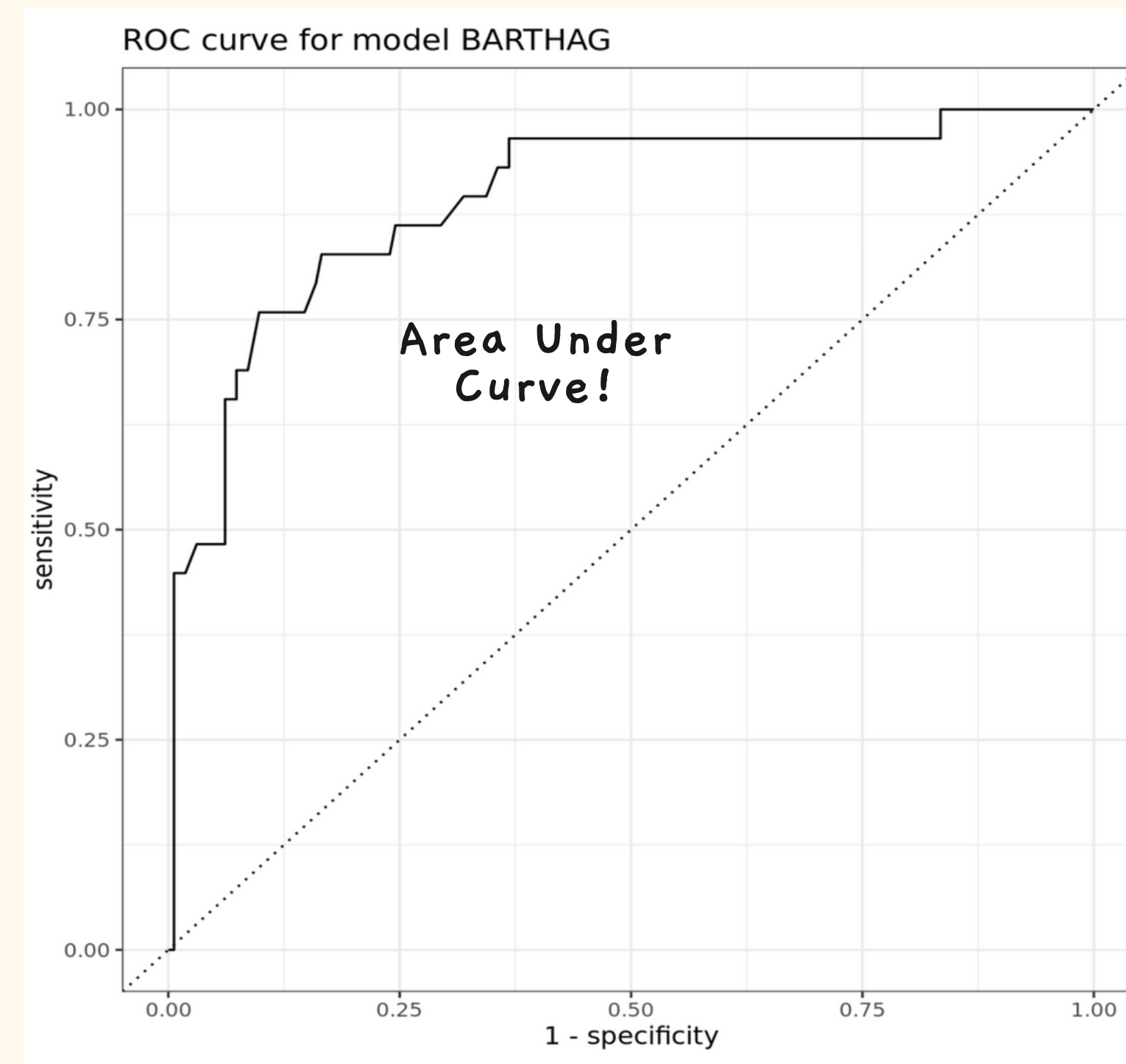
**Lastly, MIX AND MATCH variables to find the BEST model!**

# BARTHAG

EDA



BOXPLOT

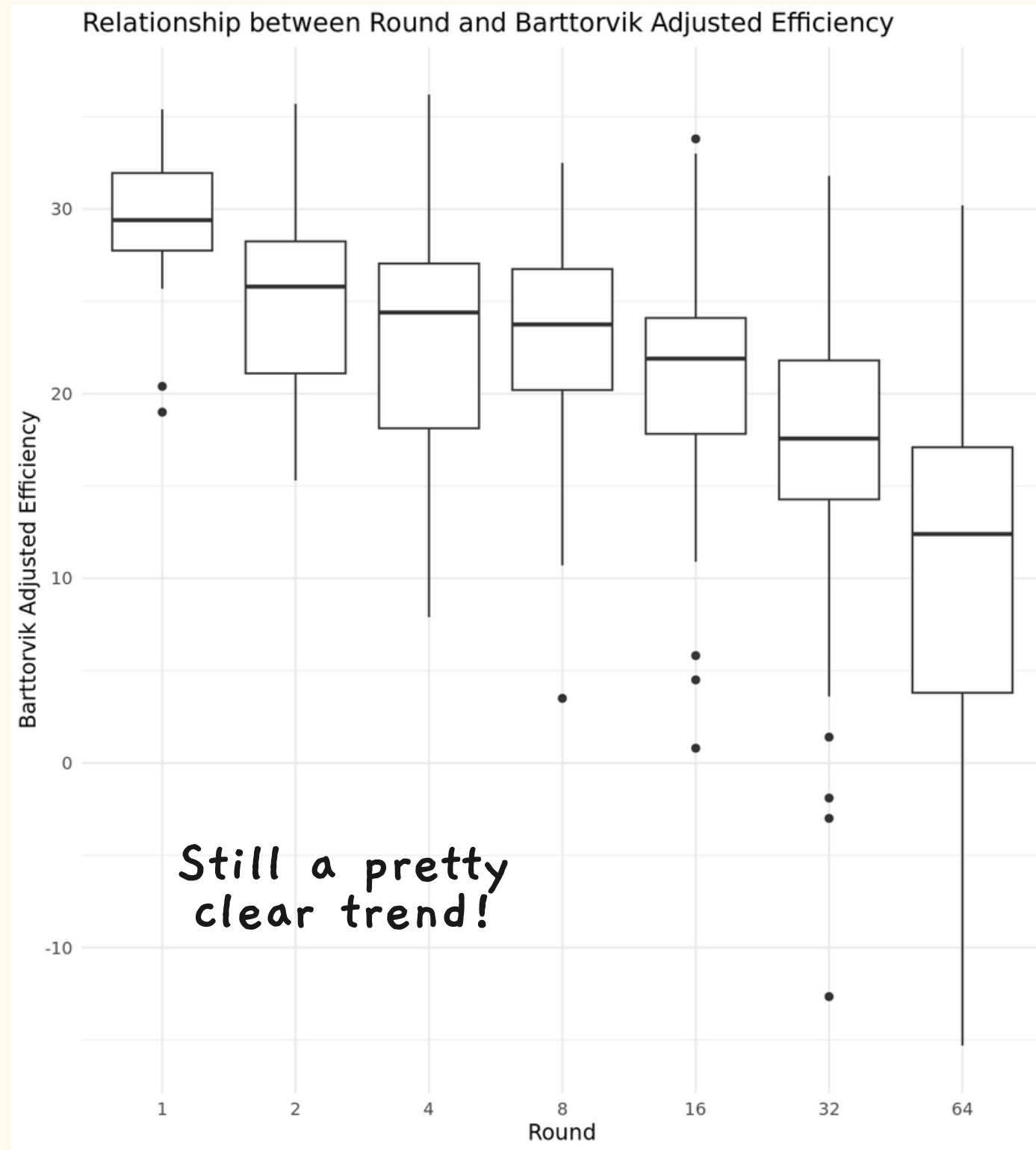


ROC

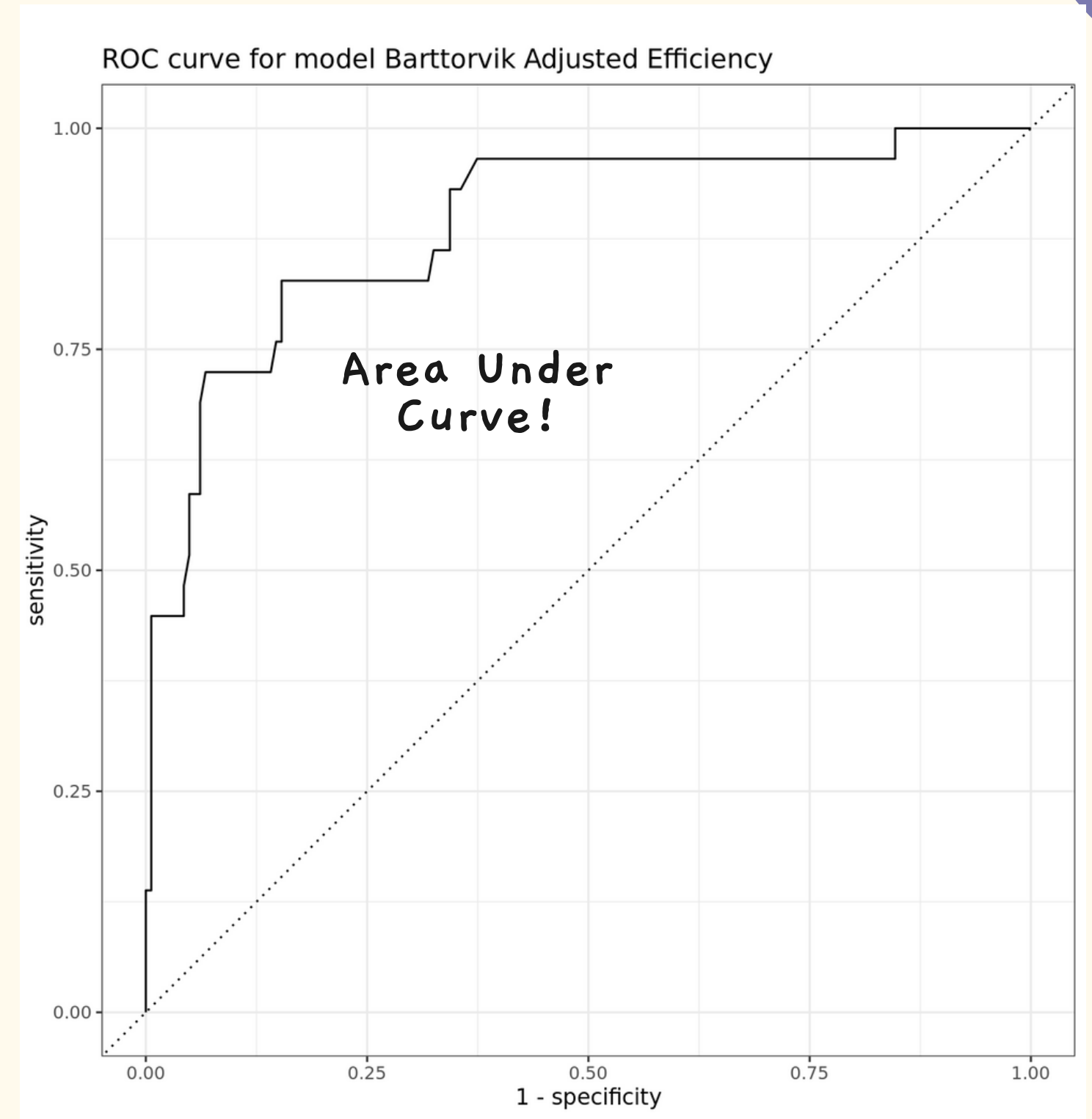


# BARTTORVIK ADJUSTED EFFICIENCY

EDA



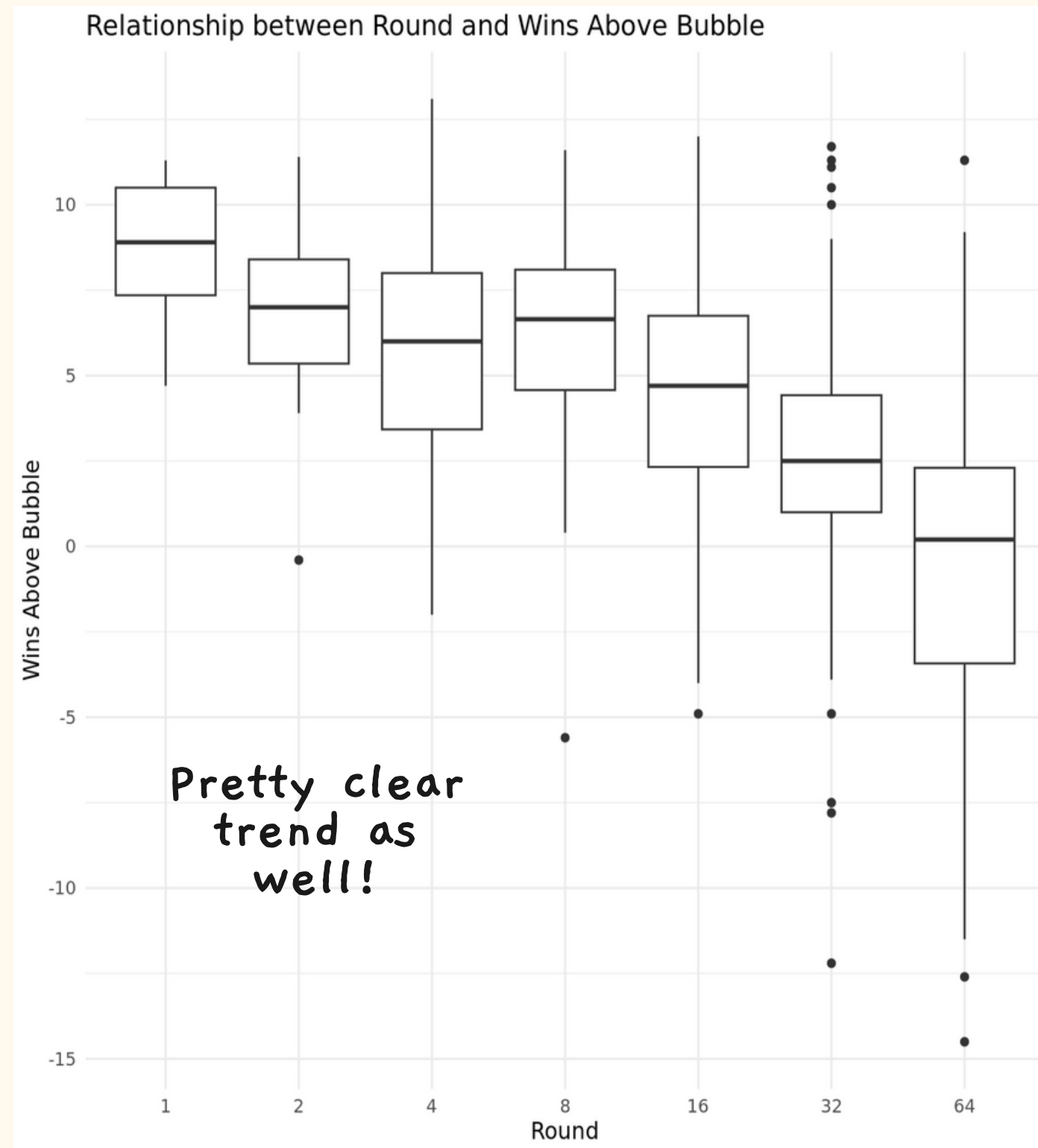
BOXPLOT



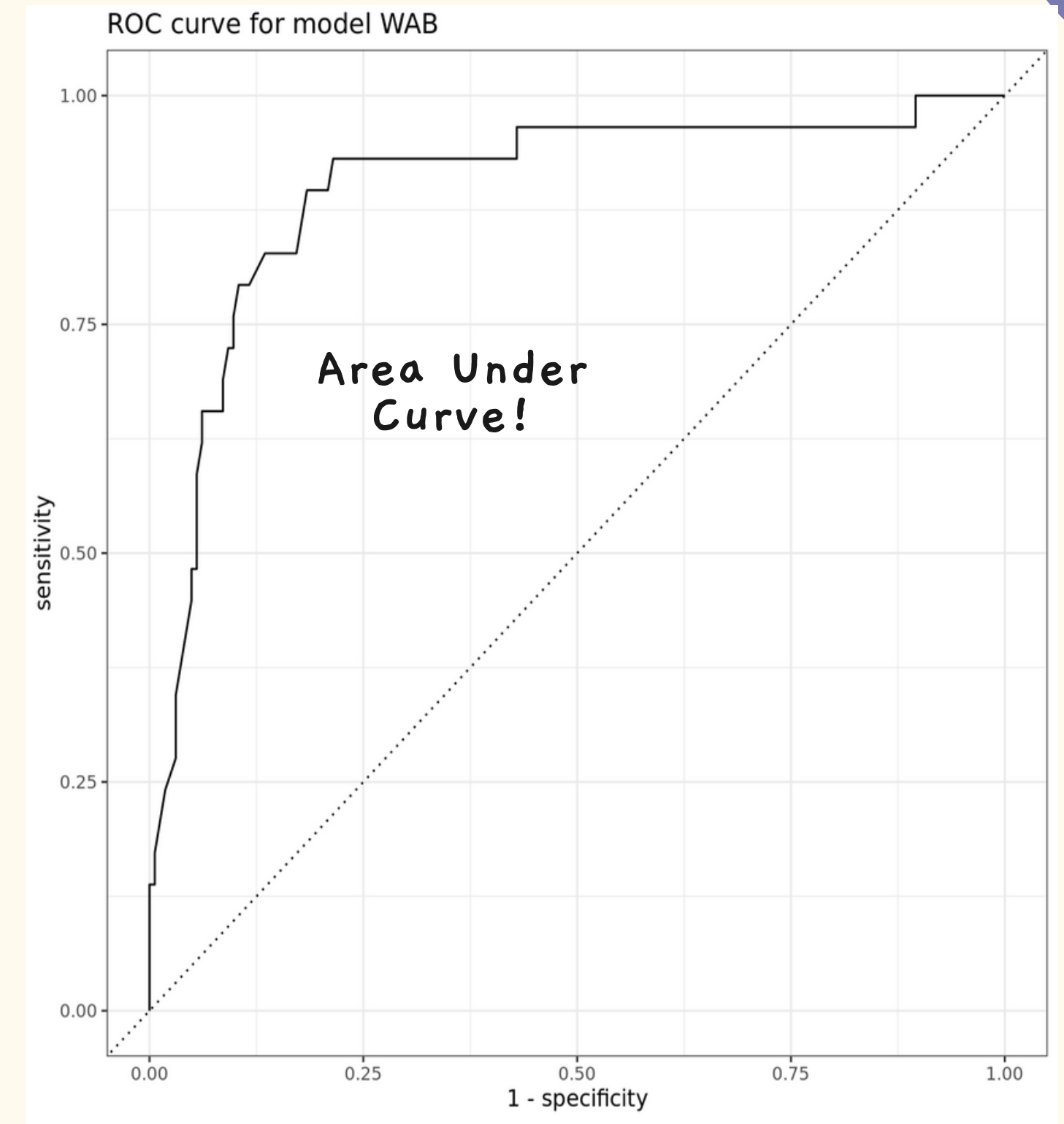
ROC

# WINS ABOVE BUBBLE

EDA



BOXPLOT



ROC



# RESULTS

## SINGLE VARIABLE MODELS

2

### BARTTORVIK ADJUSTED EFFICIENCY

Equation:

$$\log(\hat{p}/(1 - \hat{p})) = 6.09 - 0.208 * b\_adj\_eff$$

AUC Value:

0.891

1

### BARTHAG MODEL

Equation:

$$\log(\hat{p}/(1 - \hat{p})) = 20.92 - 21.32 * BARTHAG$$

AUC Value:

0.894

3

### WINS ABOVE BUBBLE

Equation:

$$\log(\hat{p}/(1 - \hat{p})) = 3.46 - 0.374 * wins\_above\_bubble$$

AUC Value:

0.898

# RESULTS

## TWO VARIABLE MODELS

1

**BARTHAG + WAB**

**Equation:**

$$\log(\hat{p}/(1 - \hat{p})) = 10.07 - 7.93 * BARTHAG - 0.26 * wins\_above\_bubble$$

**AUC Value:**

0.903

2

**BARTHAG + BAE**

**Equation:**

$$\log(\hat{p}/(1 - \hat{p})) = 4.13 + 2.80 * BARTHAG - 0.23 * b\_adj\_eff$$

**AUC Value:**

0.890

3

**WAB + BAE**

**Equation:**

$$\log(\hat{p}/(1 - \hat{p})) = 5.07 - 0.20 * wins\_above\_bubble - 0.12 * b\_adj\_eff$$

**AUC Value:**

0.901

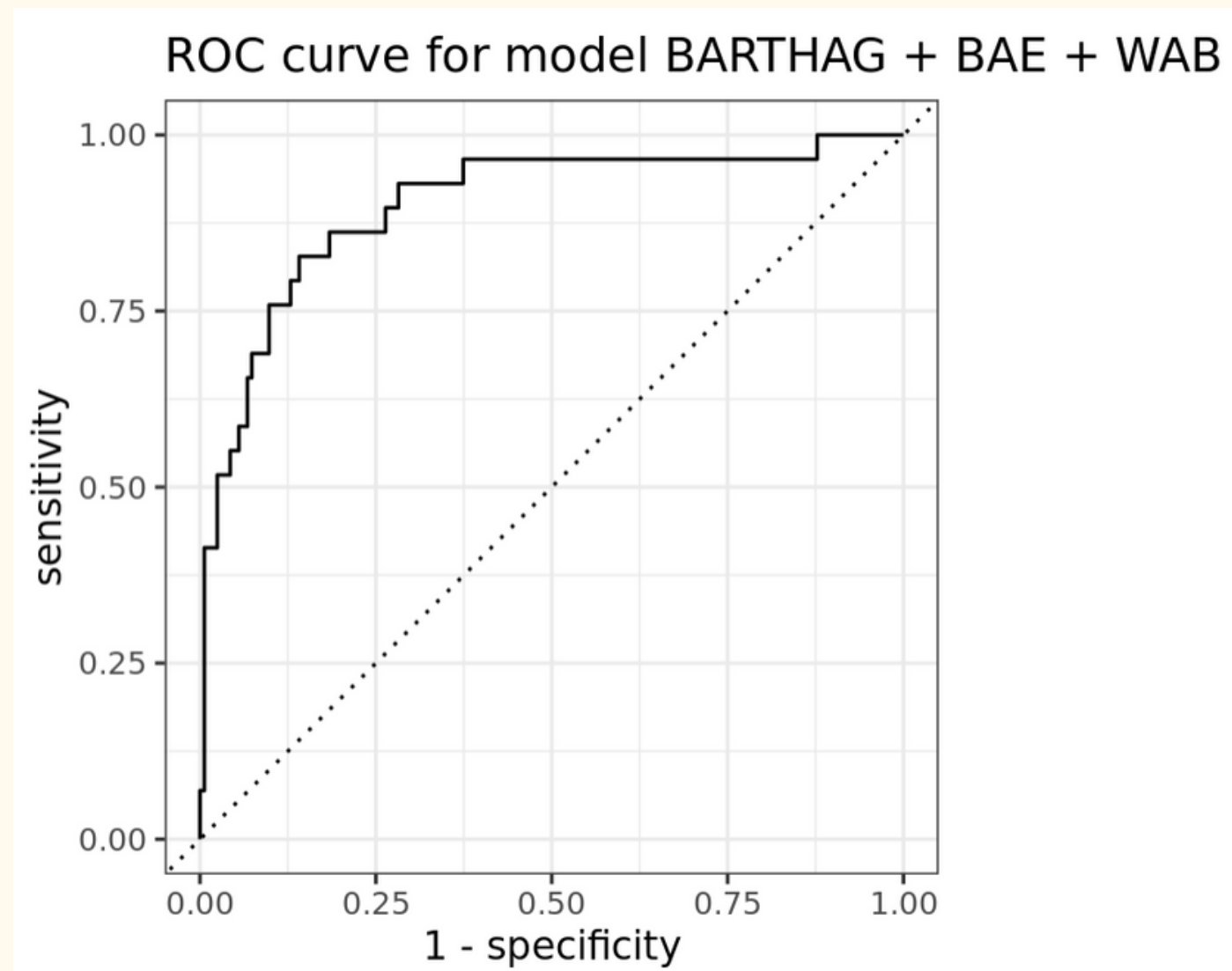
# RESULTS

## FULL MODEL

Equation:

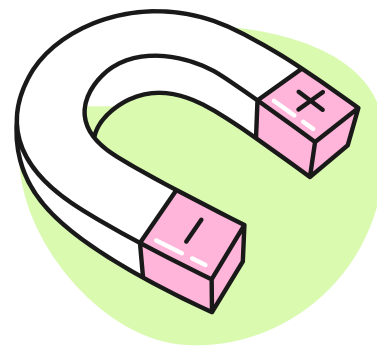
$$\log(\hat{p}/(1 - \hat{p})) = 2.86 + 3.15 * BARTHAG - 0.20 * wins\_above\_bubble - 0.14 * b\_adj\_eff$$

AUC:  
0.900



## COMMENTS

All models, no matter what variables used, performed relatively the same when comparing ROC / AUC



## CONCLUSION

Though all models performed roughly the same, we gave our best model the data for the 2023 March Madness tournament and asked it to predict which teams would make the Elite Eight. The model got 2/8 of the teams correctly—Texas and Gonzaga. Our model is far from perfect, which is to be expected given how volatile the NCAA tournament can be.

pred_elite eight <dbl>	.pred_not elite ei... <dbl>	elite_eight <fctr>	team <chr>
0.5768321	0.4231679	not elite eight	Alabama
0.4895942	0.5104058	not elite eight	Kansas
0.4497347	0.5502653	not elite eight	Purdue
0.4368582	0.5631418	not elite eight	Houston
0.4116876	0.5883124	not elite eight	UCLA
0.3586346	0.6413654	elite eight	Texas
0.3219712	0.6780288	not elite eight	Arizona
0.3077924	0.6922076	elite eight	Gonzaga



**DIGITAL MARKETING**

**THANK  
YOU**

We hope you enjoyed!  
~ heartemoji <3

