

Statistically Speaking: *Correlation is not causation*

DAVID L. TABB, PH.D.

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Overview

- Causation is a difficult beast
- Essential concepts
- Magnitudes vs. ranks: Pearson vs. Spearman

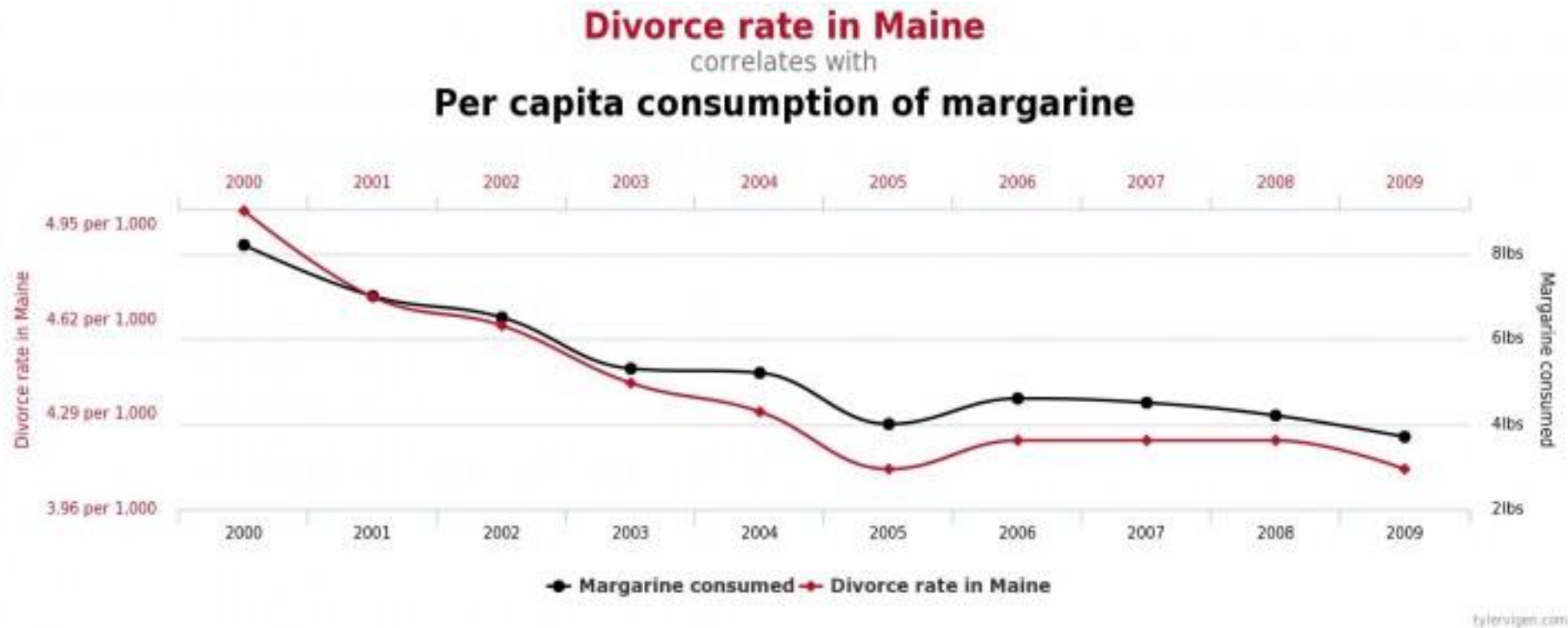
Austin Bradford Hill Causation

- Temporal Relationship
- Strength
- Dose-Response Relationship
- Consistency
- Plausibility
- Consideration of Alternate Explanations
- Experiment
- Specificity
- Coherence

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No, not even when the data lie on top of each other

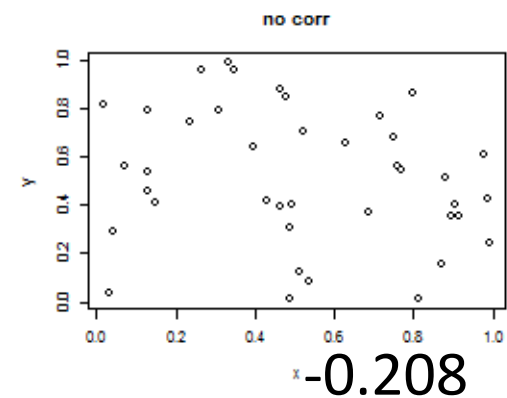
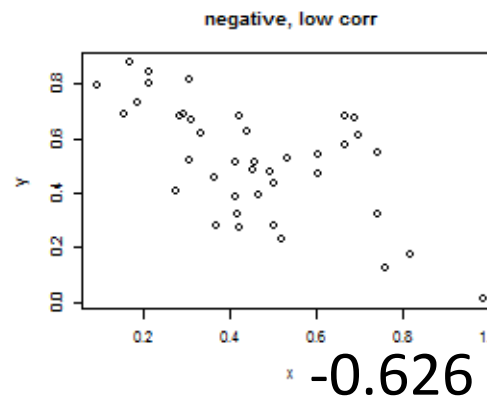
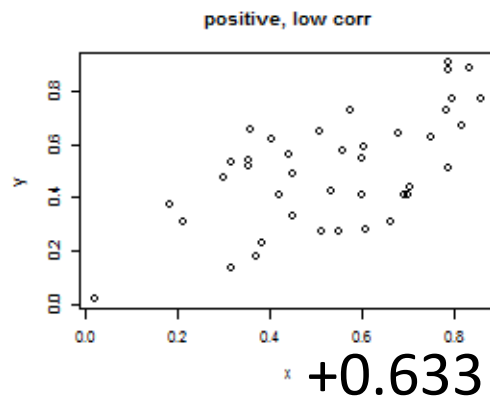
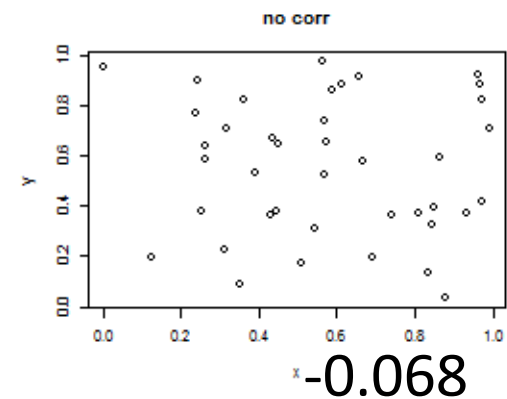
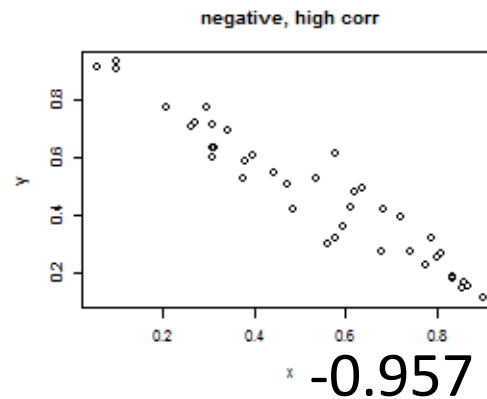
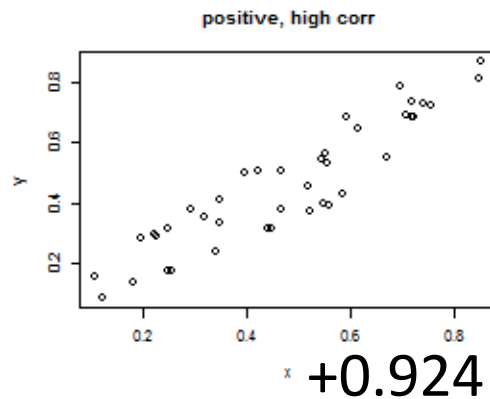


Correlation Concepts

- Positive: higher A values are associated with higher B values.
- Negative: higher A values are associated with lower B values.
- High: we observe a strong association between A and B.
- Low: we observe little association between A and B.

Correlation is commutative; $\text{cor}(a, b) = \text{cor}(b, a)$.

Examples from random simulations



Partial code for prior slide

```
#Set up for four graphs in one pane
par(mfrow=c(2,3))

#Positive, high correlation
a <- runif(40)
b <- runif(40)
x <- 0.6 * a + 0.4 * b
y <- 0.4 * a + 0.6 * b
plot(x,y,main="positive, high corr")
cor(x,y)
```

Karl Pearson: magnitudes matter

- Technically, $r = cov(x, y) / (sd(x) * sd(y))$
- “Covariance is a measure of how much two random variables vary together. It’s similar to variance, but where variance tells you how a single variable varies, covariance tells you how two variables vary together.”

<http://www.statisticshowto.com/covariance/>

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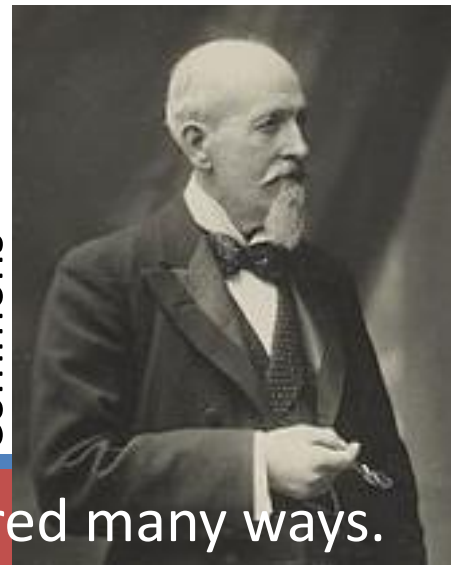


Sadly, he was also a eugenicist.

Charles Spearman: ranks matter

- Technically, $\rho = \frac{cov(ranks(x), ranks(y))}{sd(ranks(x)) * sd(ranks(y))}$
- The ranking of values within x or within y, rather than their magnitudes, drives the Spearman correlation.
- Dumping magnitude makes Spearman correlation robust against outliers.

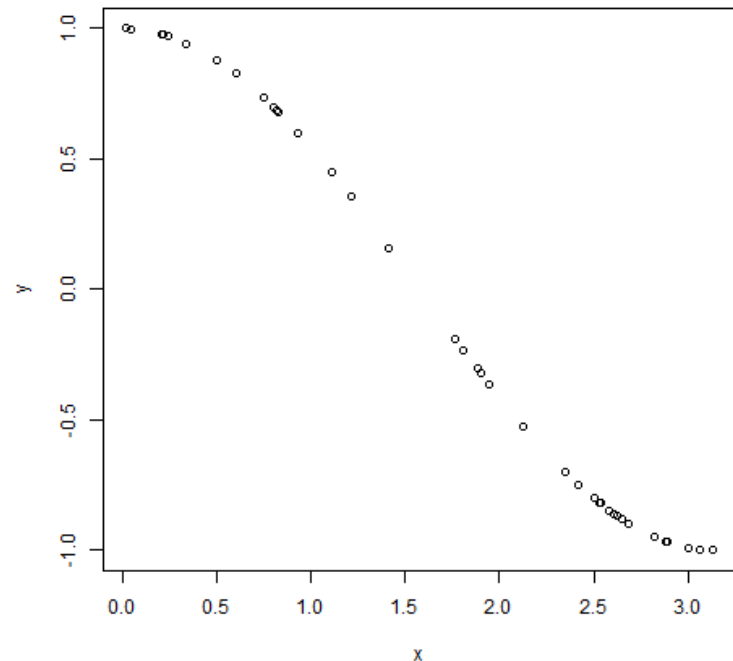
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Argued that general intelligence factor may be measured many ways.

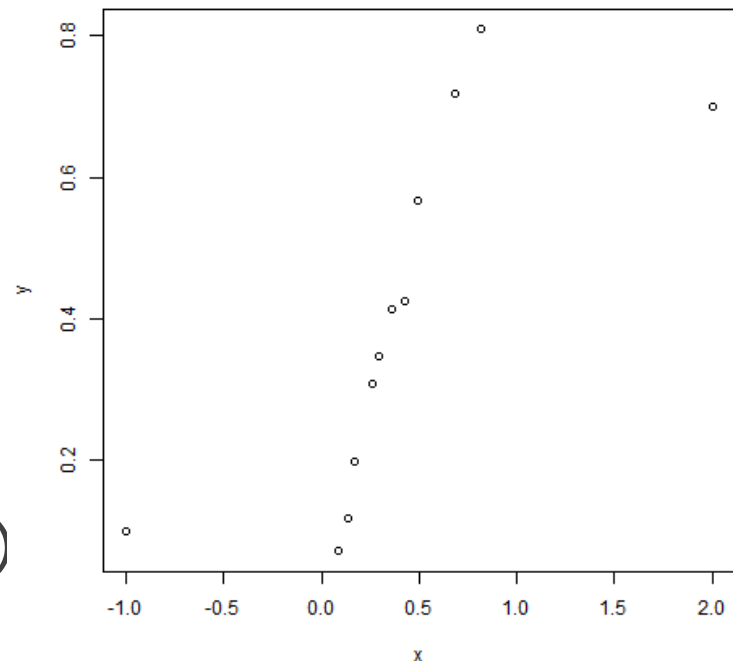
Same input, different outputs

```
x <- pi*runif(40)
y <- cos(x)
plot(x,y)
cor(x,y,
method="pearson")
[1] -0.9926561
cor(x,y,
method="spearman")
[1] -1
```



Resisting outliers

```
a <- runif(10)
b <- runif(10)
x <- c(2, -1,
(0.55*a+0.45*b))
y <- c(0.7, 0.1,
(0.45*a+0.55*b))
plot(x,y)
cor(x,y, method="pearson")
[1] 0.7461581
cor(x,y, method="spearman")
[1] 0.972028
```



Takeaways

- Correlation analysis lets us compare the values between two sets for association.
- The two sets are treated as independent of each other; neither is a function of the other.
- The non-parametric Spearman method is preferable when data contain outliers.