What You Can Do

Nick Eubank

Modeling

Developing Model to Faithfully Represent Data

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Inference

Interpreting Model Parameters
for Application

Modeling

Developing Model to Faithfully Represent Data

$\Downarrow$

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1. Are my estimates causal?

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2. Will they generalize?

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Developing Model to Faithfully Represent Data

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Inference

Interpreting Model Parameters
for Application

1. Are my estimates causal?
2. Will they generalize?
3. Do they reflect biases in an unethical way?

Day 1

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- Be able to critically evaluate causal evidence collected by others,
- Articulate causal questions,
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  Developing, along the way, great portfolio pieces!

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All while getting used to dealing with people again!

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.