

Non-Normality

Normality Tests and QQ-plots

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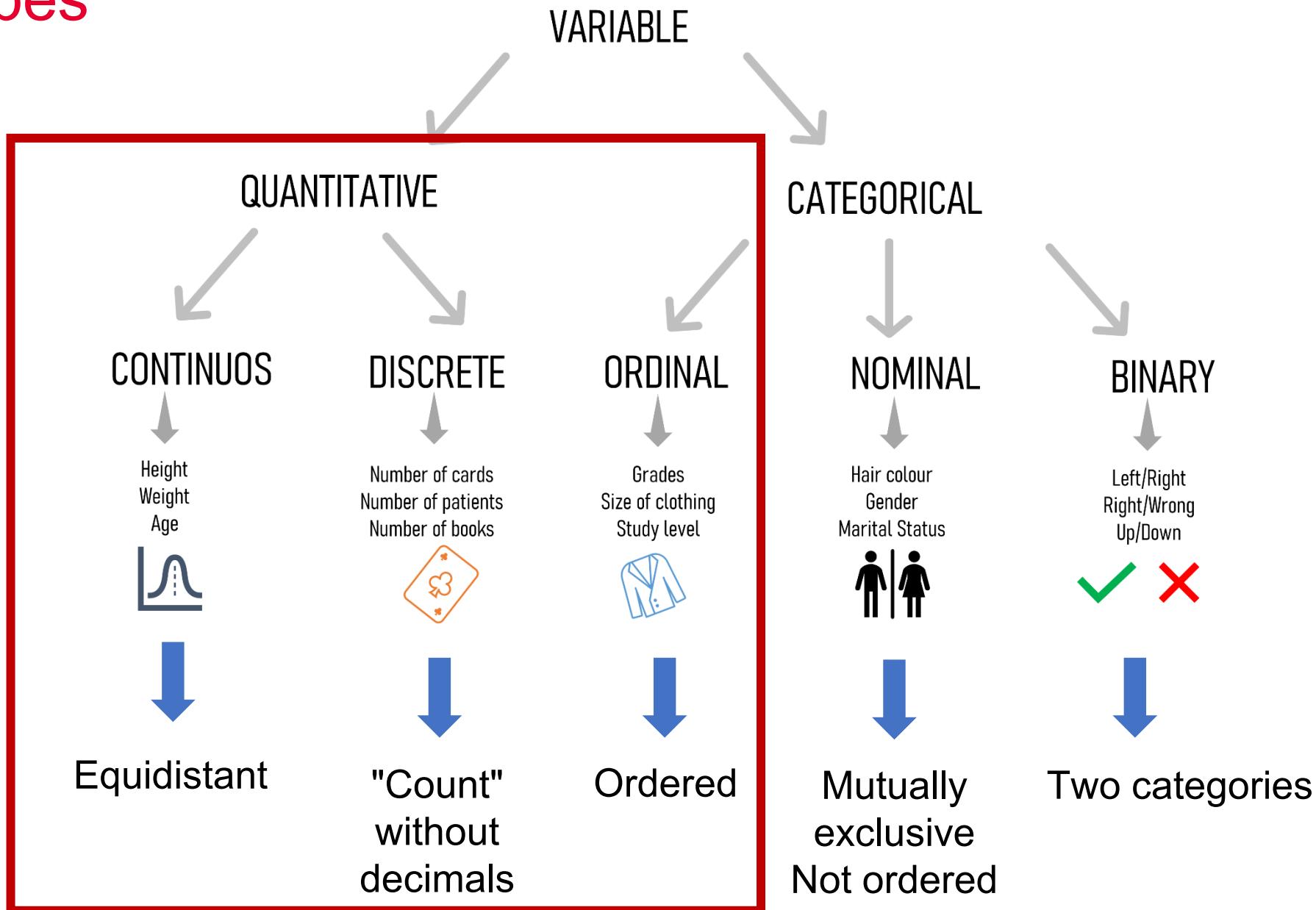
Objectives

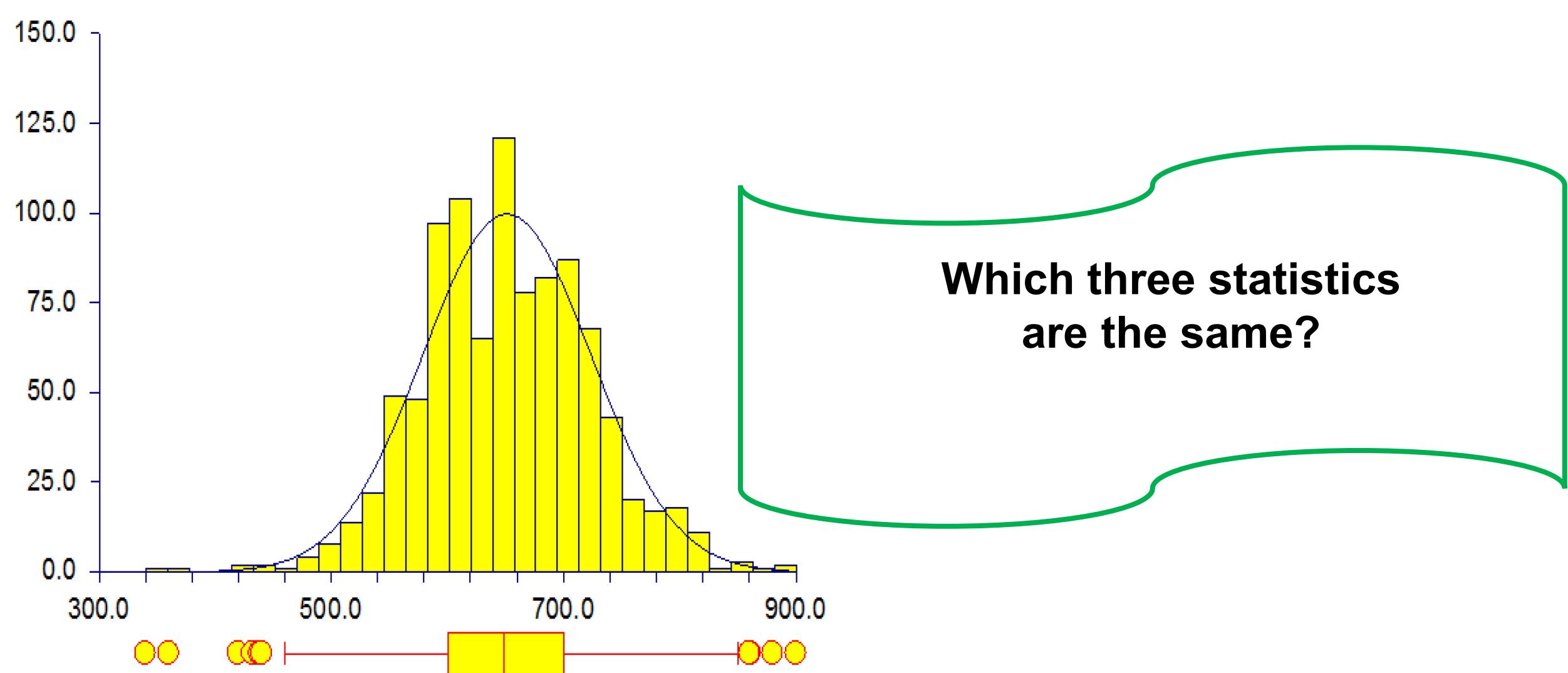
Identify non-Normal data visually and by testing for it

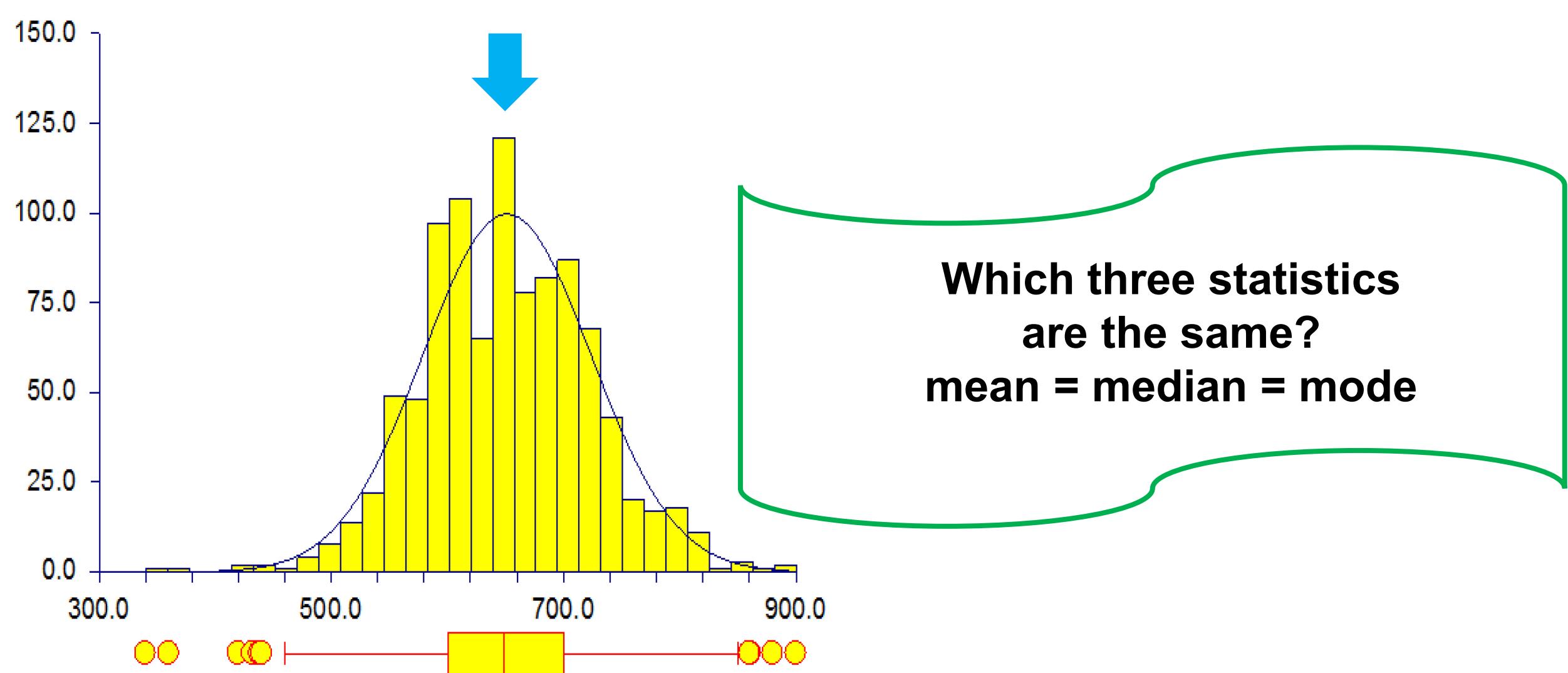
Compare non-Normal groups

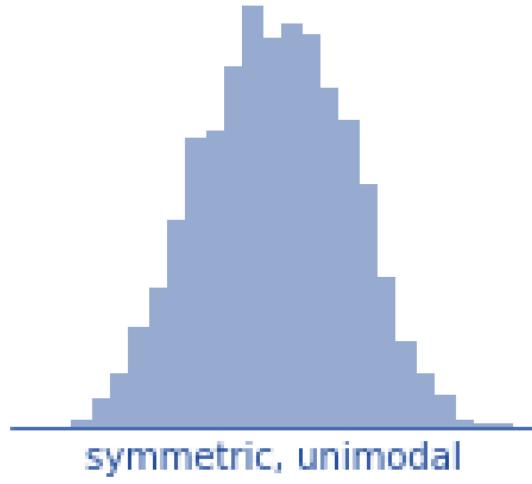
Multiple comparison corrections

Data Types

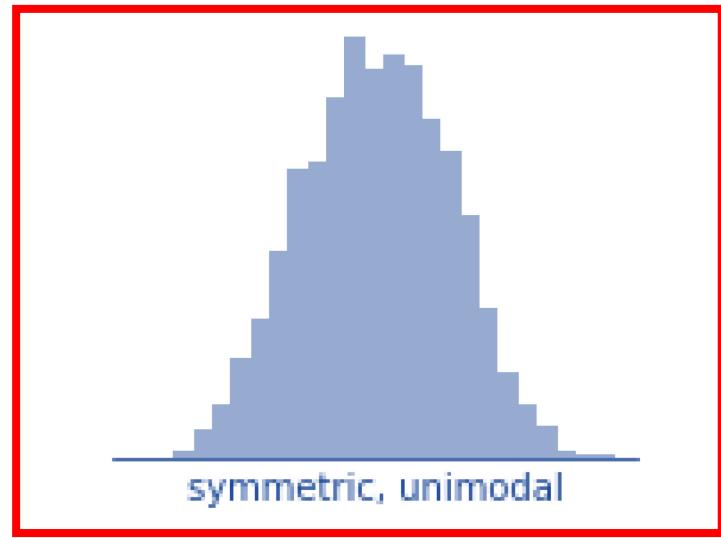




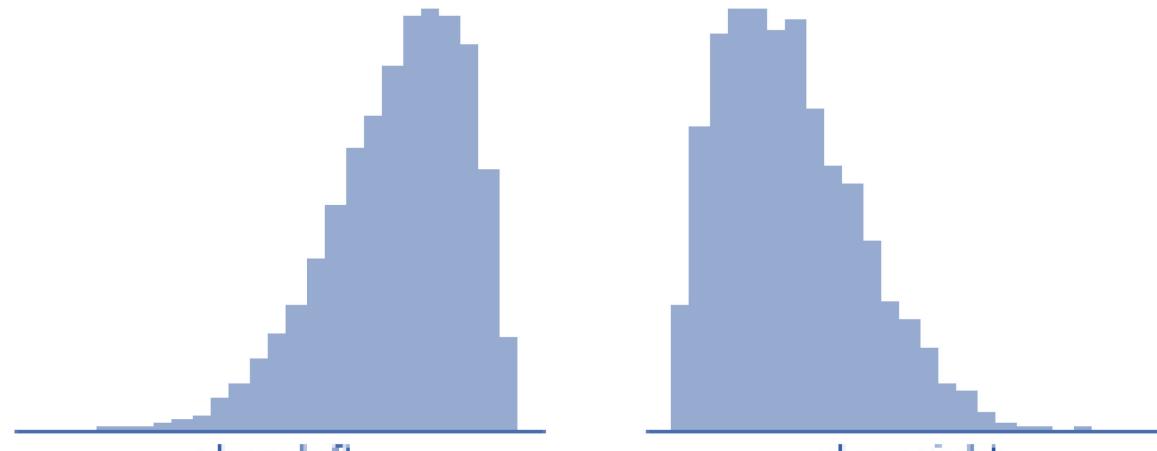




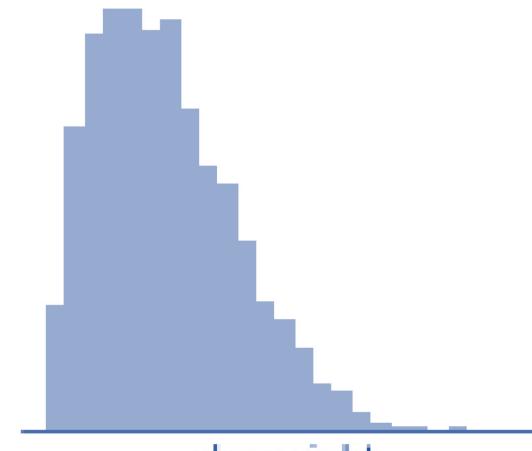
symmetric, unimodal



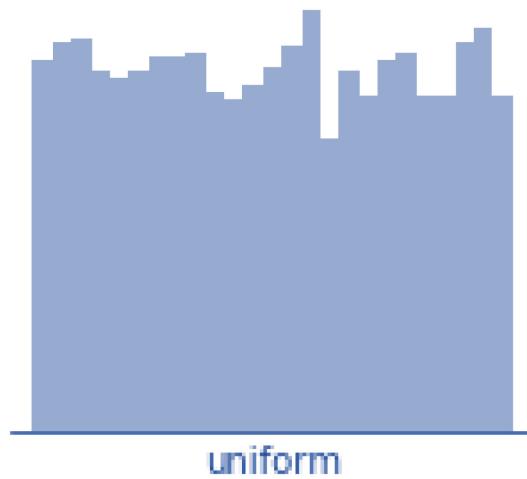
symmetric, unimodal



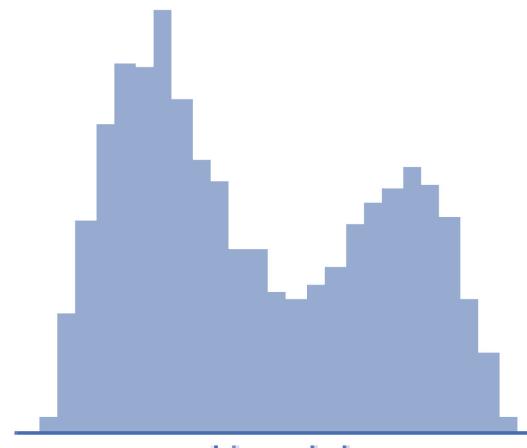
skew left



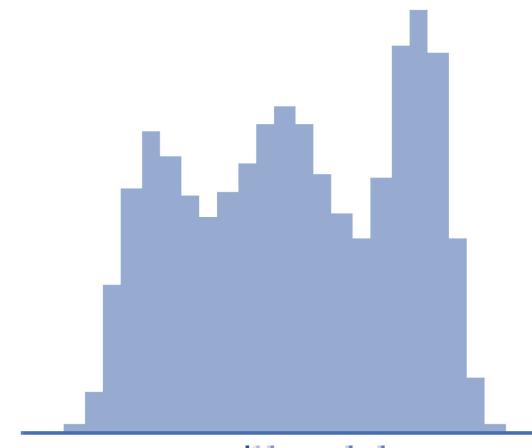
skew right



uniform

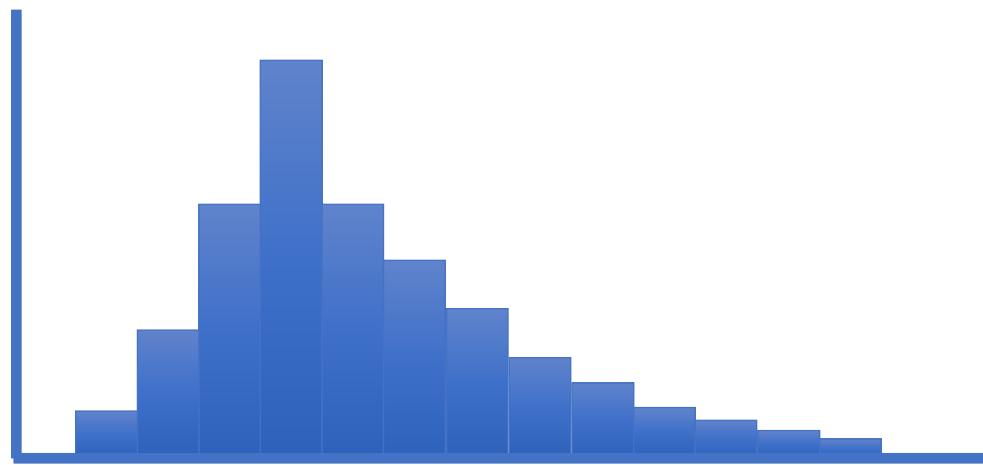


bimodal



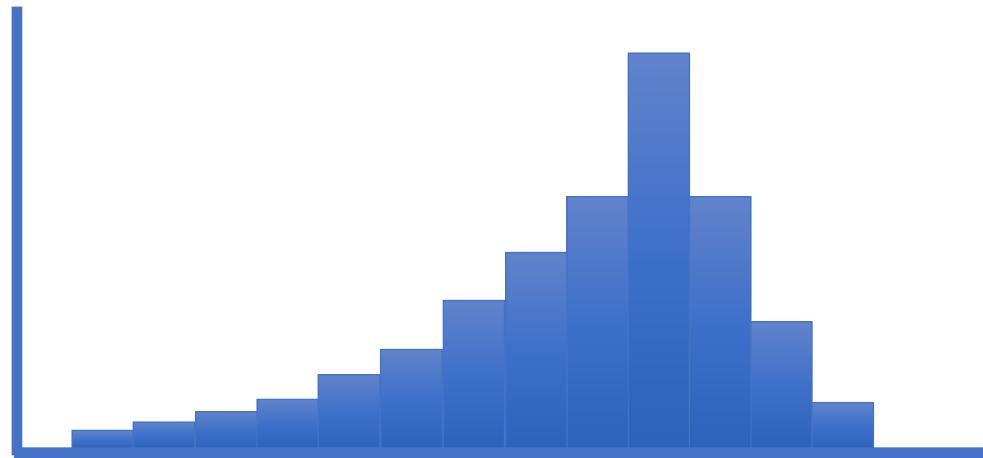
multimodal

Skew



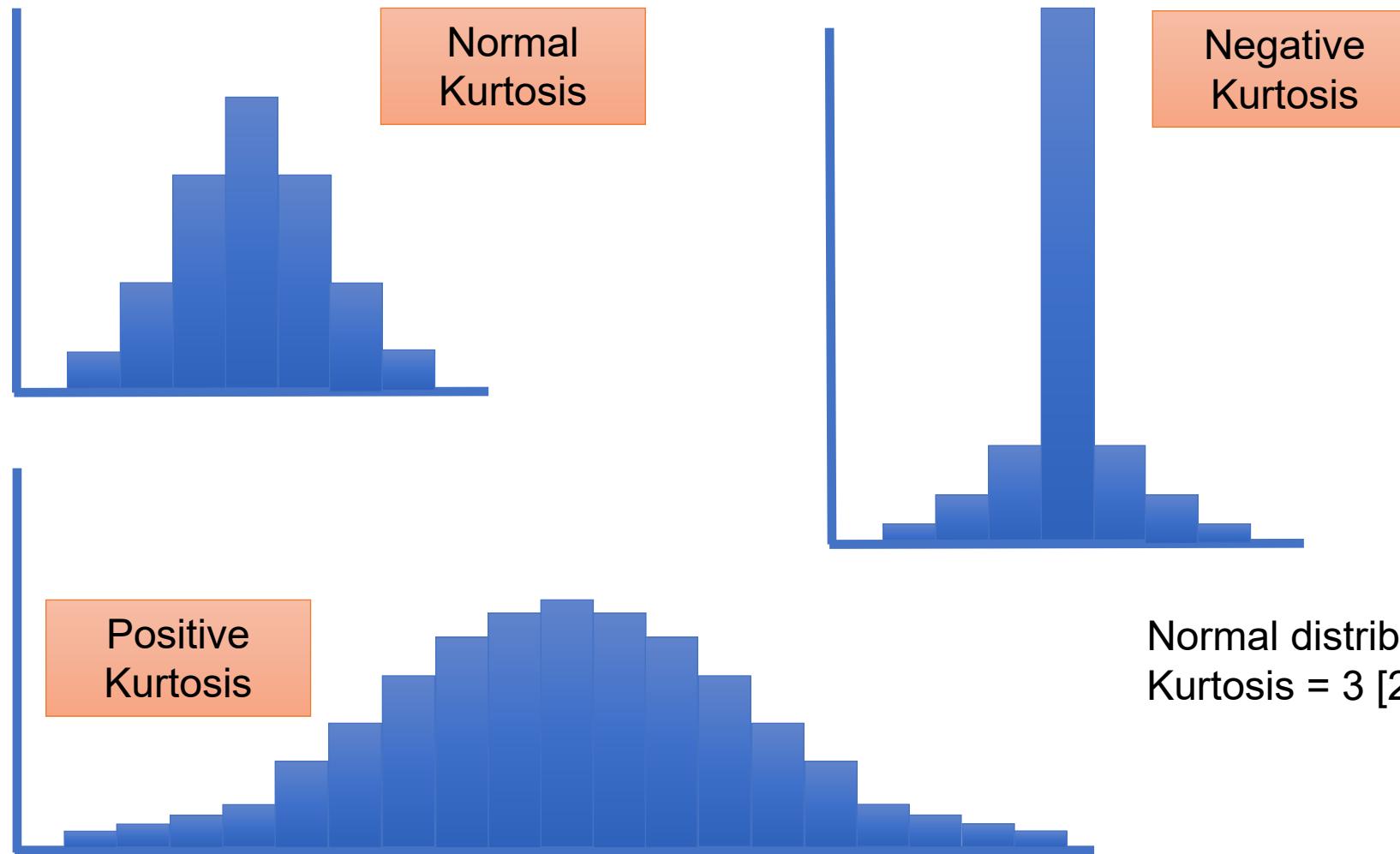
Positive Skew

Normal distribution:
Skewness = 0 [-3,3]

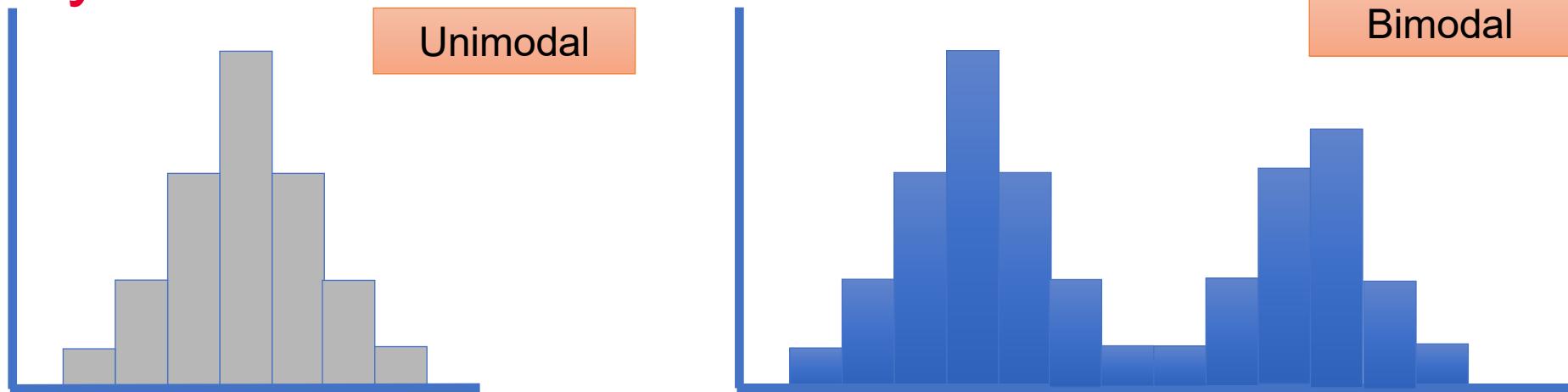


Negative Skew

Kurtosis

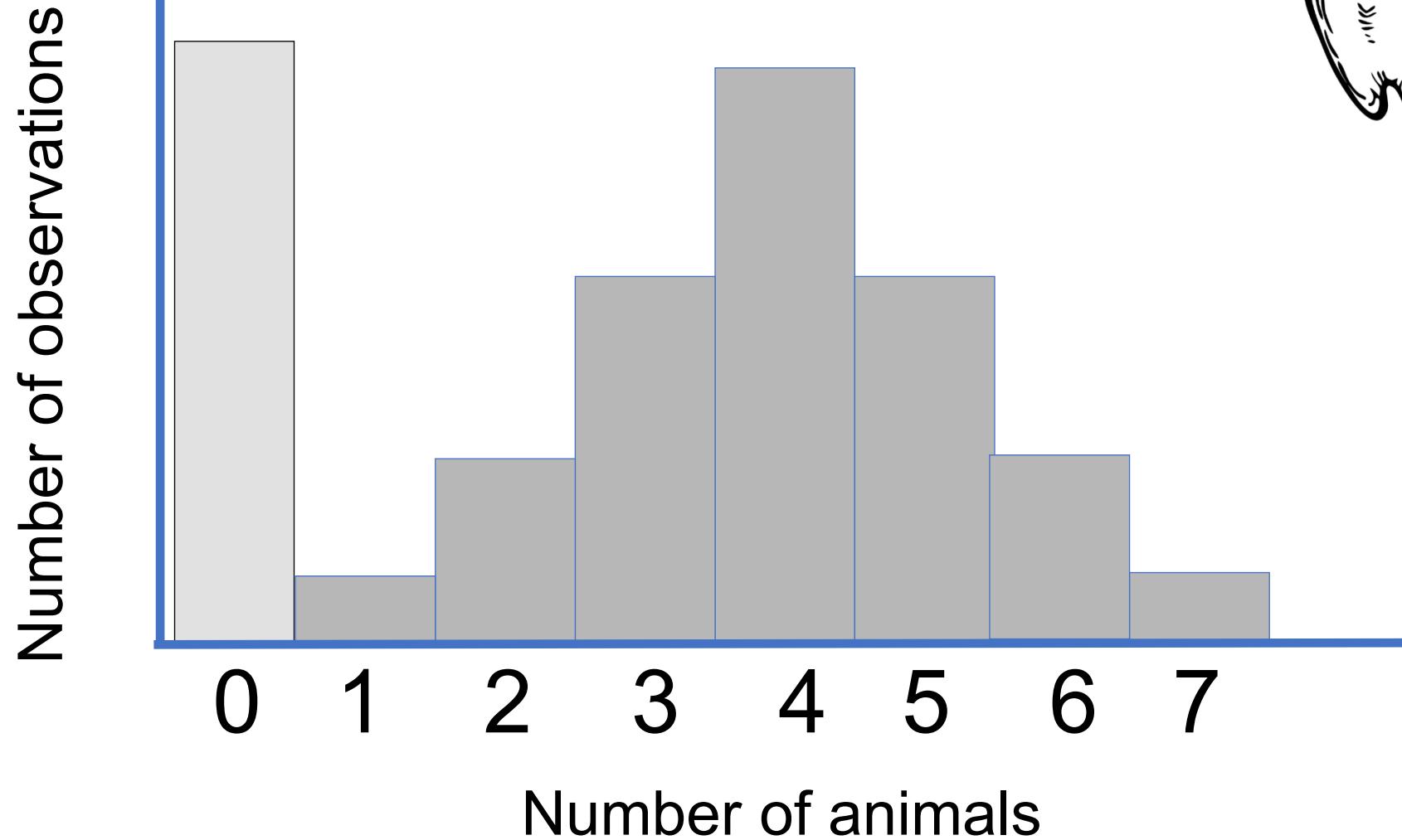


Modality



No mode
Uniform

Zero Inflation



Who you Want to Call?



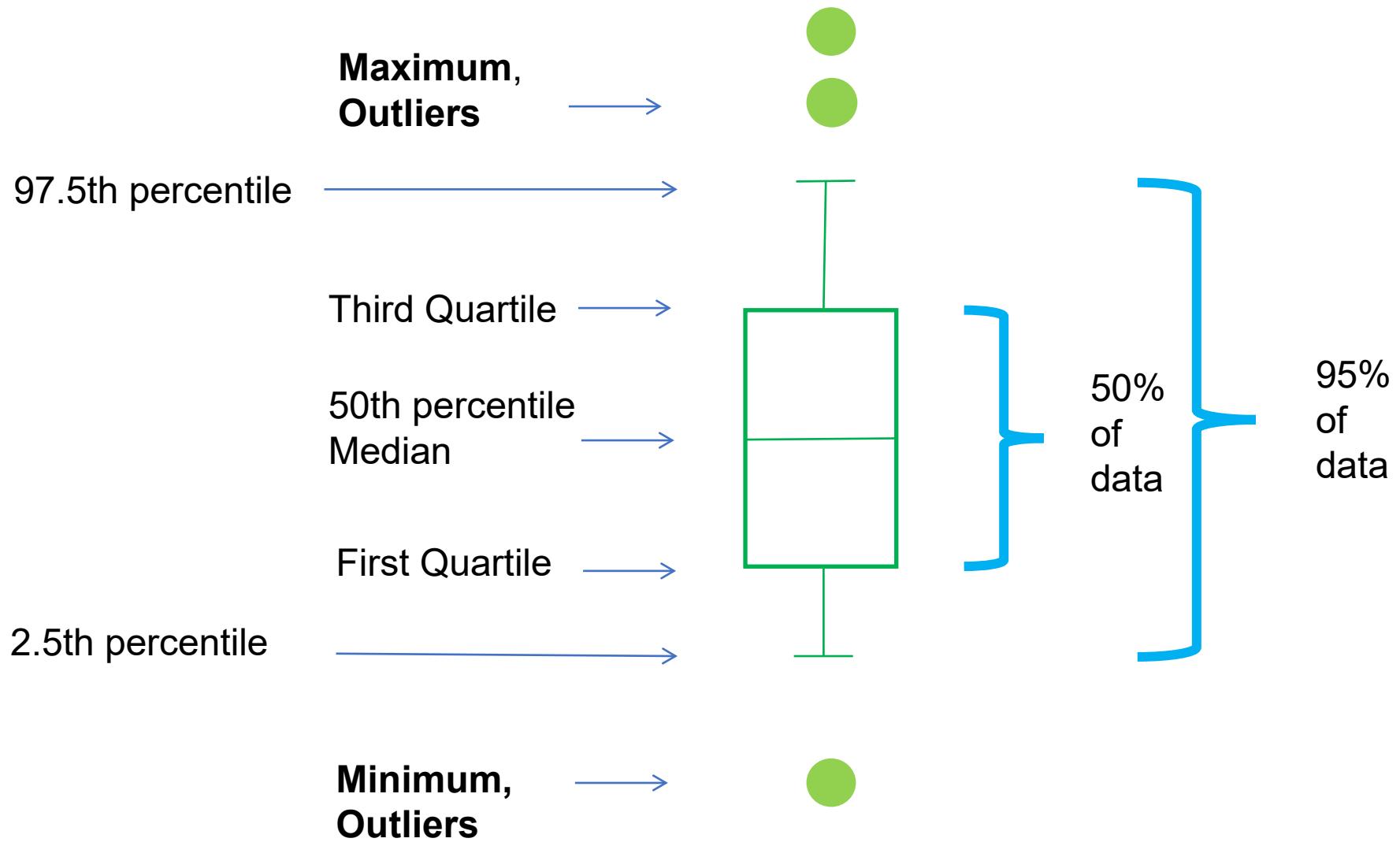
GHOSTBUSTERS
AFTERLIFE

Who you Want to Call?

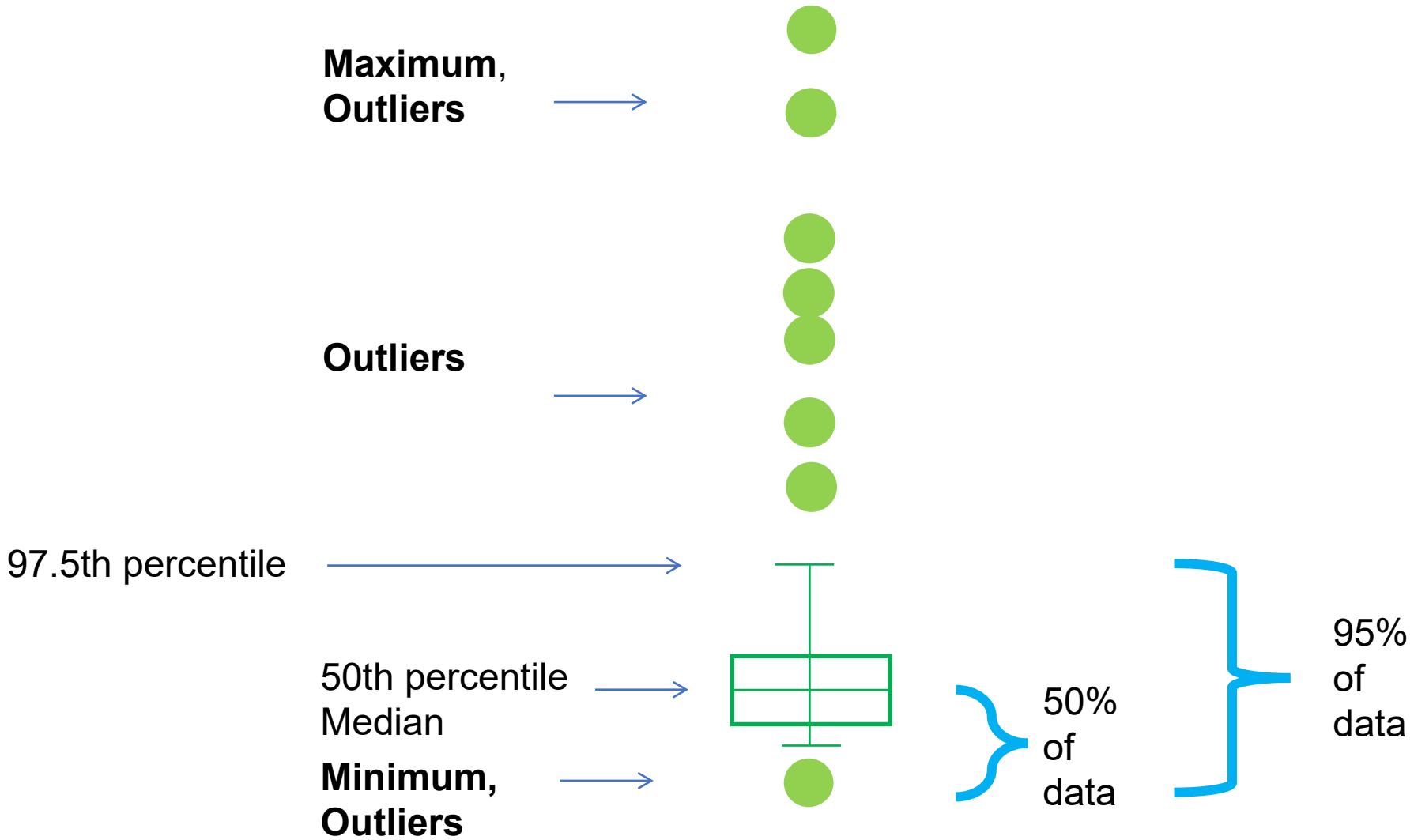


Percentiles!

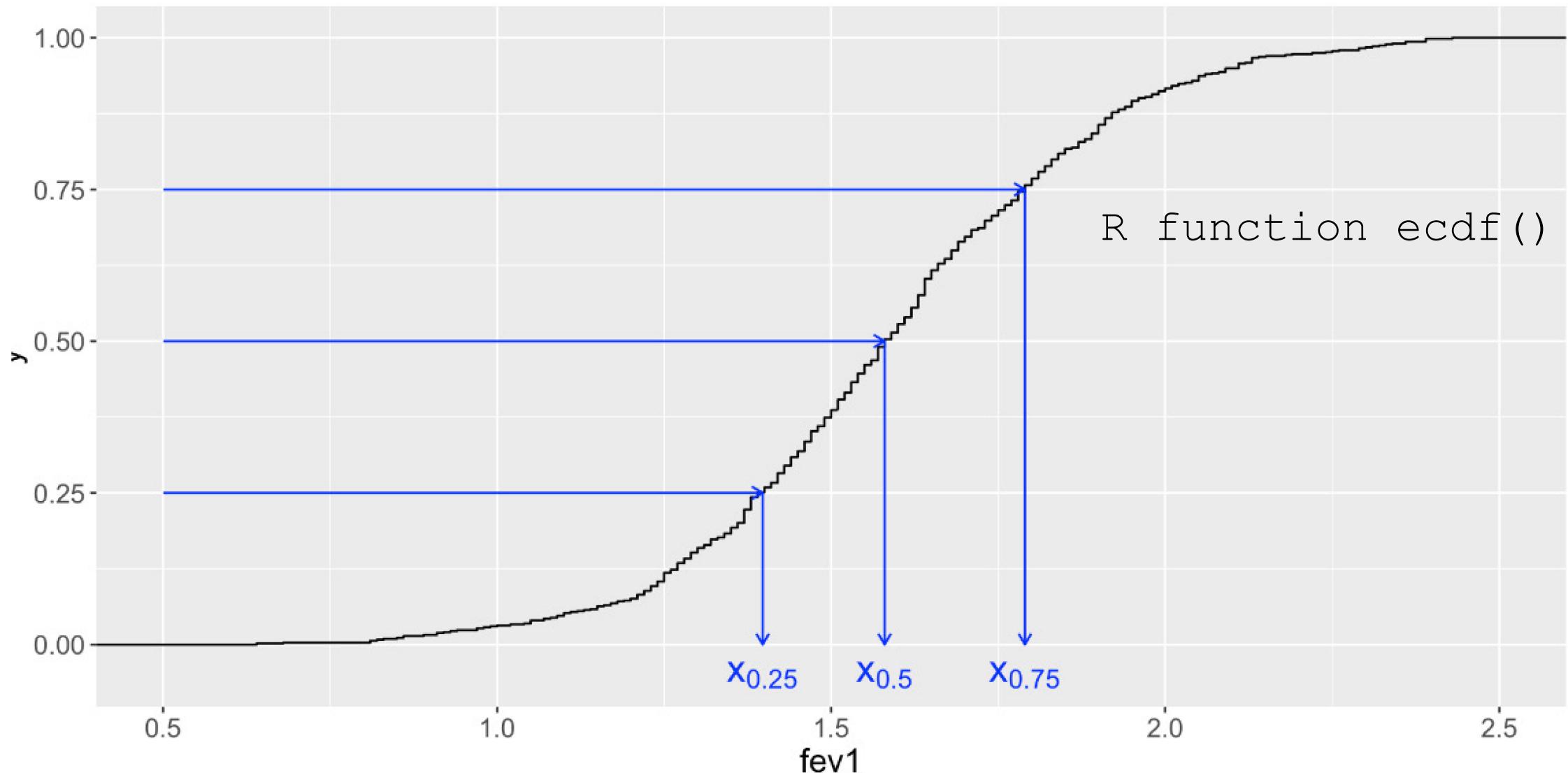
Box and Whiskers Plot (Normal)



Box and Whiskers Plot (Skewed)



Empirical Cumulative Distribution Function



Percentiles

- Sort observations from min to max
- Take 100 segments (1%-segments)
- 25th percentile is the value below which 25% of the data can be found

80th percentile is the value below which
80% of the data is found

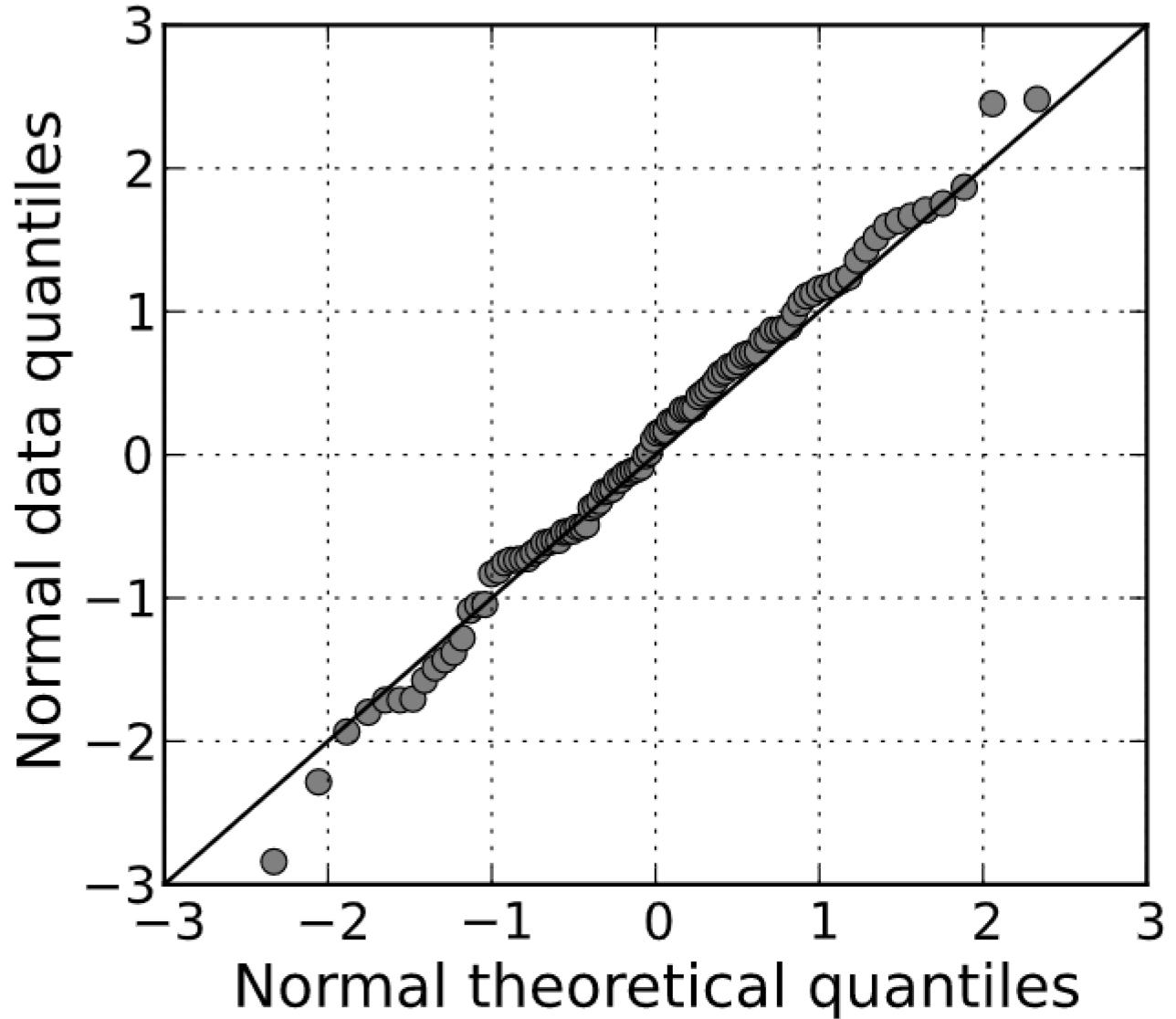


For someone as tall as the 80th percentile,
80% of people are shorter

Quantile Quantile Plot QQ-Plot

"a graphical method for comparing two probability distributions by plotting their quantiles against each other"

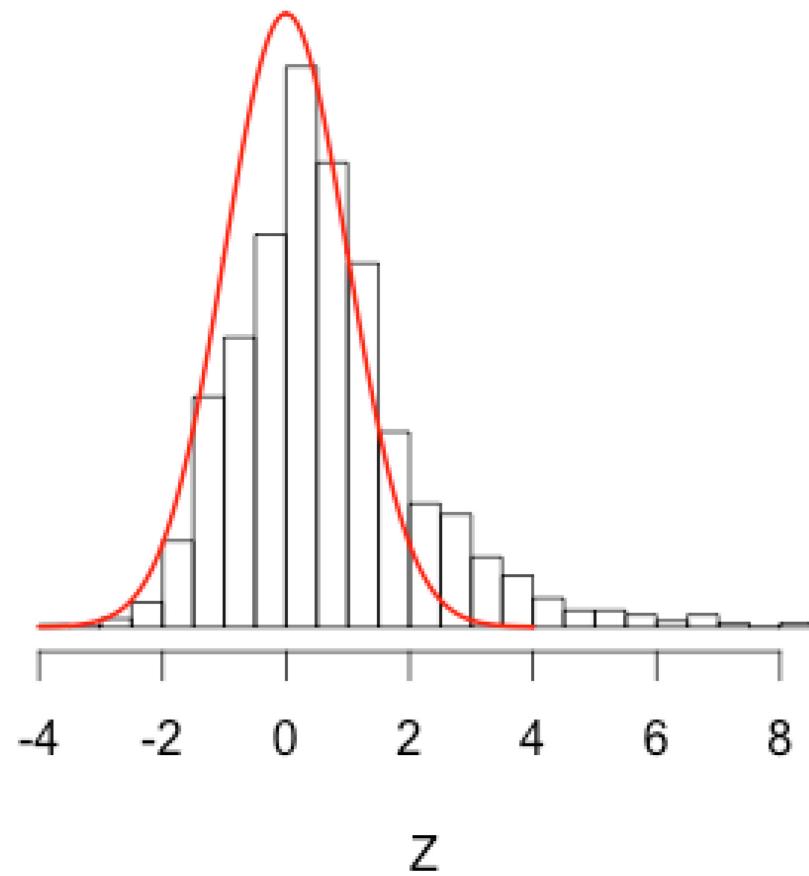
Wilk, M.B.; Gnanadesikan, R. (1968)



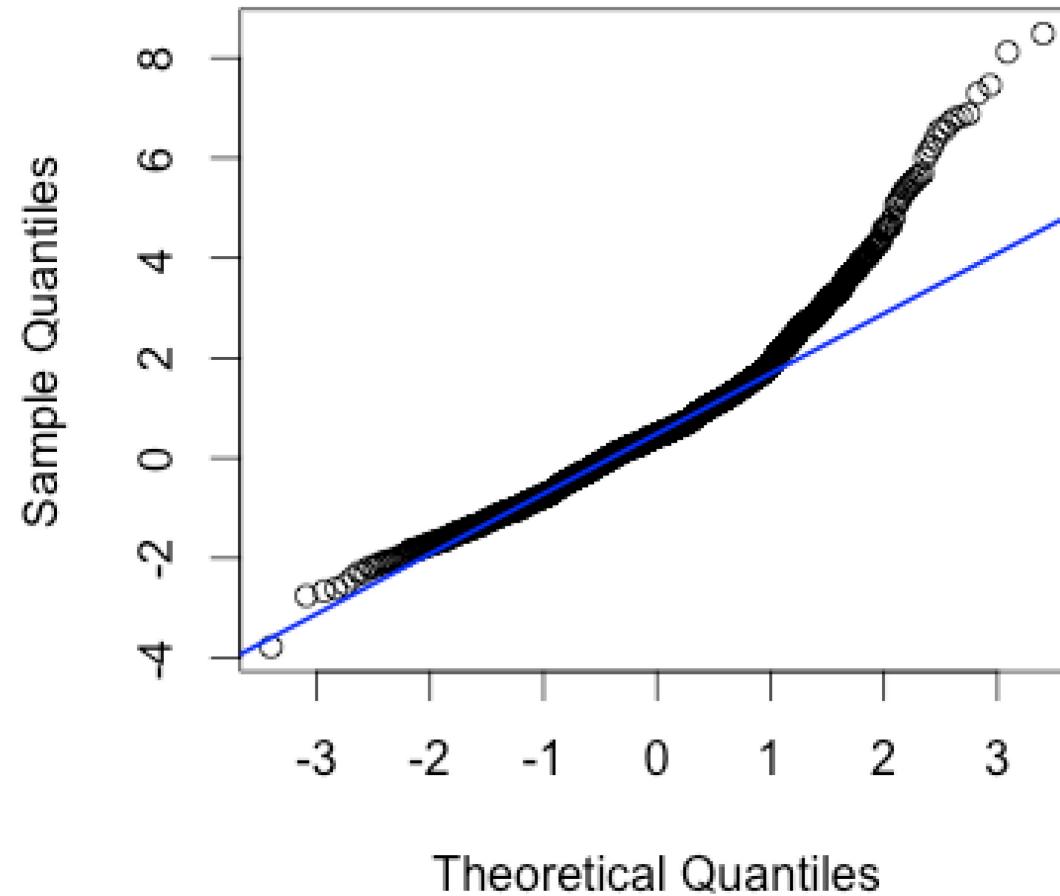
Skewed Right

<https://seankross.com/2016/02/29/A-Q-Q-Plot-Dissection-Kit.html>

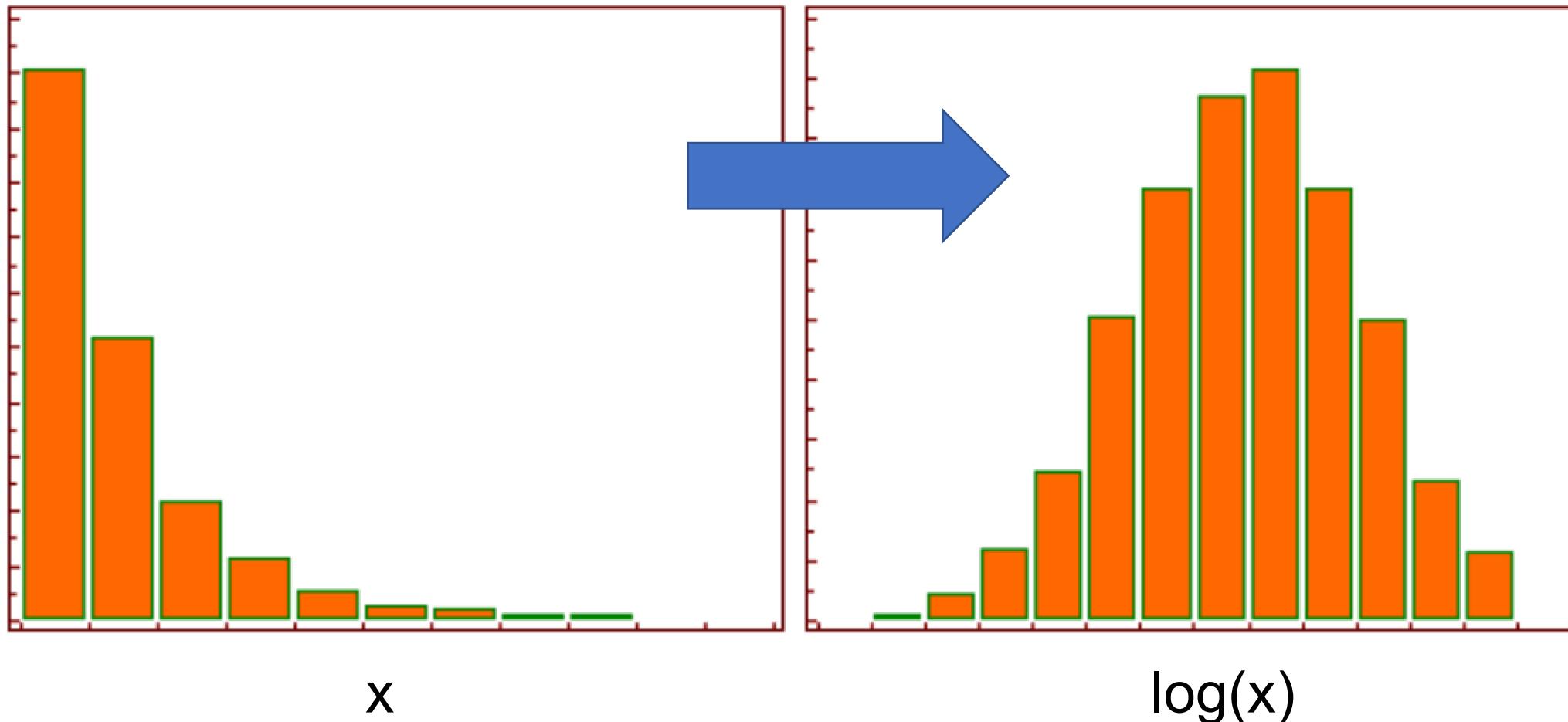
Skewed Right



Normal Q-Q Plot



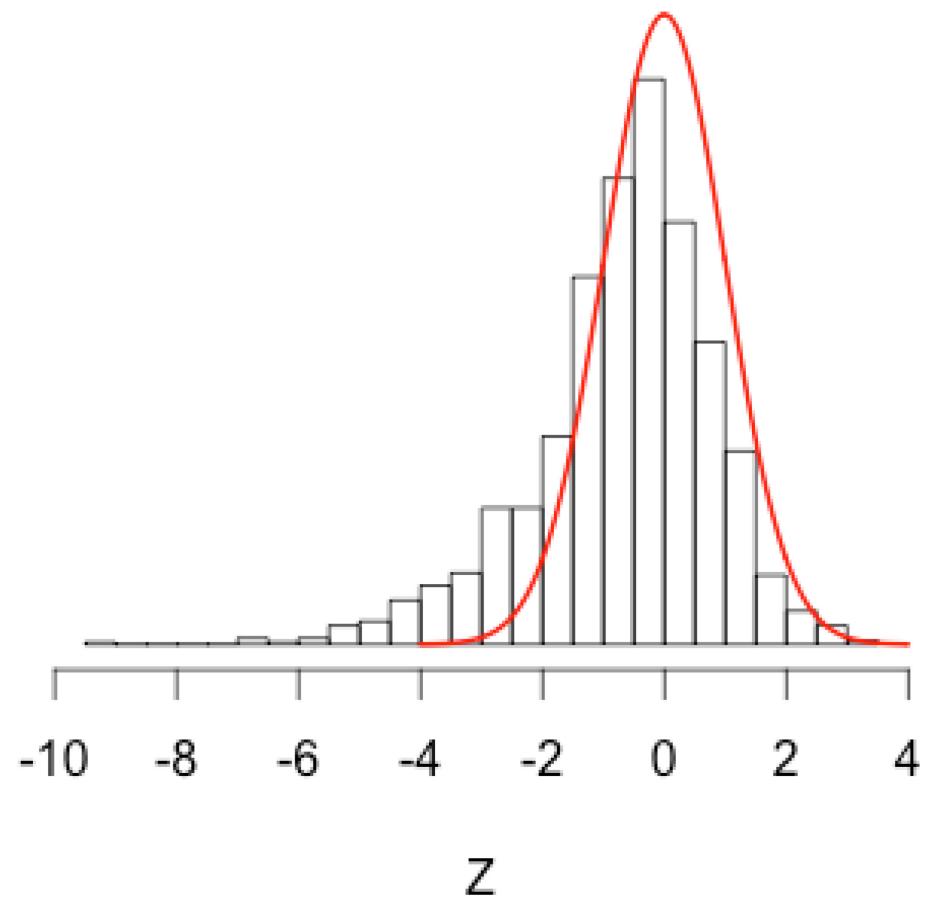
Right Skew and Log Transformation



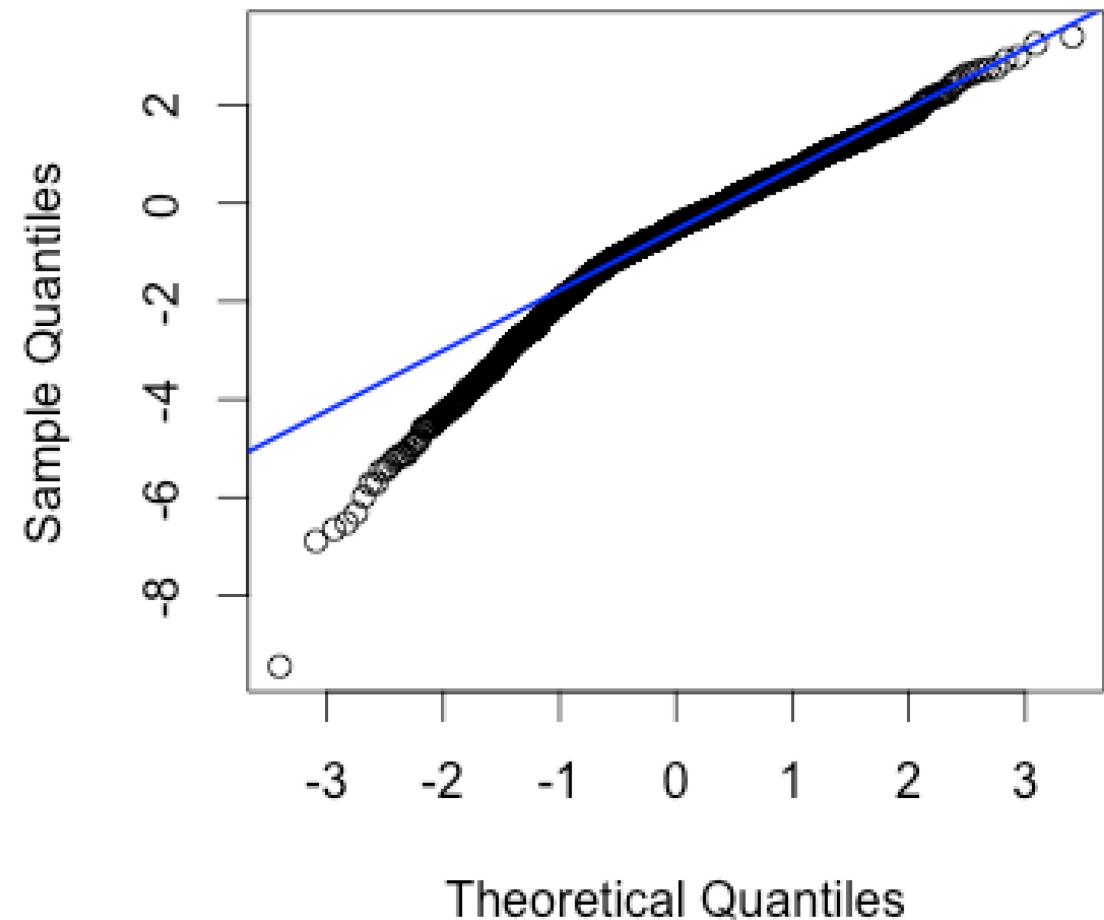
Skewed Left

<https://seankross.com/2016/02/29/A-Q-Q-Plot-Dissection-Kit.html>

Skewed Left



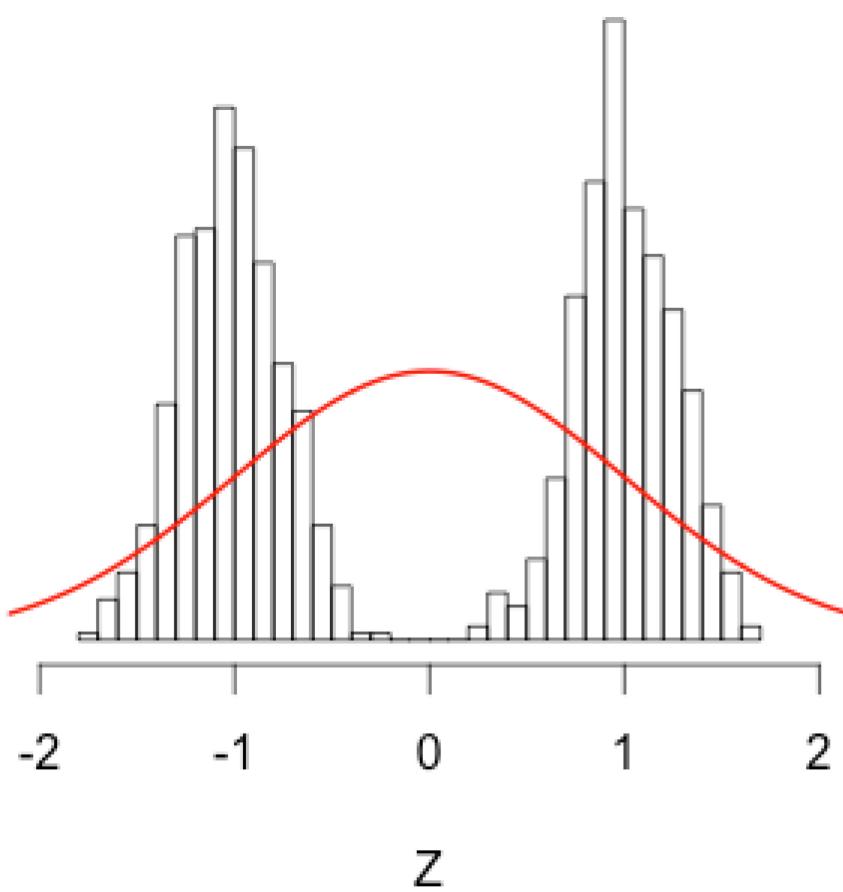
Normal Q-Q Plot



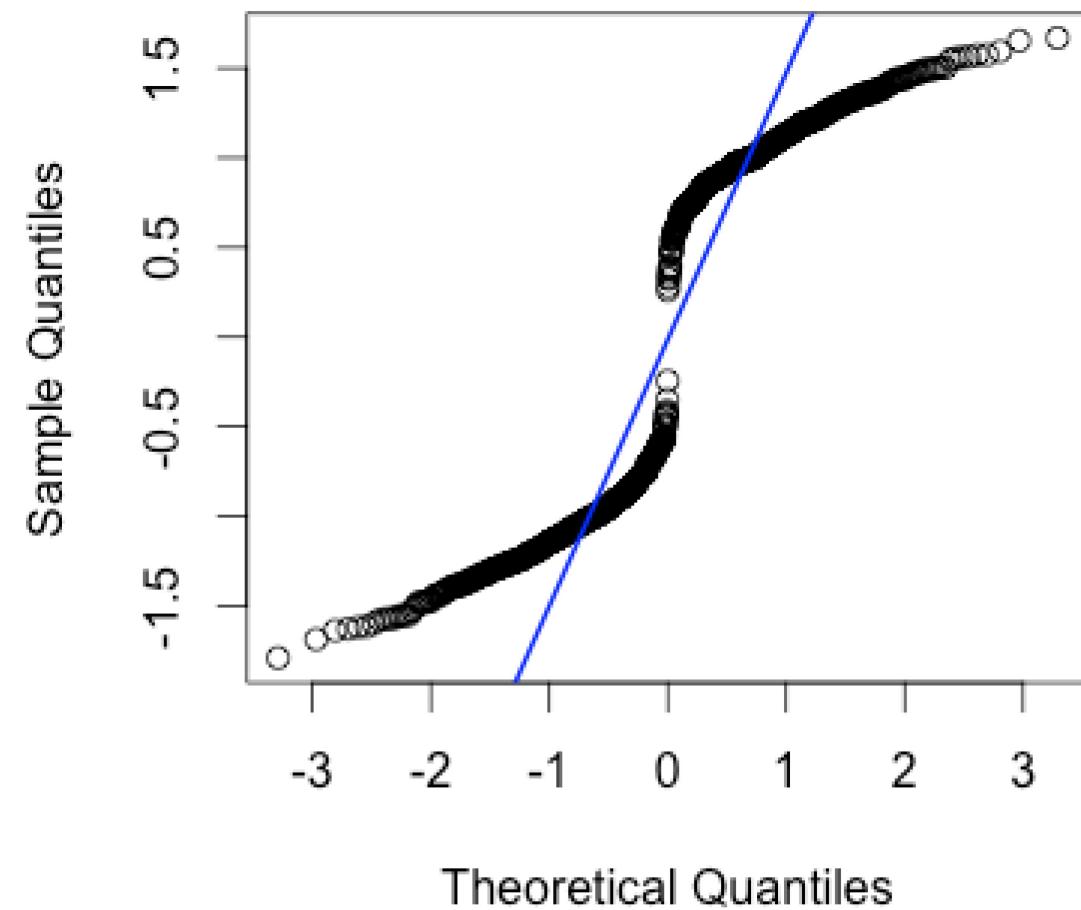
Bimodal

<https://seankross.com/2016/02/29/A-Q-Q-Plot-Dissection-Kit.html>

Bimodal

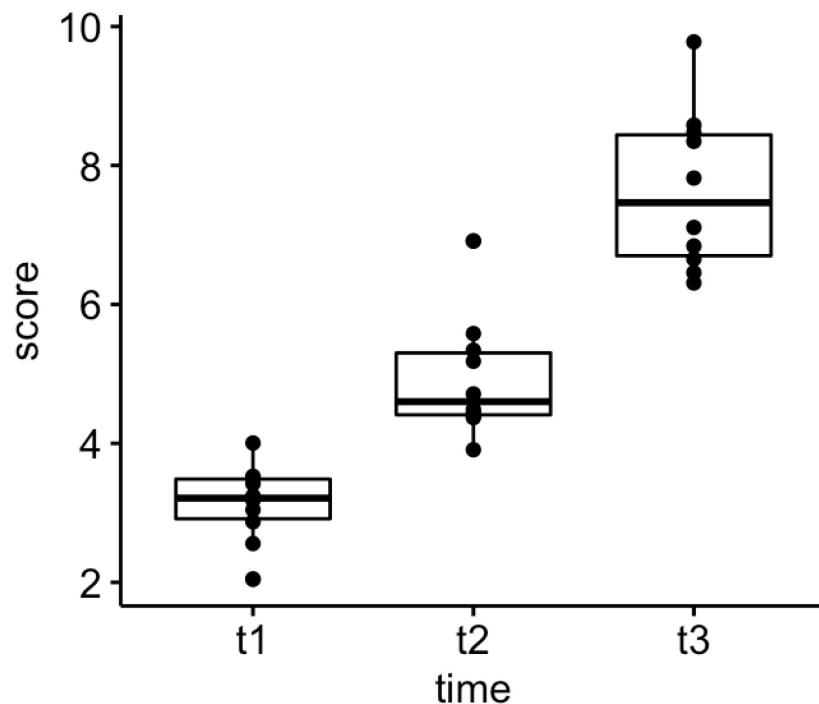


Normal Q-Q Plot

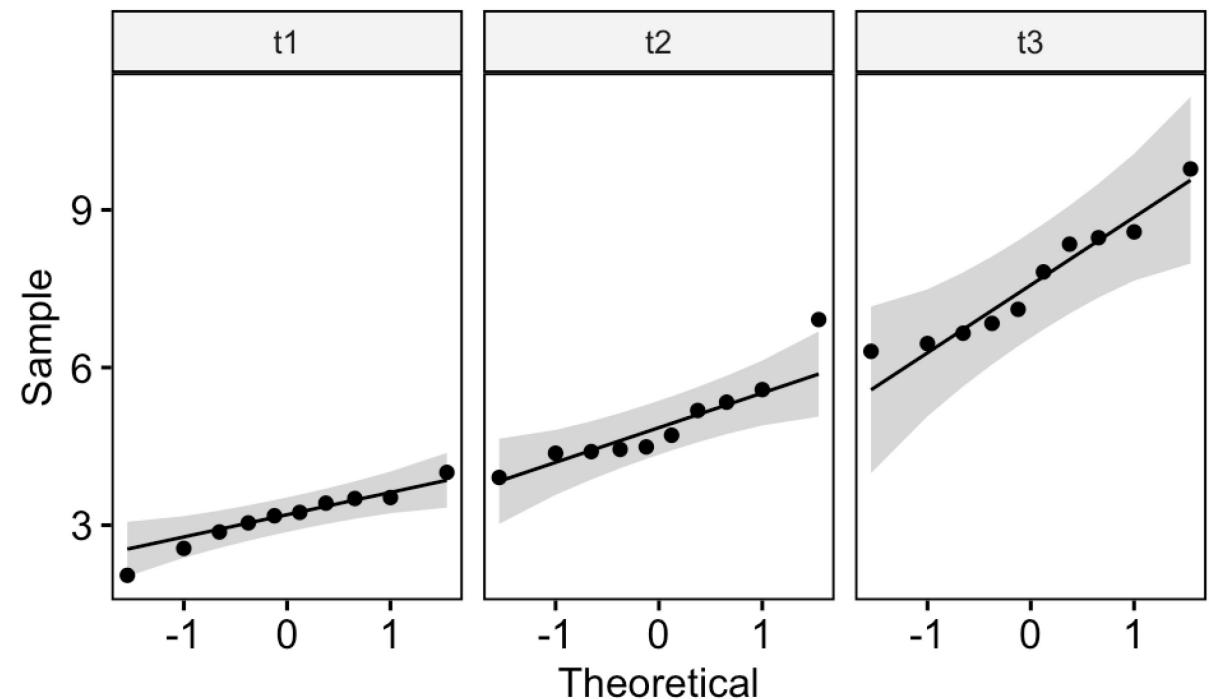


Normality Tests – Each Group Separately

When our research question is about comparing two groups
We need to test each group separately



<https://www.datanovia.com/>



Normality Tests

Shapiro-Wilk

R function shapiro.test()

Ho: data is Normally distributed

If p-value > 0.05 we can not reject Ho

Kolmogorov-Smirnov

R function ks.test()

One-sample test to compare to Normal or a two-samples test

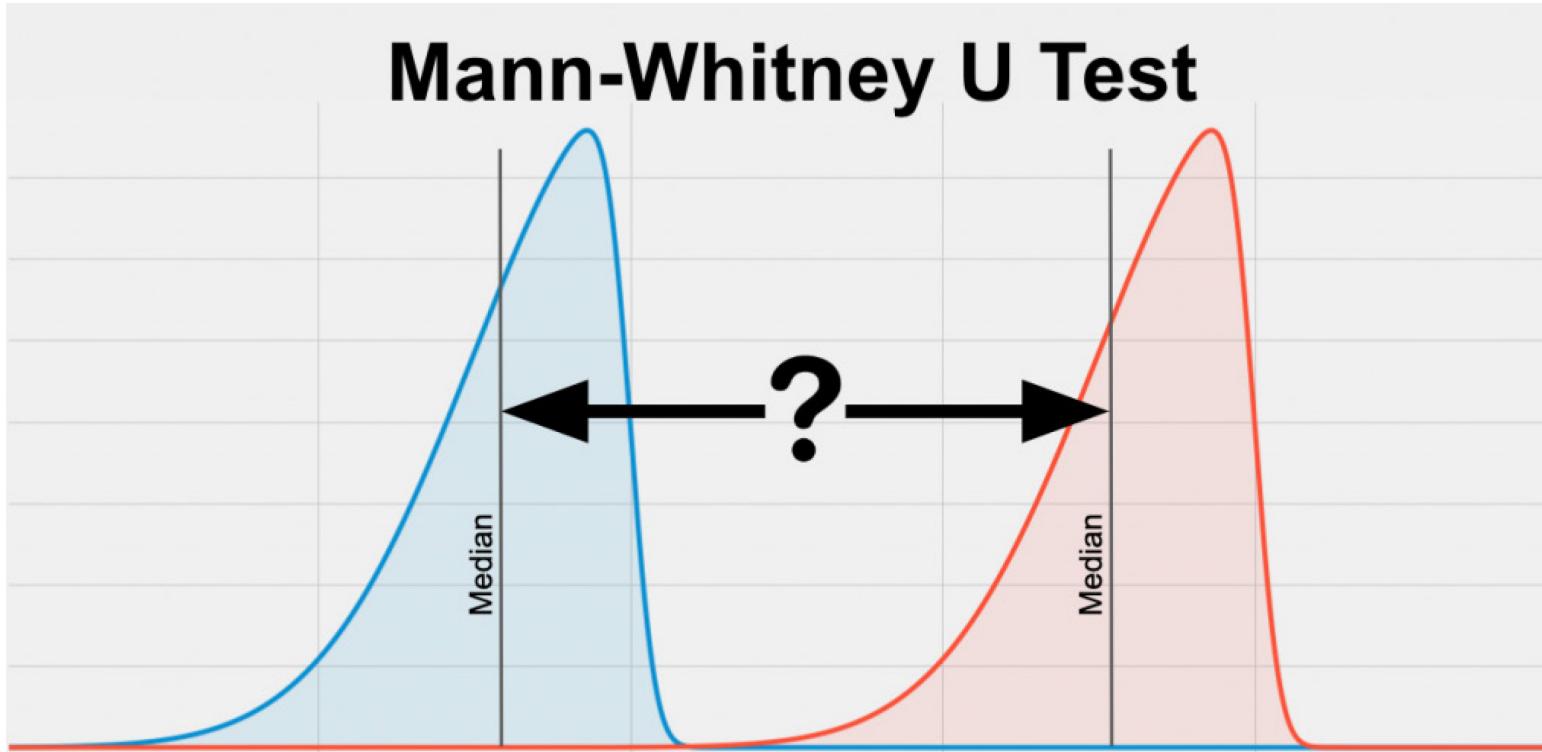
Ho: two samples were drawn from the same distribution

```
x <- rnorm(50) ; y <- runif(30)  
# Do x and y come from the same distribution?  
ks.test(x, y)
```

Comparing Independent non-Normal Distributions

Mann–Whitney U Test

<https://www.statstest.com/mann-whitney-u-test/>



Given two identically shaped and scaled distributions,
H₀: are the medians different?

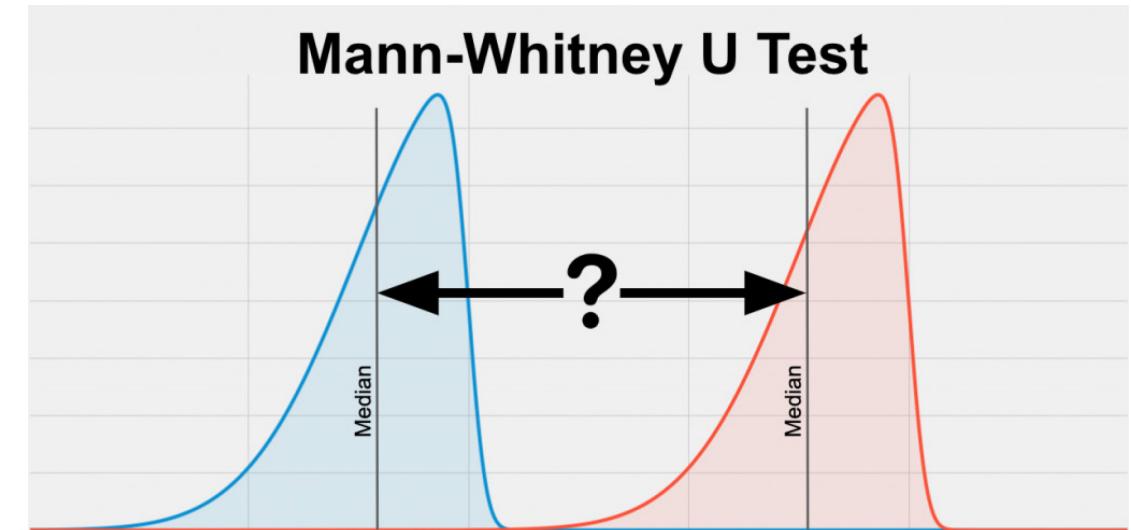
Mann–Whitney U Test

<https://www.statstest.com/mann-whitney-u-test/>

Wilcoxon 1945

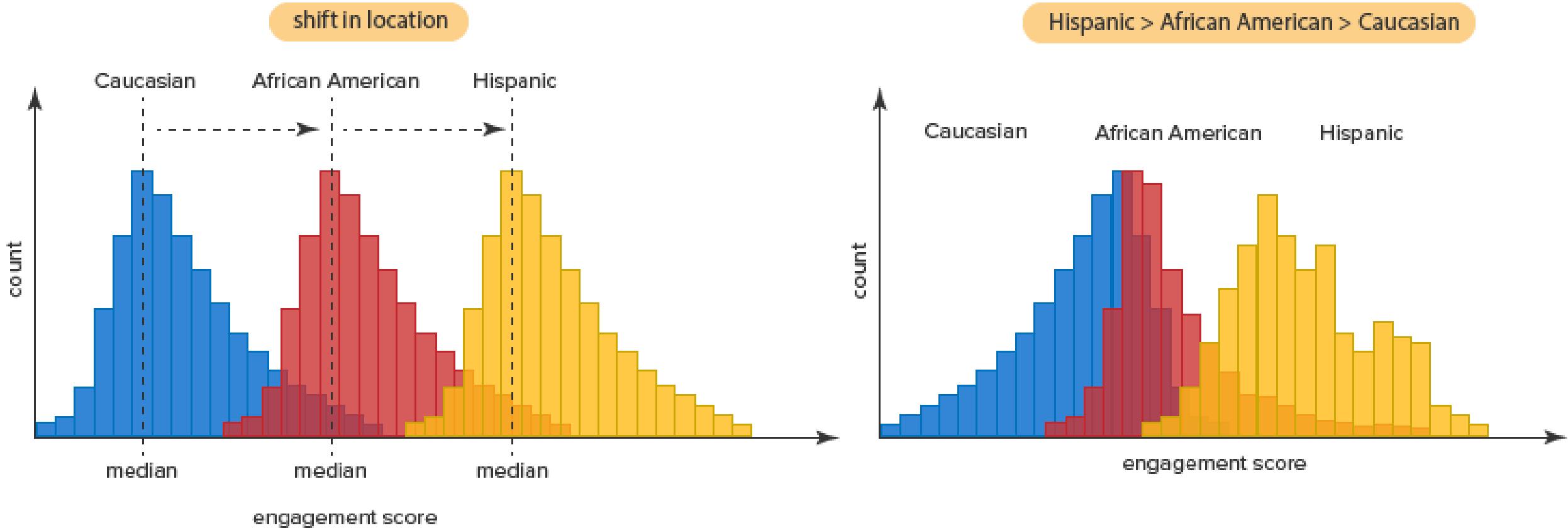
Mann & Whitney 1947

Mann–Whitney–Wilcoxon



R function `wilcox.test(y~x, paired = FALSE)`

Kruskal Wallis Test: More than Two Groups



Kruskal Wallis Test

W. H. Kruskal, W. A. Wallis 1952

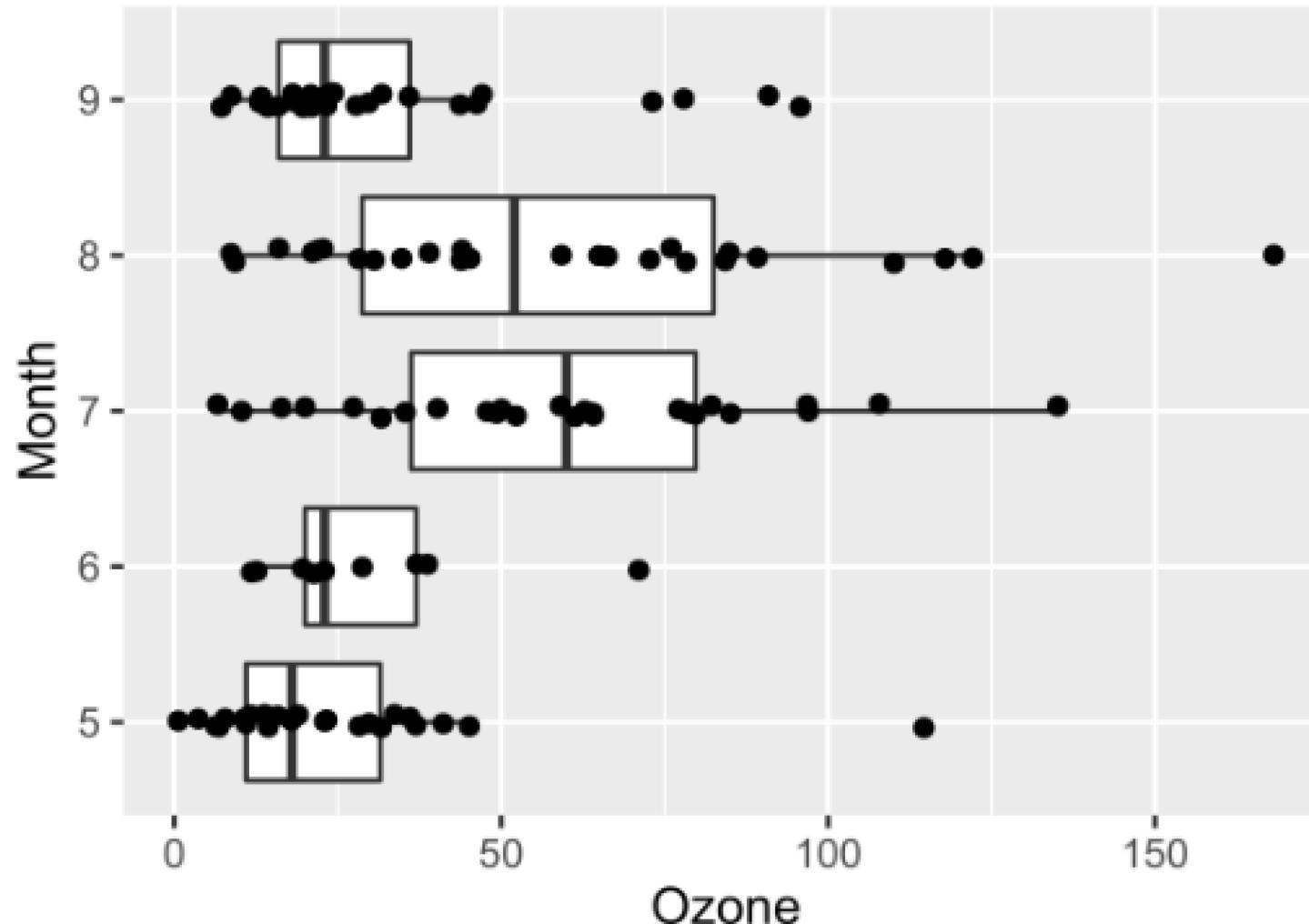
One-way ANOVA on ranks

Similar assumption as Mann-Whitney U

For identically shaped and scaled distribution for all groups,

Ho: are the medians different?

Kruskal Wallis Test



R function
kruskal.test()

Multiple Comparisons Corrections



The more comparisons we make,
the higher the chances of rejecting H_0



Multiple Comparisons Corrections



One comparison
Significance level = 0.05



Reduce the level of significance (alfa)
for each comparison
So that the probability of getting
one wrong is the same as if only one
comparison was made



Multiple Comparisons Corrections

Goal: compare each pair of medians/means (or the one of interest)
WHILE achieving an overall Type I error < 5%

R function dunn.test()

```
dunn.test(y~x, method = "bonferroni")
```

Performs a Kruskal Wallis test with Bonferroni correction

Multiple Comparisons Corrections

For Non-Normal (independent) groups comparisons

```
dunn.test(y ~ x, method = "bonferroni")
```

```
pairwise.wilcox.test(y, x, p.adjust.method = "bonferroni")
```

```
pairwise_wilcox_test(y, x, p.adjust.method = "bonferroni")
```

For Normal (independent) group comparisons

```
stats::pairwise.t.test(y ~ x, p.adjust.method = "bonferroni")
```

```
rstatix::pairwise_t_test(y ~ x, p.adjust.method = "bonferroni")
```

<https://rpkgs.datanovia.com/rstatix/>

Reasons Not to Correct p-values

- The sample size (n) calculation depends on both type I (alfa) and type II (beta) errors
- So if we decrease alfa and don't increase n , then beta error increases (to keep the equation)
- Result: We loose statistical power
- We would need to increase the sample size to keep the same power
- This is a problem when sample size is small (occurs often)
- Solution: focus on fewer comparisons

Summary

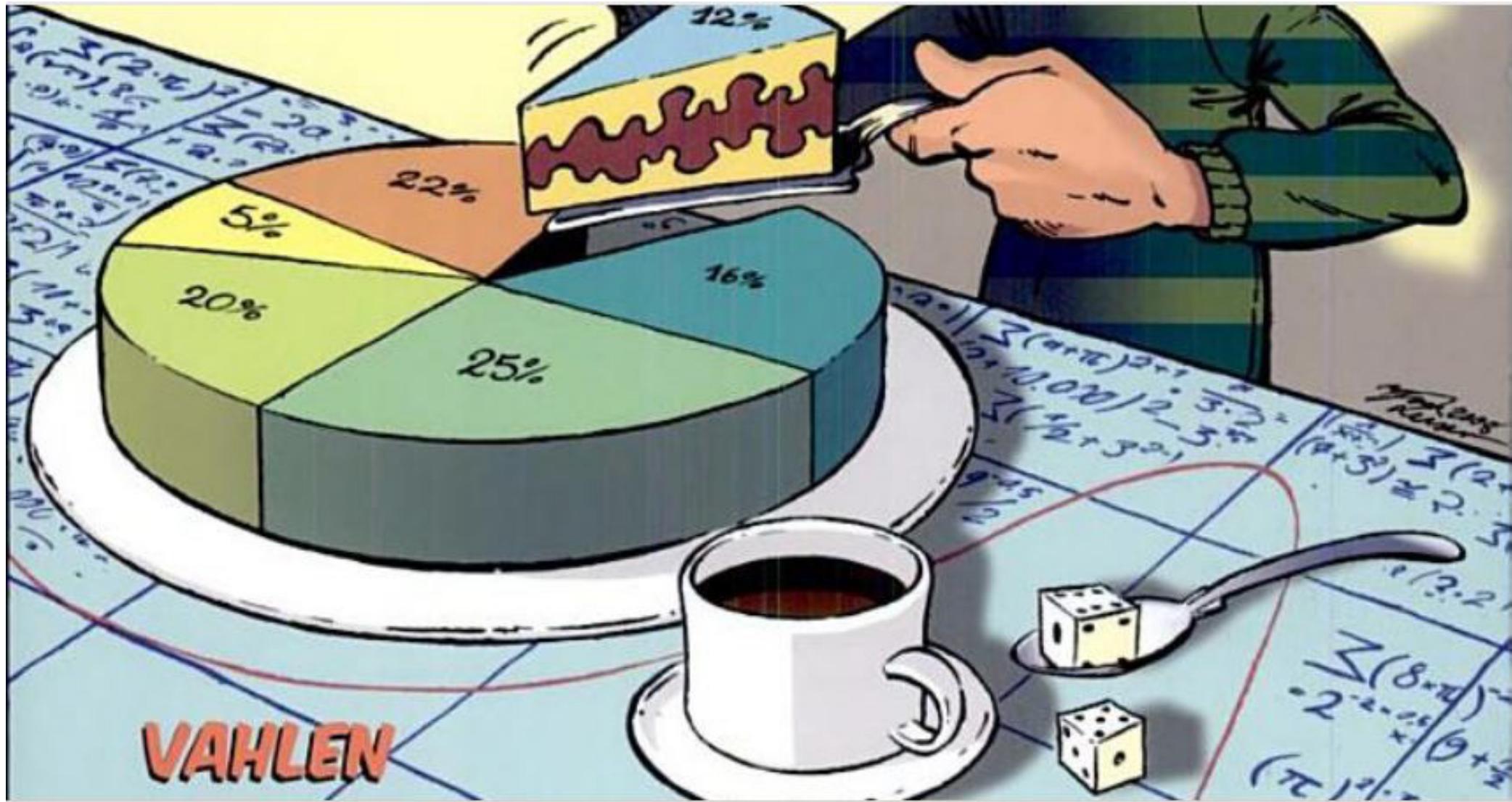
- Skewness, kurtosis, modality, zero-inflation
- Percentiles (Ghostbusters! ☺) and ecdf
- QQplots and Normality tests (Shapiro-Wilk, Kolmogorov-Smirnov)
- Mann Whitney U test to compare 2 groups (`wilcox.test()`)
- Tests that include multiple comparison corrections (pairwise comparisons with Bonferroni correction)

Table to Choose the Right Test

1. Measurement scale and distribution of dependent (outcome) variable (-> COLUMNS)
2. Study objective related to type and scale of independent (often group) variable (-> ROWS)
3. Independent or dependent observations between subjects?

	Type of dependent variable ("outcome" or y-variable)				
Study objective	Interval data (normally distributed measurements)	Ranks, Scores, or non-normally distributed measurements	Binary outcome (two levels)	Survival Time	
Describe a single study group (summaries, frequencies)	Mean, median, mode, SD, SEM, percentiles	Median, range, interquartile range, percentiles	Freq., Proportion (Prevalence) + 95% CI	Kaplan Meier survival curve	
Compare a single study group to a fixed (population) value	Single sample t test	Wilcoxon Signed-Rank test	Single sample proportion test		
Compare two independent study groups	Independent sample t test	Two-sample Wilcoxon (Rank Sum) test	Two-sample proportions test (Chi-square, FET)	Log-rank test or Mantel-Haenszel	
Compare two paired (dependent) sample study groups	Paired sample t test	Paired sample Wilcoxon (Signed-Rank) test	McNemar's test, Kappa statistic	Conditional proportional hazards regression	
Compare three or more unmatched (independ.) study groups	One-way ANOVA	Kruskal-Wallis Test	Proportions test (Chi-square), Logistic Regression	Cox proportional hazard regression	
Compare three or more matched (dependent) groups	Multiway ANOVA	Friedman test	Cochrane Q	Conditional proportional hazards regression	
Quantify correlation between two variables	Pearson correlation	Spearman Rank correlation	Contingency coefficients		
Predict outcome value from another interval or categorical variable	Generalized linear Models: Simple Linear regression or ANOVA	Nonparametric regression	Cross tabulation (Odds ratio), Simple logistic regression	Cox proportional hazard regression	
Predict outcome value from several measured (interval), categorical or binomial variables	Generalized linear Models: Multivariable Linear regression or ANOVA	Generalized linear models accommodating nonparametric components	Stratified cross tables, multiple logistic regression	Cox proportional hazard regression	

Questions



Thanks for your attention

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