# Communicating Your Data to Statisticians





BERDC Special Topics Talk 4



# DacCota Dakota cancer collaborative ON TRANSLATIONAL ACTIVITY

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Biostatistics, Epidemiology, and Research Design Core



# Introduction



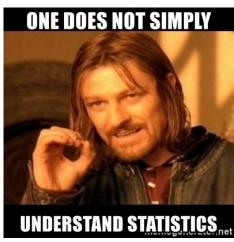
Goal: Improve communication between researchers and statisticians

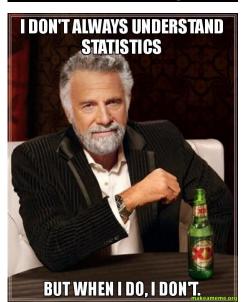
A pragmatic approach to statistics:

- Do I understand what the statistical method is doing?
- Is it interpretable?
- Is it comparable to other studies?
- Can I justify and defend it?

Bridging the gap between experiment and analysis:

- A major obstacle is the move from the specific (researcher) to the general (statistician)
- The target outcome is to train researchers to generalize their data so that statisticians can grasp it quickly and be able to understand the specifics of it
- The approach I'll be using is what I call "The 4 Big Topics"







# The 4 Big Topics



#### **T1: Big Picture Goal**

- Sum up your experiment in a few, clear sentences
- Sum up your experiment in terms of Y and X variables

#### T2: General Test Type

- Style of test (t-test, ANOVA regression, etc.)
- Discrete results (tables/graphs) expected

#### **T3: Variables Information**

- Variable name and type
- Variable measurement info

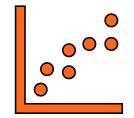
#### **T4: Sample Data**

- Provide preliminary or mock data
- Show report of similar experiments

## "Explanation"

Y-variable=X-variables

## Test-style



	Var 1	Var 2	Var 3
	Num.	Num.	Cat.
	Cont.	Discr.	3 Grps
	50-100	1-5	A,B,C

Var 1	Var 2	Var 3
<i>50.2</i>	1	A
<i>65.0</i>	2	В
75.7	5	С







#### 1. Sum up your experiment in a few, clear sentences

- Translate details into the big picture
- Abstracts work well here
- May be several sub-experiments to explain

## 2. <u>Sum up your experiment in terms of Y and X variables</u>

- Examples:
  - We wanted to model **Y-variable** as a function of **X-variable**(**s**).
  - We wanted to see if **X-variable(s)** predicted **Y-variable**.
  - We wanted to see if **Y-variable** was affected by **X-variable**(s).
- Can also add more information:
  - ... while correcting for **confounding-variable(s)**.
  - ... across time.
  - ... etc.

#### **Examples**

We wanted to model nematode growth rate (Y) as a function of gene copy number (X).

We wanted to see if the presence of nuclear power reactors in a state (X) predicted brain cancer incidence rates (Y).

We wanted to see if COVID-19 county death rates (Y) were affected by county rural-urban status (X)







#### 1. Determine the general style of test

- What type of test do you anticipate your experiment(s) to be?
  - Comparing means across 2 groups: T-test
  - Comparing means across 3+ groups: ANOVA
  - Comparing 2 or more numerical values: regression
  - Comparing frequencies across categories: Chi-Square
- More complicated or specific?
  - Generalized or mixed model?
  - Logistic, Poisson, or other regression?
  - Survival analysis?
  - Etc

### 2. Sketch out the expected resulting graph or table

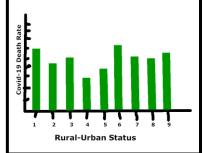
- Helps visualize what the outcome will be and therefore what will need to go into it
- Can be a simple as sketching out hand-drawn graphs and labeling the axes

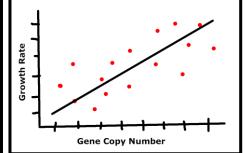
#### **Examples**

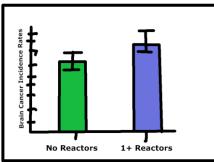
Modeling nematode growth rate as a function of gene copy number: **Regression** 

State nuclear power reactor status as a predictor for brain cancer incidence rates: **T-test** 

COVID-19 county death rates across rural-urban county type (9 groups): **ANOVA** 











# T3: Variable Information

#### 1. Write down the variable names and types

- Provide full name rather than abbreviation
  - LBXWBCSI [unhelpful] -> White blood cell count [helpful]
  - Include units if possible-> White blood cell count (1000 cells/uL)
- Numerical or categorical:
  - Numerical: continuous or discrete?
  - Categorical: ordinal or nominal?
- Other considerations:
  - Controlling for confounding variables or random effects?

#### 2. Write down the variable measurement info

- Numerical: central tendency, spread, and distribution
- Categorical: number of groups, group size

### **Important point**: need to translate from method to measurement

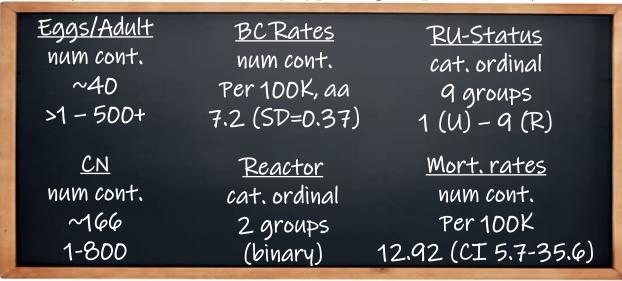
• Ex. 'running a Western Blot' -> translates to generating protein concentrations in (μL)

#### **Examples**

Modeling nematode growth rate as a function of gene copy number

State nuclear power reactor status as a predictor for brain cancer incidence rates

COVID-19 county death rates across ruralurban county type (9 groups)









#### 1. Provide preliminary or mock data

- Best would be preliminary, pilot, sample data
- Doesn't need to be full dataset, just enough to get a feel for the analysis
- Also good for power analysis
- If not, could also try to create mock data through simulation
- <a href="https://aosmith.rbind.io/2018/08/29/getting-started-simulating-data/">https://aosmith.rbind.io/2018/08/29/getting-started-simulating-data/</a>

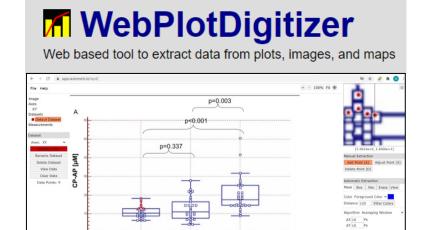
#### 2. Show report of similar experiments

- Previous experimental design and results
- Published papers (proposal references work well)

#### **Simulation Example**

```
Y <-rnorm(n=100, mean=20, sd=5)
Y <-round(Y,2)
X <-rpois(n=100, lambda=15)
C <-c(rep('A',25), rep('B',25), rep('C',25), rep('D',25))
sim_data <-data.frame(Y,X,C)
View(sim_data)</pre>
```

#### **Data Extraction Example**



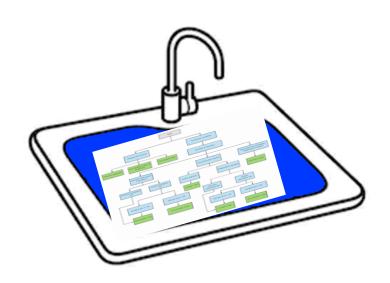
_	<b>Y</b>	<b>X</b>	<b>c</b> ‡
1	13.14	18	Α
2	19.82	19	Α
3	16.81	15	Α
4	17.40	11	Α
5	22.34	12	Α
6	19.61	19	А
7	16.41	19	Α
8	27.08	14	Α
9	30.76	19	Α
10	9.97	21	А



## Other Considerations



- More involved experiments will have more information to consider
  - Is time involved (ex. repeated measures)?
  - Are there fixed (age, sex, etc.) or random effects (plot, batch, etc.) you want to account for?
  - Is there missing data?
  - Anything else that would make the analysis harder?
- Better to have the researcher start to answer the questions than have the statistician try to answer them with less intimate knowledge of the research
- Don't have to have all the answers
- Useful procedure is to start with a basic setup and add complexity from there





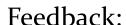
## What's Next



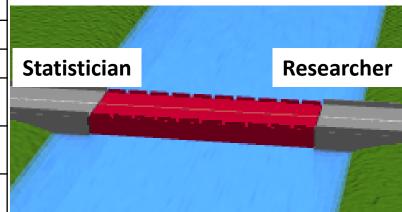
#### Closing points

- 4 Big Topics helps statistician understand research project and helps researcher understand, frame, and communicate their research
- The clearer the design, the better the chance for successful outcome
- Well-defined data plan protects against failure due to faulty planning
- Researchers are encouraged to think about their data

	BERDC project support by priority		
	1	Career development awards (DaCCoTA CTR)	
	2	Pilot projects (DaCCoTA CTR)	
	3	Resident/trainee research	
	4	Federally funded research	
e	5	General consultation on research design and methods	
	6	Grant application support for clinical research	
	7	Industry-funded research and clinical trials	
	8	Manuscript support	
	9	Other funded research	
	10	Privately funded research	
	11	Unfunded research	



- Survey: <a href="https://und.qualtrics.com/jfe/form/SV\_exoFvxRYpkddzhz">https://und.qualtrics.com/jfe/form/SV\_exoFvxRYpkddzhz</a>
- Questionnaire: <a href="https://und.qualtrics.com/jfe/form/SV">https://und.qualtrics.com/jfe/form/SV</a> dcyUSPLhD4cmP5Q









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