## Math 2311

Bekki George - bekki@math.uh.edu
Office Hours: MW 11am to 12:45pm in 639 PGH Online Thursdays 4-5:30pm

And by appointment
Class webpage: http://www.math.uh.edu/~bekki/Math2311.html

## 5.5 - Non-Linear Methods

Many times a scatter-plot reveals a curved pattern instead of a linear pattern.
We can transform the data by changing the scale of the measurement that was used when the data was collected. In order to find a good model we may need to transform our $x$ value or our $y$ value or both.

In this example from section 5.4, we saw that the linear model was not a good fit for this data:

| Year | 1790 | 1800 | 1810 | 1820 | 1830 | 1840 | 1850 | 1860 | 1870 | 1880 |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
| People per square mile | 4.5 | 6.1 | 4.3 | 5.5 | 7.4 | 9.8 | 7.9 | 10.6 | 10.09 | 14.2 |
| Year | 1890 | 1900 | 1910 | 1920 | 1930 | 1940 | 1950 | 1960 | 1970 | 1980 |
| People per square mile | 17.8 | 21.5 | 26 | 29.9 | 34.7 | 37.2 | 42.6 | 50.6 | 57.5 | 64 |

Let's investigate other models.

## 5.6 - Relations in Categorical Data

A two-way table organizes the data for two categorical variables.
The totals of each row and column are considered marginal distributions because they appear in the margins of the table.

Example:
The following two-way table describes the preferences in movies and pizza toppings for a random sample of 100 people.

| Movie | Pepperoni | Hamburger | Mushrooms |  |
| :--- | :--- | :--- | :--- | :--- |
| Jurassic Park | 20 | 5 | 10 |  |
| Star Wars | 15 | 15 | 12 |  |
| Gone with the Wind | 8 | 2 | 13 |  |
|  |  |  |  |  |

Enter the marginal distributions in the table.
Draw a bar chart to display the marginal distribution of pizza topping preference.

What percent of our sample likes Gone with the Wind?

What percent of pepperoni lovers like Star Wars?

A conditional distribution is made up of the percentages that satisfy a given condition.
Compare the conditional distributions of movie preference for hamburger lovers and mushroom lovers. Back up your description with percentages.

Popper 12
The following two-way table describes the preferences in music genre and pizza toppings for a random sample of 100 people.

|  | Cheese | Peperoni | Mushrooms |
| :--- | :--- | :--- | :--- |
| Techno | 20 | 5 | 15 |
| Country | 8 | 12 | 11 |
| Rock | 14 | 2 | 13 |

2. What percent of the sample likes rock music?
3. What percent of cheese lovers like techno music?

Always be careful if combining data to make a comparison. Simpson's Paradox is the reversal of the direction of a comparison or an association when data from several groups are combined to form a single group.

This is adapted from Subsection 2.3.2 of A. Agresti (2002), Categorical Data Analysis, 2nd ed., Wiley, pp. 48-51.
In a 1991 study by Radelet and Pierce of the effect of race on death-penalty sentences, the following table was obtained tabulating the death-penalty sentences (Death) and non-death-penalty sentences (No death) in murder convictions in the state of Florida.

| Defendant's <br> race | Death | No death | Percent death |
| :--- | :--- | :--- | :--- |
| Caucasian | 53 | 430 | 11.0 |
| African- <br> American | 15 | 176 | 7.9 |

From this table, we see Caucasian defendants received the death penalty more often than AfricanAmerican defendants.

Now, we consider the very same data, except that we stratify according to the race of the victim of the murder. Below is the table.

| Victim's <br> race | Defendant's <br> race | Death | No death | Percent <br> death |
| :--- | :--- | :--- | :--- | :--- |
| Caucasian | Caucasian | 53 | 414 | 11.3 |
| Caucasian | African- <br> American | 11 | 37 | 22.9 |
| African- <br> American | Caucasian | 0 | 16 | 0.0 |
| African- <br> American | African- <br> American | 4 | 139 | 2.8 |

Here we see that when considering the cases involving Caucasian victims separately from the cases involving African-American victims, that the African-American defendants are more likely than Caucasian ones to receive the death penalty in both instances ( $22.9 \%$ vs $11.3 \%$ in the first case and $2.8 \%$ vs. $0.0 \%$ in the second case).

Example 3 (from text): A drug company tests two new treatments for an illness. In trial 1, drug $f$ cures 45 out of 200 of the patients with the illness and drug B cures 32 out of 200 patients with the illness. In trial 2, 100 patients with the illness are given drug A and 85 of them are cured. Drug B is given to 500 patients in trial 2 and 400 are cured.
A. Create a table for each trial and compare results. Which treatment would you conclude is better based on the data in the tables?

| Trial <br> 1 | Cur <br> ed | Tot <br> al | Perce <br> nt |
| :--- | :--- | :--- | :--- |
| Drug | 45 | 200 | 22.5 <br> A |
| Drug <br> B | 32 | 200 | $16 \%$ |


| Trial <br> 2 | Cur <br> ed | Tot <br> al | Perce <br> nt |
| :--- | :--- | :--- | :--- |
| Drug <br> A | 85 | 100 | $85 \%$ |
| Drug <br> B | 400 | 500 | $80 \%$ |

B. Put the data together into one table and calculate the percentage cured for the aggregated data. Which treatment would you conclude is better based on the data in this table?

| Trials 1 and 2 combined | Cured | Total | Percent |
| :--- | :--- | :--- | :--- |
| Drug A | 130 | 300 | $43.3 \%$ |
| Drug B | 432 | 700 | $61.7 \%$ |

## Popper 12

4. Simpson's Paradox occurs when
5. LSRL stands for:
