Graphs: The art of designing information

"A picture tells a thousand words"

- Lake Blanche

1



2

General definition of a graph

- Visual representation of a relationship between two or three variables (and more sometimes).
- Variables can be of any type (e.g., categorical or numerical).
- They commonly consist of two axes: x-axis (horizontal or abscissa) and y-axis (vertical or ordinate).







Why graphs?

- Powerful way of summarizing data that is easy to read (i.e., quick and direct).
- Highlight the most important information (i.e., facilitate communication).
- Facilitate (summarize) data understanding.
- Help convince others.
- Easy to remember (general trends).
- Aid in detecting unusual features in data.
- Tell stories.













BAR GRAPH: Two categorical variables (often from a contingency table)									
Is reproduction risky to health?									
	Control group	Egg-removal group	Row total						
Malaria	7	15	22						
No Malaria	28	15	43						
Column total	35	30	65						
Parus major	Treatment (egg removal/control) & outcome (malaria – yes/no) - categorical Frequency - numerical (discrete) Female birds put more energy in generating eggs to make up for those removed, thus reducing energy allocation towards immunocompetence.								



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Explanatory versus Response variables

- One major use of BioStatistics is to *relate* one variable to another, by examining associations between variables or differences between groups.

- When association between two variables is investigated, a common goal is to assess how well one of the variables, deemed the *explanatory* variable, *predicts* or *affects* (explain) the other variable, called the *response* variable.

16

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"Assumed" explanatory power may depend on the type of study:[1] experimental versus [2] observational studies





Explanatory and response variables (experiment)

When conducting an experiment (e.g., malaria study in the last slides), the treatment variable (the one manipulated by the researcher) is the *explanatory* variable, and the measured effect of the treatment is the *response* variable.



19



20

"Assumed" explanatory power may depend on the type of study

Observational study - Researchers have no control over which observational units fall into which treatment or values of the explanatory variable. Examples:

- Studies on the health consequences of cigarette smoking in humans (unethical to assign smoking and no-smoking treatments to observational units, i.e., people).
- Growth of fish in warm versus cold lakes (observational units, i.e., fish are already in lakes; the research has no control on which fish goes in which lake).















Independent versus dependent variables = explanatory versus response variables, respectively

Strictly speaking, if one variable depends on the other, then neither is independent, so we rather say *explanatory* and *response*.

Sometimes you will hear variables referred to as "*independent*" and "*dependent*". These are the same as *explanatory* and *response* variables, respectively.

26

Independent versus dependent variables = Explanatory versus response variables, respectively

Strictly speaking, if one variable depends on the other, then neither is independent, so we rather say *explanatory* and *response* (e.g., in Whitlock and Schluter).

Sometimes you will hear variables referred to as "*independent*" and "*dependent*". These are the same as *explanatory* and *response* variables, respectively.









































Some "rules" of Data visualization

40

How to make a *bad* plot:

- 1. Hide the data.
- 2. Make patterns hard to see.
- 3. Display magnitudes dishonestly.
- 4. Draw graphics unclearly.

How to make a *good* plot:

- 1. Show the data.
- 2. Make patterns easy to see.
- 3. Display magnitudes honestly.
- 4. Draw graphics clearly.

41

Mistakes in displaying data Mistake 1. Hide the data

Mistake 1: Hide the data

How to hide data:

• Provide only statistical summaries (e.g., means).

How to reveal data:

• Present all data points, while allowing all to be seen.

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43















Mistakes in displaying data Mistake 2. Making patterns hard to see

47

Mistake 2: Making Patterns Hard to See

How to hide patterns:

- Make one plot and call it good.
- Use unreasonable scales.
- Arrange factors nonsensically.

How to reveal patterns:

- Explore multiple potential plots.
- Use appropriate scales.
- Arrange factors meaningfully. Arrange in order for ordinal, by mean for nominal.

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52

