

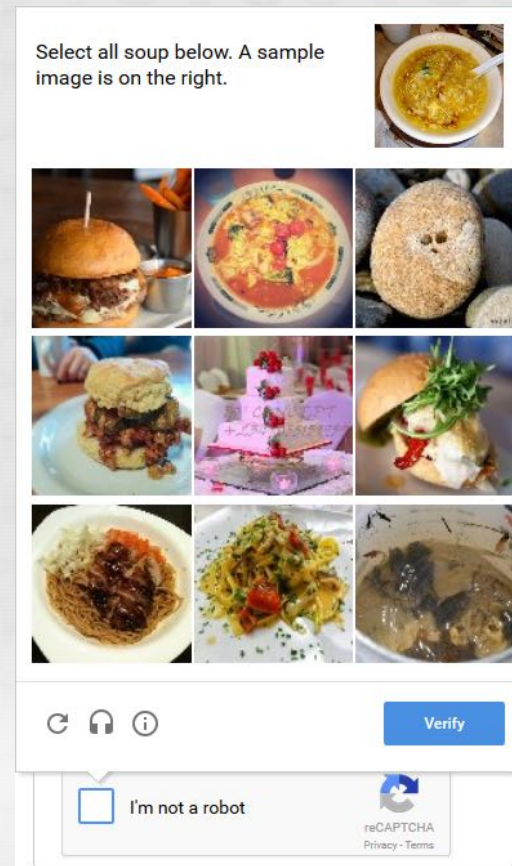
Coding and Reliability

DSA8022 Frontiers in Analytics

Dr Gary McKeown

School of Psychology, David Keir Building

0G3.536



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OVERVIEW

- Supervised Learning
 - Creating a Ground Truth
 - Training Sets and Test Sets
- Observational Methods
- Inter-rater reliability
 - Categorical Data - Cohen's Kappa
 - Continuous Data - icc
 - Both - Krippendorff's alpha

SUPERVISED LEARNING

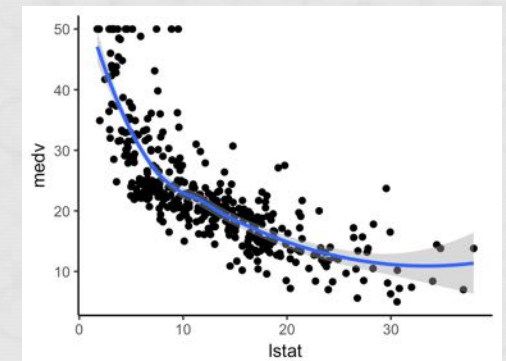
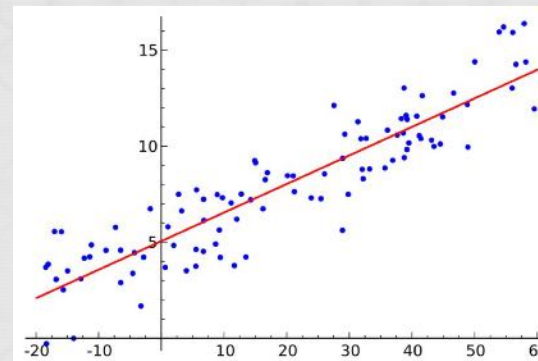
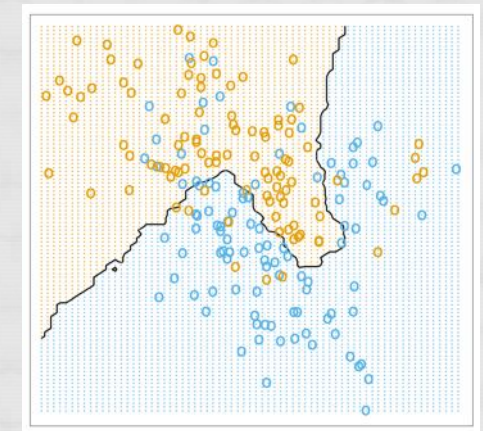
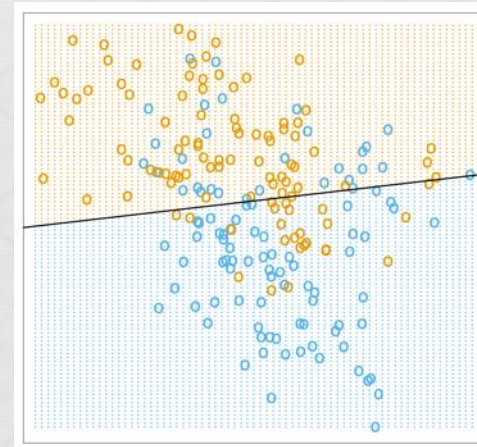
Ground Truth

UNSUPERVISED LEARNING

- Find the structure in the data without any explicit predefinition of the structure
 - Categorical Data
 - Clustering - k means
 - Topic modelling - Latent Dirichlet Allocation (LDA)
 - Continuous Data
 - Dimensionality Reduction
 - e.g Bag of Words Sentiment Analysis
 - Latent Semantic Analysis

SUPERVISED LEARNING

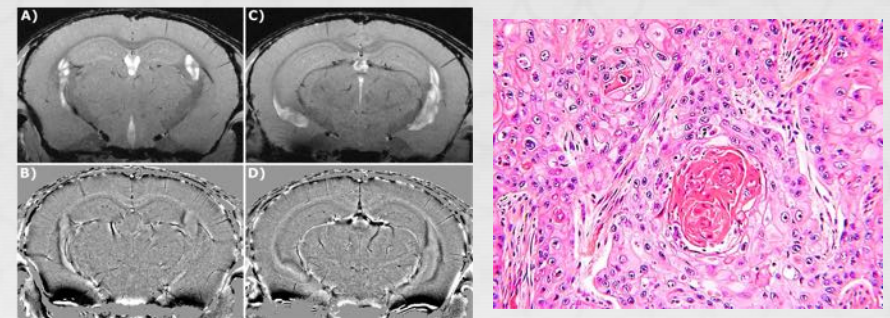
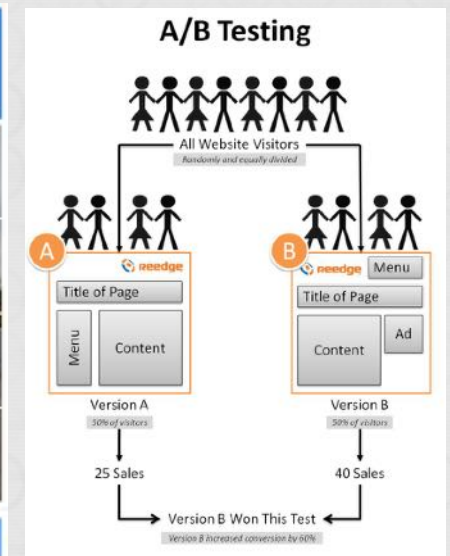
- Give a ground truth that defines the data and learn the relationship
- Training and test datasets
 - Categorical Data
 - Classification
 - Continuous Data
 - Regression



OBSERVATIONAL METHODS

MEASUREMENT

- We can measure things from many different sources
- Surveys
- Self-report
- A/B Testing
- Getting lots of people to rate things and tell us what they are
- Strategies
 - Lots of naive raters
 - Smaller numbers of expert raters



MEASUREMENT

- Lab settings
 - Specific highly controlled tasks
 - Lower ecological validity
 - Overfitted
 - Does not generalise
- In the wild
 - Natural behaviour
 - Uncontrolled
 - High ecological validity
 - More noisy



INTERRATER RELIABILITY

RELIABILITY

$$\text{Observed Score} = \text{True Score} + \text{Measurement Error}$$

- Observed Score is typically the Score on a Psychometric test
- Reliability tries to estimate the proportion of variance that is captured by any measurement tool
- Typically scores are between 0 and 1.
- A reliability of 0.8 means 80% is True Score 20% is Error variance
- High reliability means we have a better measure of what it is we are trying to measure - it is closer to the true score
- Assuming error is independent - random not systematic



MEASURING RELIABILITY

Nominal Binary

Subject	Rater 1	Rater 2
1	Yes	No
2	Yes	Yes
3	No	No
4	No	Yes
5	Yes	Yes
6	No	No
7	No	Yes

Nominal Multiple Categories

Subject	Rater 1	Rater 2
1	Cat3	Cat3
2	Cat2	Cat4
3	Cat2	Cat2
4	Cat1	Cat1
5	Cat4	Cat2
6	Cat3	Cat2
7	Cat4	Cat4

MEASURING RELIABILITY

Ordinal

Subject	Rater 1	Rater 2
1	Cat3	Cat3
2	Cat2	Cat4
3	Cat2	Cat2
4	Cat1	Cat1
5	Cat4	Cat2
6	Cat3	Cat2
7	Cat4	Cat4

Continuous

Subject	Rater 1	Rater 2
1	66.43	65.33
2	45.98	54.88
3	67.98	66.98
4	76.99	64.87
5	34.55	56.72
6	78.99	57.91
7	76.88	89.43



MEASURING RELIABILITY



- How to measure the relationship between raters?
 - Pearson's correlation?
 - Percentage agreement?
- Issues of chance agreement
- Issues of systematic error

Subject	Rater 1	Rater 2
1	66.43	65.33
2	45.98	54.88
3	67.98	66.98
4	76.99	64.87
5	34.55	56.72
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7	76.88	89.43

INTER RATER RELIABILITY



- Quantifies the degree of agreement between two or more raters who make independent ratings about the features of a set of subjects.
- Subjects can be people, behaviours, images, social media posts, adverts, movie reviews
- Many options that depend on:
 - The nature of the data
 - The number of raters
 - The goal of the rating
 - The amount of ratings that have been conducted - full coverage or partial coverage

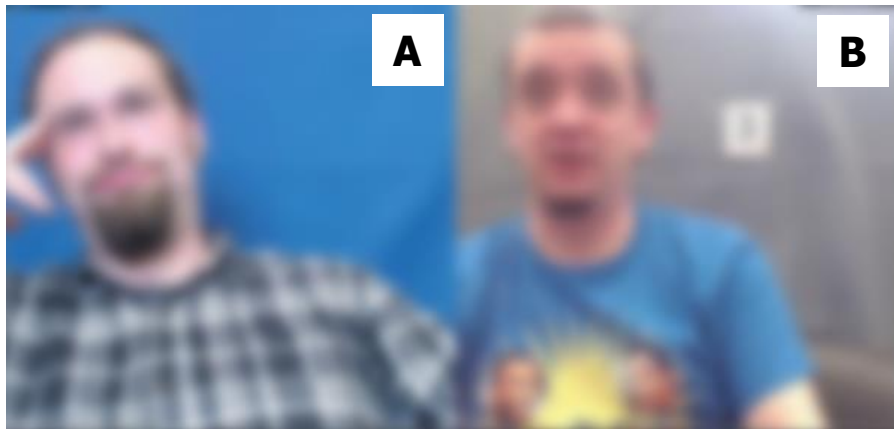
INTER RATER RELIABILITY

- Many options:
 - Percentage agreement
 - Cohen's Kappa
 - Nominal data - 2 raters
 - Cohen's weighted Kappa
 - Ordinal data - 2 raters
 - Fleiss' Kappa
 - Nominal data - 2 or more raters
 - Intra Class Correlation
 - Continuous/interval data - multiple raters
 - Krippendorff's alpha
 - Nominal, Ordinal, Interval data, many raters

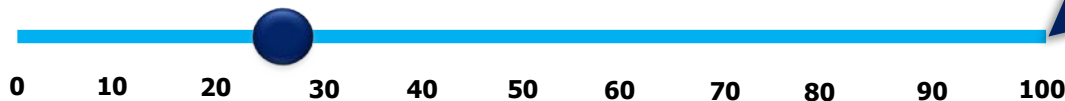


An example of behaviour analysis - non-verbal expressivity slider measurement on a continuous scale

Example: online non-verbal rating task



How non-verbally expressive is Person A?



Moveable, marked slider.

0 to 100.

2 decimal points.

Micro-level nonverbal annotation

head

- Nodding
- Cock to side
- Shaking
- Other positional

smiling

- Duration
- Intensity 1 - 3

scowling

- Duration
- Intensity 1 - 3

Vocalisation/verbalisation behaviours

- All audible sounds
- Speaking duration
- Interruptions
- Back-channelling
- Laughter duration
- Laughter intensity

eyebrows

- Raise
- Furrow
- Other

eyes

- Gaze direction
- Mutual gaze
- Exaggerated opening
- Closing

- hand.**
- Self-touch
 - Self-adaptors
 - Object-adaptors
 - Gestures



THIN SLICING IN TIME

Non-verbal behaviour score

Time (seconds)	Rater 1	Rater 2	Rater 3
1-15	66.43	65.33	64.11
16-30	45.98	54.88	67.35
31-45	67.98	66.98	67.35
46-60	76.99	64.87	76.23
61-75	34.55	56.72	61.87
76-90	78.99	57.91	65.98



COHEN'S KAPPA

- Percentage agreement
 - Does not take chance into account
 - Cohen's Kappa
 - The original measure by Jacob Cohen
 - Accounts for chance
- $$\kappa = 1 - \frac{1 - p_o}{1 - p_c}$$
- p_o is the relative observed agreement
 p_c probability of chance agreement
- Limitations
 - Can be biased in certain circumstances
 - Only nominal data - Only 2 raters
 - Requires fully-crossed designs

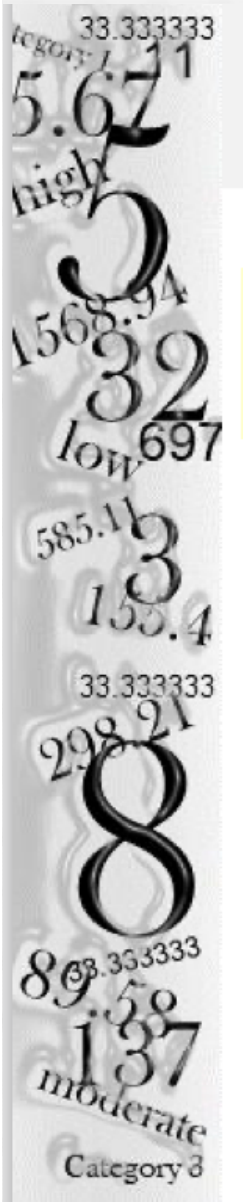


COHEN'S WEIGHTED KAPPA

- There are many variations on Cohen's Kappa
- Cohen's Weighted Kappa
 - Provided by Cohen
 - The weights are the penalties given for different ways of disagreeing
 - The weighting makes it suitable for ordinal data.
 - Many possible weightings
- Limitations
 - Only 2 raters
 - Equivalent to two-way mixed, single-measures, consistency ICC (just use ICC?)



FLIESS' KAPPA



- There are many variations on Cohen's Kappa
- Fleiss' Kappa
 - Provided by Joseph Fleiss
 - Two or more raters
 - Raters are ample from a larger population and new ones are sampled for each subject
- Limitations
 - Only nominal data
 - Assumes a new sample of raters for each subject
 - Therefore not useful for fully crossed designs
 - Might be good for recaptcha traffic lights

INTRACLASS CORRELATION

- ICC - Intraclass correlation
- Flexible
 - Can be used for ordinal, interval or ratio data
 - Two or more coders
 - Can do fully crossed where all subjects are rated by multiple raters
 - Also not fully -crossed where multiple raters rate a subset and a single rater rates the rest
 - Values between 0 and 1
 - 1 equals higher reliability/agreement
- Limitations
 - Many forms can be confusing (6 or more forms)
 - Some implementations deal poorly with missing data



INTRACLASS CORRELATION

- Hallgren (2012) provides a nice tutorial
- Four decisions need to be made before running an ICC:
 - Which model: One-way or Two-way?
 - Absolute agreement or consistency?
 - Unit of Analysis: Average-measures or Single-measures?
 - Random or fixed effects?



Open ICCData.csv

**Using the irr package
conduct an intra-class correlation**

You will need to know the options:

Oneway or twoway?

Absolute agreement or consistency?

Average measures or single measures?

Random or fixed effects?

**You can answer two of these from looking directly
at the data can you work out which of the two
need more information?**

Two-way, consistency, average

```
icc_result <- icc(ICCData, model="twoway", type="consistency",unit="average")
```

Average Score Intraclass Correlation

Model: twoway

Type : consistency

Subjects = 12

Raters = 26

ICC(C,26) = 0.947

F-Test, $H_0: r_0 = 0$; $H_1: r_0 > 0$

$F(11,275) = 18.9$, $p = 5.51e-28$

95%-Confidence Interval for ICC Population Values:

$0.892 < ICC < 0.982$

Intraclass Correlation Coefficient

	Intraclass Correlation ^b	95% Confidence Interval		F Test with True Value 0			
		Lower Bound	Upper Bound	Value	df1	df2	Sig
Single Measures	.407 ^a	.241	.675	18.862	11	275	.000
Average Measures	.947	.892	.982	18.862	11	275	.000

Two-way random effects model where both people effects and measures effects are random.

- The estimator is the same, whether the interaction effect is present or not.
- Type C intraclass correlation coefficients using a consistency definition-the between-measure variance is excluded from the denominator variance.

McGraw and Wong (1996)

Selecting an ICC

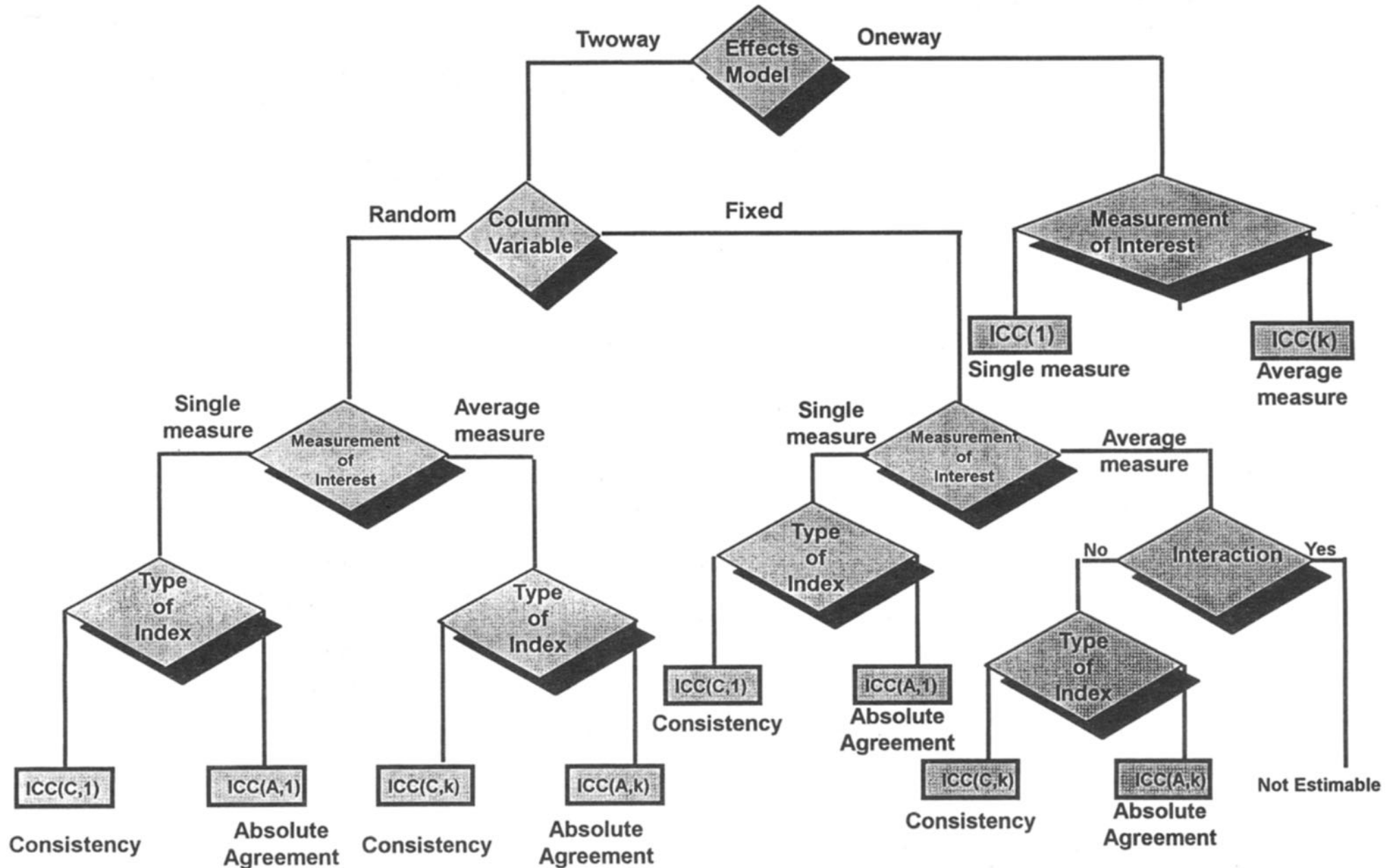
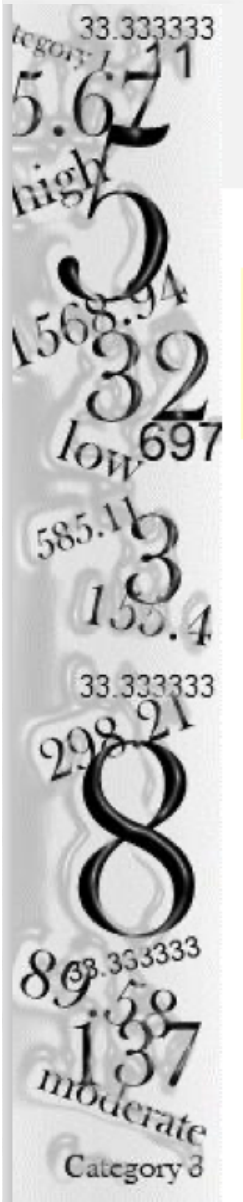


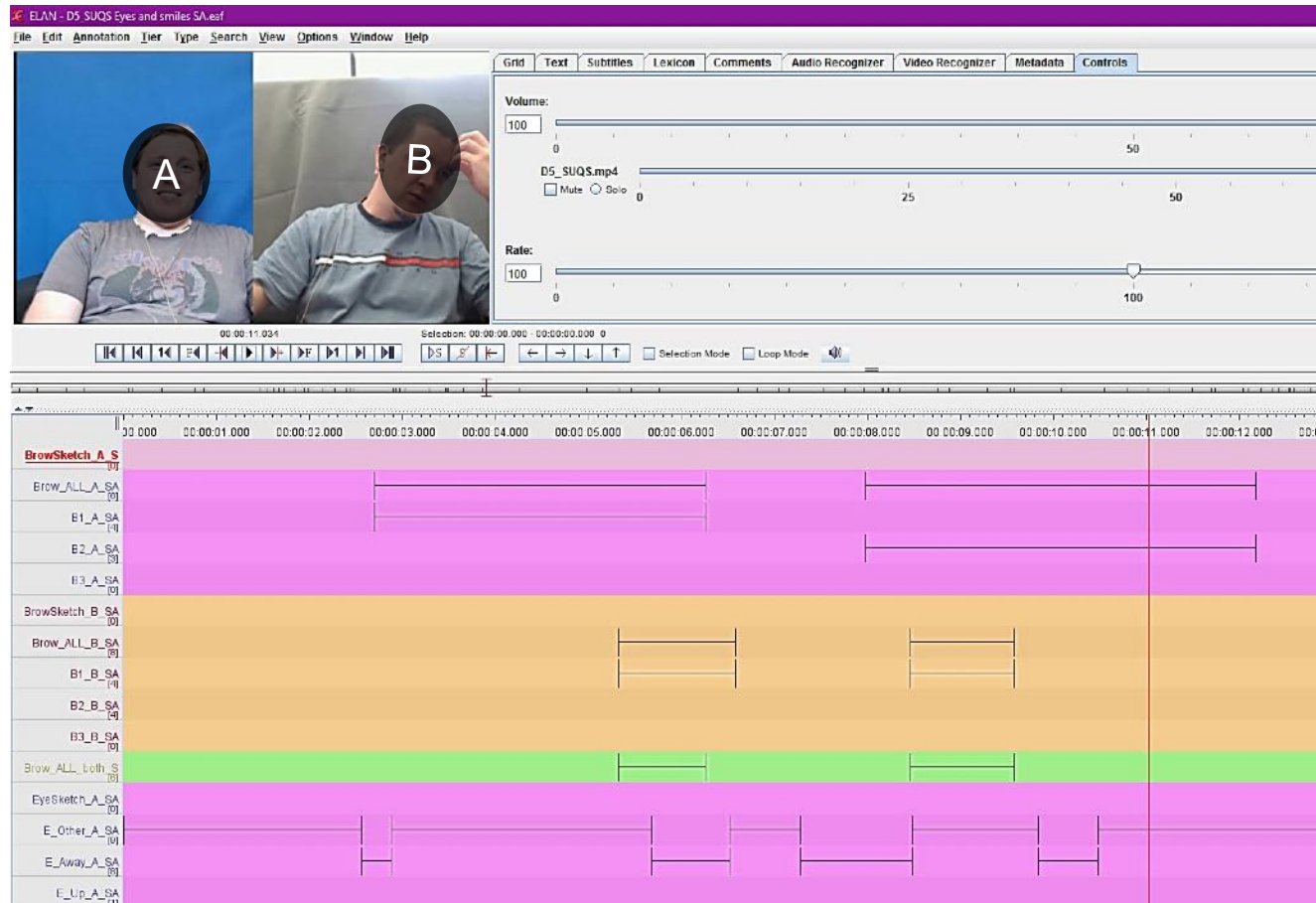
Figure 1. Flow chart for selecting an appropriate intraclass correlation coefficient (ICC).

KRIPPENDORF'S ALPHA



- The most flexible in terms of data types
- Works for nominal, ordinal, interval and ratio data
- Good at handling missing data
- Values between 0 and 1
 - 1 equals higher reliability/agreement
- Allows comparison across data types
- Limitations
 - Has not been widely adopted
 - Less flexible than ICC for continuous data

ELAN manual annotation software



<http://tla.mpi.nl/tools/tla-tools/elan/>

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