

UCLA STAT 13
**Introduction to Statistical Methods for
the Life and Health Sciences**

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**Chapter 3: Exploratory Tools for
Relationships**

Tools for assessing relationships between

- **Two** quantitative variables
- **A** quantitative and a qualitative variable
- **Two** qualitative variables

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Use **scatter plots** to explore relationships
between **quantitative variables**

Figure 3.1.1 Scatter plot of SYSVOL versus DIAVOL
for the heart-attack data in Table 2.1.1.

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Example: Deaths and radiation in milk after Chernobyl Accident

Region	Peak radioactivity in milk (picocuries/L)	Percentage increase in death rate
Middle Atlantic	23	2.2
South Atlantic	20	2.4
New England	22	1.9
East North-Central	29	3.9
West North-Central	32	3.6
East Southern	21	2.6
Central Southern	16	0
Mountain	37	4.2
Pacific	44	5

Figure 3.1.2 Chernobyl data.

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Example: Computer timings data

Number of terminals:	40	50	60	45	40	10	30	20
Time Per Task (secs):	9.9	17.8	18.4	16.5	11.9	5.5	11	8.1
Number of terminals:	50	30	65	40	65	65		
Time Per Task (secs):	15.1	13.3	21.8	13.8	18.6	19.8		

Figure 3.1.3 Computer timings data.

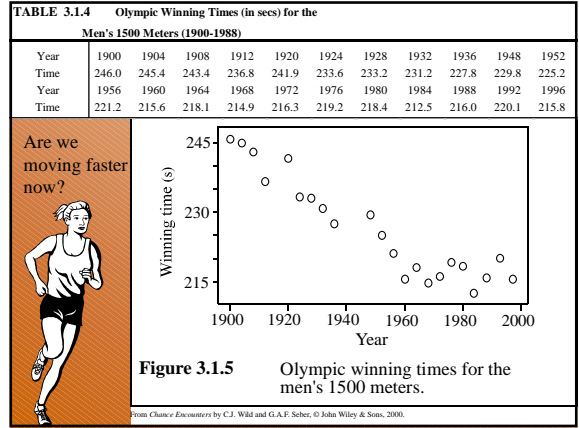
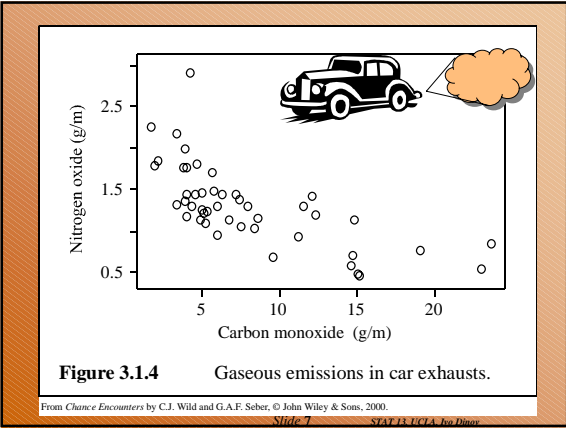
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Example: Car emissions

Car	HC	CO	NOX	Car	HC	CO	NOX	Car	HC	CO	NOX
1	0.50	5.01	1.28	17	0.83	15.13	0.49	32	0.52	4.29	2.94
2	0.65	14.67	0.72	18	0.57	5.04	1.49	33	0.56	5.36	1.26
3	0.46	8.60	1.17	19	0.34	3.95	1.38	34	0.70	14.83	1.16
4	0.41	4.42	1.31	20	0.41	3.38	1.33	35	0.51	5.69	1.73
5	0.41	4.95	1.16	21	0.37	4.12	1.20	36	0.52	6.35	1.45
6	0.39	7.24	1.45	22	1.02	23.53	0.86	37	0.57	6.02	1.31
7	0.44	7.51	1.08	23	0.87	19.00	0.78	38	0.51	5.79	1.51
8	0.55	12.30	1.22	24	1.10	22.92	0.57	39	0.36	2.03	1.80
9	0.72	14.59	0.60	25	0.65	11.20	0.95	40	0.48	4.62	1.47
10	0.64	7.98	1.32	26	0.43	3.81	1.79	41	0.52	6.78	1.15
11	0.83	11.53	1.32	27	0.48	3.45	2.20	42	0.61	8.43	1.06
12	0.38	4.10	1.47	28	0.41	1.85	2.27	43	0.58	6.02	0.97
13	0.38	5.21	1.24	29	0.51	4.10	1.78	44	0.46	3.99	2.01
14	0.50	12.10	1.44	30	0.41	2.26	1.87	45	0.47	5.22	1.12
15	0.60	9.62	0.71	31	0.47	4.74	1.83	46	0.55	7.47	1.39
16	0.73	14.97	0.51								

Source: Lorenzen [1980].
HC = hydrocarbons; CO=carbon monoxide; NOX = nitrogen oxides; grams/mile measurements; 46 identical vehicles tested.

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Review

- What is a **quantitative** variable?
- What basic **tool** is used for **exploring relationships** between quantitative variables?
- What is a **controlled** variable? (variables whose values are determined in the experimental design, as opposed to random variables who are evaluated once the experiments are conducted (e.g., number of terminals vs. task completion time))
- What is the difference between a random and a nonrandom variable? (variables whose values are not to be observed as random events during the experiment, i.e., these are controlled, deterministic or predictable variables, e.g., **year** for the Running Time experiment).

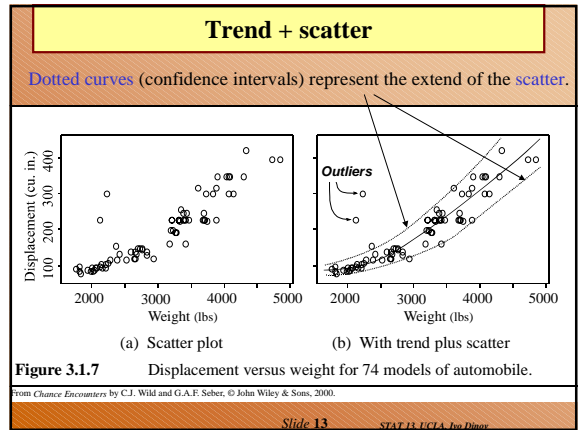
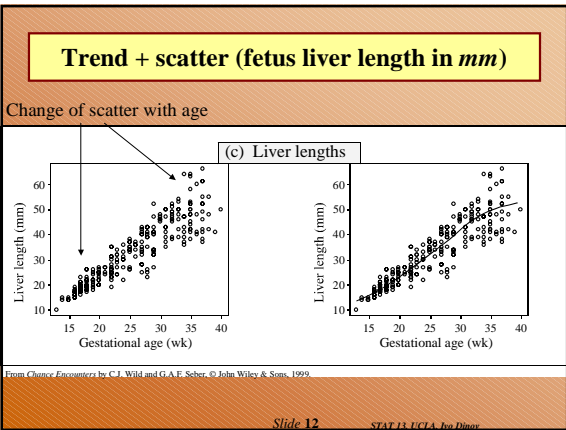
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Regression relationship = trend + residual scatter

(a) Sales/income

- **Regression** is a way of **studying relationships** between variables (random/nonrandom) for **predicting** or **explaining** behavior of 1 variable (**response**) in **terms of** others (**explanatory variables** or **predictors**).

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Outliers – odd, atypical, observations (errors, B, or real data, A)

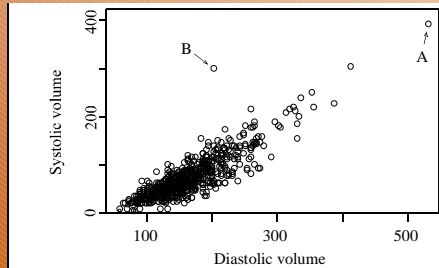


Figure 3.1.9 Scatter plot from the heart attack data.

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A weak relationship

58 abused children are rated (by non-abusive parents and teachers) on a psychological disturbance measure.

How do we quantify weak vs. strong relationship?

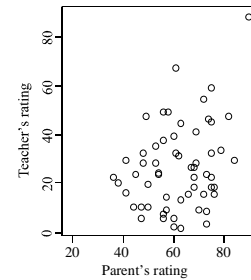
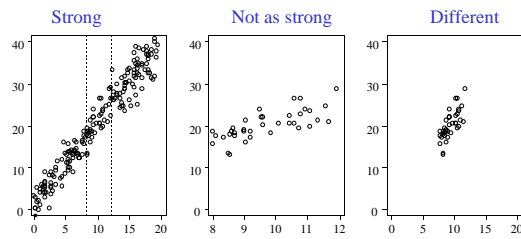


Figure 3.1.10 Parent's rating versus teacher's rating for abused children.

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Looking at plots



(a) Full data set (b) Middle portion (c) More white space
Figure 3.1.11 Visual impressions from scatter plots.

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Strong and weak relationships

- Plotting a strong relationship only on a small X-range will make the relationship much weaker (So, be ware sample size and sample representativeness do matter).
- The x-range scale and y-range scale need to be taken into account when investigating strong/weak relationships (extending or compressing any of the axes could significantly change the relation trend).

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Questions ...

- When people talk about plotting Y versus X , which variable is conventionally represented on the horizontal axis and which on the vertical axis?
- What are the roles of the response variable and the explanatory variable in regression?
- On a scatter plot, which axis is conventionally used for the explanatory variable and which for the response?

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Questions ...

- What are the two main components of a regression relationship?
- What do we call observations that are further from the trend curve than expected when compared with the usual level of scatter?
- Should outliers simply be discarded when analyzing data?

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Questions ...

- What should you immediately do when you identify an outlier?
- What makes some relationships look weak and others look strong?
- Under what circumstances can a strong relationship look weak in a scatter plot?
- What do we mean by association between two variables? (scatter plot trend that can not be explained by chance alone, implies the two variables are associated) A positive association? (If y and x are associated and y increases with x). A negative association? (If y and x are associated and y decreases with x).

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The prediction problem – can not predict a new response form a weak relationship

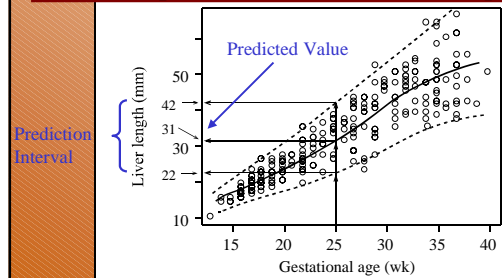


Figure 3.1.12 The prediction problem for liver-length data.

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Problems with prediction ...

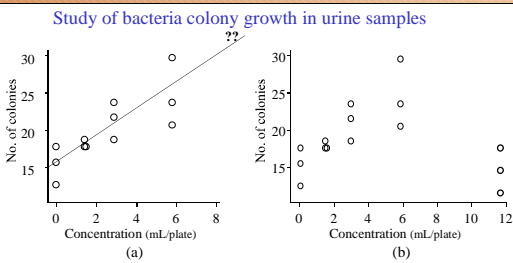


Figure 3.1.13 The dangers of predicting outside the range of the data. (Plotted from data in Margolin [1988]).

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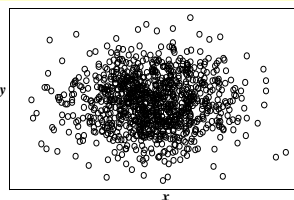
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Be very cautious when (extrapolating) predicting outside the range of the data.

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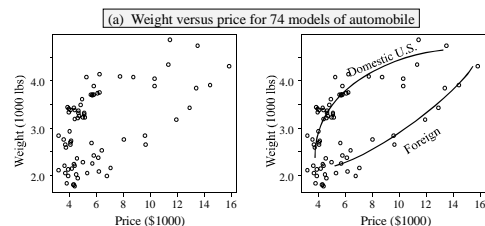
Questions ...

- Why can we not predict with any precision from a weak relationship?
- Under what circumstances can prediction be used with any confidence?
- Why is prediction outside of the range of the data dangerous?



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Other patterns

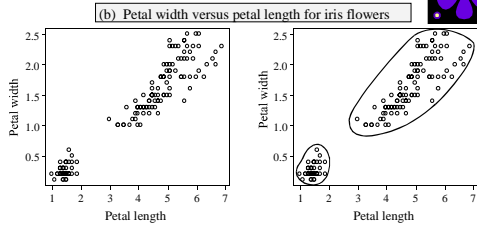
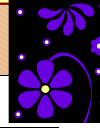


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Does mean, regression or prediction make sense in these situations? How to fix the problem? These are some of the issues we'll address later.

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Other patterns - clusters



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