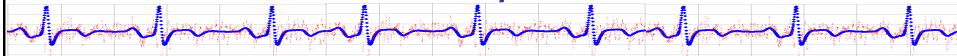


Empirical Research Methods in Information Science

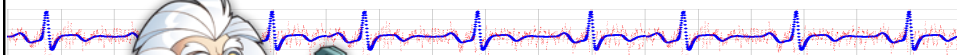
IS 4800 / CS 6350



Lecture 21
Non-parametric Stats

1

Experiment



Data transforms?

2

When to Use Non-parametric Stats

- DV is nominal
- DV is ordinal
- DV is interval or ratio and
 - Skewed
 - Outliers
 - Multi-modal
 - Generally, non-normal
- aka "distribution-free tests"
- Note: Non-Parametric Stats still have assumptions, but much less restrictive

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Non-parametric statistics

- Make few assumption about underlying population distribution.
- If your data does meet assumptions for parametric statistics (t-test etc), then non-parametric stats may be used.
- Provide less power than parametric equivalents
 - IF data is normal
- Generally operate on rank order-transformed or ranked data.
- A hodge-podge of techniques with no underlying theoretic framework.

4

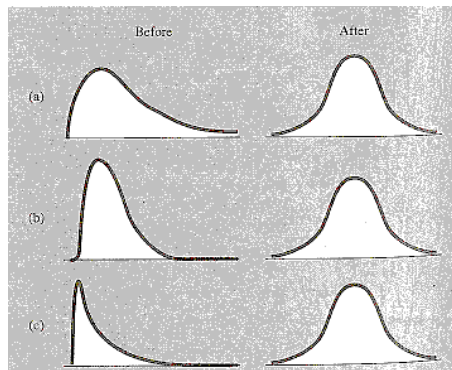
R: tests for normality

```
#Plot Q-Q line  
>qqline(DV)  
  
#Shapiro-Wilk Normality test  
#Probability my data came from a  
# normal distribution  
>shapiro.test(DV)
```

5

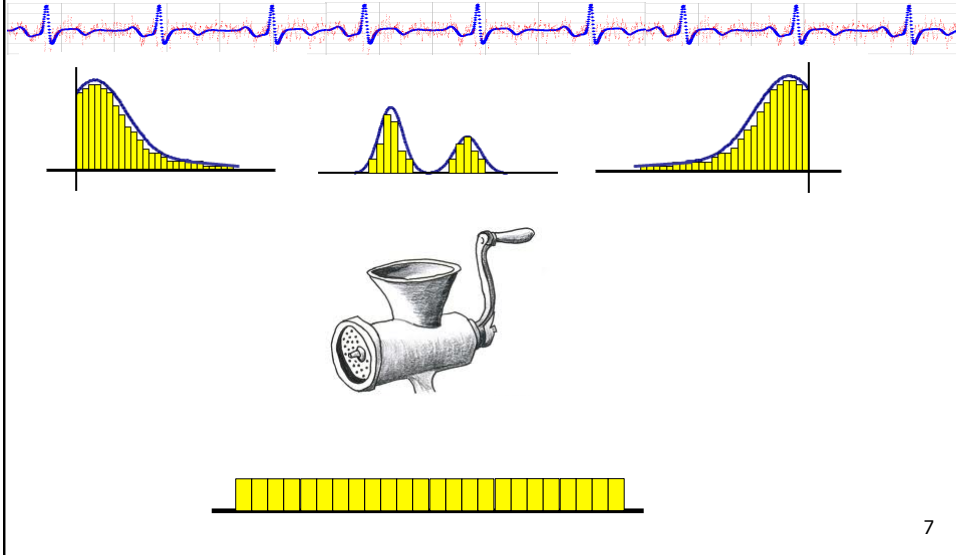
Other alternatives?

- Interval or Ratio DV, but skewed...



6

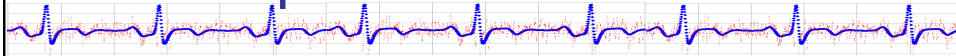
Rank-Order Transform



Rank-Order Transform

- Another kind of data transformation.
 - For non-uni-modal distributions or distributions with extreme outliers, or for ordinal measures
 - Transform
 - Take all data from a given measure (from all groups), sort and assign ranking (ignore ties).
 - Then, replace the original values with the corresponding rank.
- 8

Rank-Order Transform Example

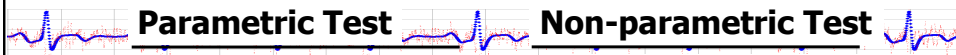


- Note For Ties: take average of ranks that would have been assigned
- DV Measure: Satisfaction
 - With sign in food court: 5.8, 7.1, 6.2, 6.7
 - Without sign: 2.3, 3.4, 1.7, 5.5
- Rankings:

1.7	2.3	3.4	5.5	5.8	6.2	6.7	7.1
1	2	3	4	5	6	7	8
- Resulting data for comparison:
 - With sign in food court: 5, 8, 6, 7
 - Without sign: 2, 3, 1, 4

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Non-parametric Tests



Parametric Test

Non-parametric Test

Pearson correlation

Chi-square test for independence

Spearman's rho
Just compute Pearson on the ranks

t test for independent means

Wilcoxon rank-sum test
aka
Mann-Whitney U test

t test for dependent means

Wilcoxon signed rank test

One-way ANOVA

Kruskal-Wallis H-test

Full-factorial ANOVA

Aligned Rank Transform**

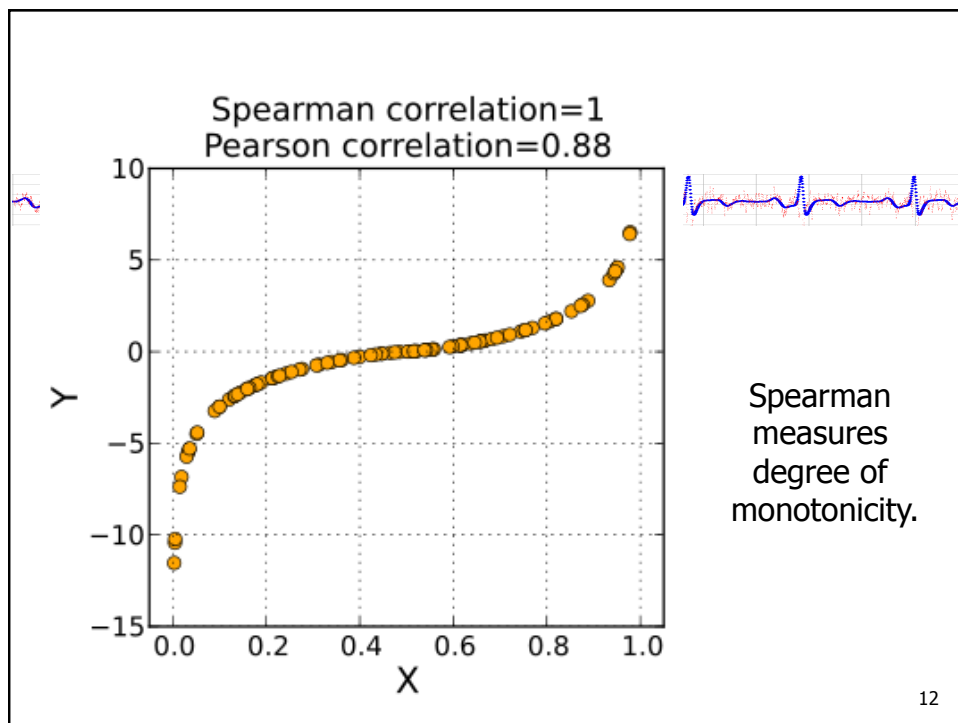
Note: Chi^{sq} on nominals, others on ordinal or better

χ^2 Test for Independence in R

```
#use contingency table
chisq.test(table(x,y))

#e.g., data$favcolor, data$ownmac
chisq.test(table(data$favcolor,
                  data$ownmac))
```

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Spearman's rho in R

```
> cor.test(x,y,method="spearman")
```

- Hyp test: likelihood we could have obtained a correlation this large, assuming H_0 (true correlation is 0).

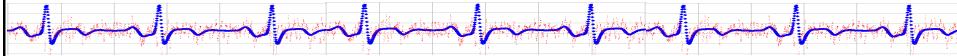
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Wilcoxon Rank Sum

- Two groups of independent measurements
 - Cf: t-test for independent means
- For each group
 - Compute $W =$
 - Sum of ranks for group measures
 - Minus "mean rank" for group = $\sum_{i=1}^N i$
(corrects for size of group)
 - Pick smaller W and test

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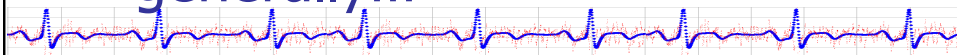
Wilcoxon Rank Sum (independent samples)



```
> wilcox.test(nonsmokers, smokers)
Wilcoxon rank sum test with
continuity correction
data: nonsmokers and smokers
W = 76.5, p-value = 0.04715
alternative hypothesis: true
location shift is not equal to 0
```

Report: $W=76.5$, $p<.05$

Wilcoxon in R more generally...



```
wilcox.test(dv-group1,  
            dv-group2)
```

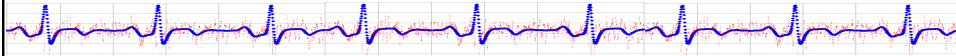
-or-

#or

```
wilcox.test(dv ~ iv-factor,  
            paired=FALSE/TRUE, data=d)
```


Effect size – for both indep & paired tests

from "Discovering Statistics Using R"



$$r = \frac{z}{\sqrt{N}}$$

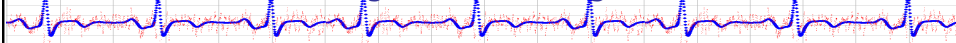
```
wilcoxModel <- wilcox.test(...)  
z <- qnorm(wilcoxModel$p.value/2)  
r <- z/sqrt(N)
```

Cohen: 0.1=Small, 0.3=Medium, 0.5=Large

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Sample report

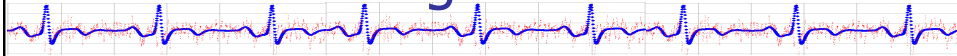
from "Discovering Statistics Using R"



- Depression levels in ecstasy users (Mdn = 17.50) did not differ significantly from alcohol users (Mdn = 16.00) the day after the drugs were taken, $W = 35.5$, $p = 0.286$, $r = -.25$. However, by Wednesday, ecstasy users (Mdn = 33.50) were significantly more depressed than alcohol users (Mdn = 7.50), $W = 4$, $p < .001$, $r = -.78$.

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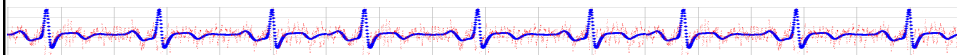
Wilcoxon signed rank



- Paired/dependent samples
 - Cf. paired sample t-test
- Compute difference for each pair of measures, discarding zeros
- Rank order transform the differences
- T_+ = Sum of ranks from positive diffs
- T_- = Sum of ranks from negative diffs
- T = smaller of T_+ and T_-
 - Note: R calls this "V"

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Wilcoxon signed rank in R



```
wilcox.test(dv-group1,  
            dv-group2, paired=TRUE)
```

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Wilcoxon Signed Rank Sample report

from "Discovering Statistics Using R"

- For ecstasy users, depression levels were significantly higher on Wednesday (Mdn = 33.50) than on Sunday (Mdn = 17.50), $p = .047$, $r = -.56$. However, for alcohol users the opposite was true: depression levels were significantly lower on Wednesday (Mdn = 7.50) than on Sunday (Mdn = 16.0), $p = .012$, $r = -.45$.

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Differences between several independent groups

- Kruskal-Wallis (cf. 1-way ANOVA)
 - Based on ranked data
- H_0 : all samples came from same underlying population
- H_1 : there is at least one difference
- "H" statistic based on ranks within each group

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Kruskal-Wallis in R

```
> kruskal.test(dv ~ ivfactor,  
              data=d)
```

```
#posthocs – Wilcoxon Rank sum on  
# all pairs of groups
```

```
> library(pgirmess)
```


```
> kruskalmc(dv ~ iv, data=d)
```

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Sample output from kruskalmc

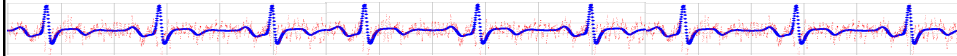
Multiple comparison test after Kruskal-Wallis, treatment vs control
(two-tailed)
p.value: 0.05
Comparisons

	obs.dif	critical.dif	difference
No Soya Meals-1 Soya Meal	2.2	15.63787	FALSE
No Soya Meals-4 Soya Meals	2.2	15.63787	FALSE
No Soya Meals-7 Soya Meals	19.0	15.63787	TRUE

Significant Difference?


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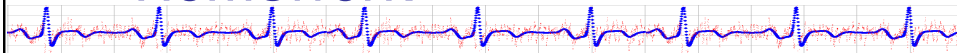
Non-parametric equivalent of multi-factor ANOVA



- J. Wobbrock, et al. 2011. The aligned rank transform for nonparametric factorial analyses using only anova procedures. *Proceedings of the SIGCHI Conference on Human Factors in Computing Systems (CHI '11)*.

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Homework



- Start T3
 - Experiment
 - Between or Within
 - At least 8 subjects.
 - Proposals due Friday.
 - 2 pages.
 - Include a power analysis. Should be prospective, based on a prior study.
- Read
 - Advanced study designs (B&A Ch 11).

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