

# What You Can Do

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Nick Eubank

## **Modeling**

Developing Model to Faithfully Represent Data

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## **Inference**

Interpreting Model Parameters  
for Application

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## **Inference**

Interpreting Model Parameters  
for Application

1. Are my estimates causal?
2. Will they generalize?
3. Are they biased?



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- Articulate causal questions,
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Developing, along the way, great portfolio pieces!

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And all in the middle of a pandemic.



erika  
@brownbearika



Maybe...just maybe.... the real treasure...was the friends  
we made along the way... (end scene)



I hope you feel really proud of everything you've accomplished.