

Inter-Annotator Agreement I

Cohen's Kappa and Basic Agreement Measures

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Today's Agenda

- ① Why measure agreement?
- ② Observed agreement
- ③ Problems with raw agreement
- ④ Cohen's Kappa (2 annotators)
- ⑤ Calculating Kappa step by step
- ⑥ Interpreting Kappa values
- ⑦ Worked examples

Project: IAA evaluation due this week

Assignment: HW 3 assigned

Why Measure Agreement?

IAA serves multiple purposes:

- ① **Quality assessment:** How reliable is our data?
- ② **Task validation:** Is the task well-defined?
- ③ **Guideline evaluation:** Are guidelines clear?
- ④ **Annotator calibration:** Are annotators consistent?
- ⑤ **Upper bound:** What's the best model can achieve?

Key insight: If humans can't agree, models can't learn reliably

Observed Agreement

Simplest measure: Proportion of items where annotators agree

$$A_o = \frac{\text{Number of agreements}}{\text{Total items}}$$

Example:

Item	Ann. 1	Ann. 2	Agree?
1	Positive	Positive	✓
2	Negative	Negative	✓
3	Positive	Neutral	×
4	Negative	Negative	✓
5	Neutral	Neutral	✓

$$A_o = \frac{4}{5} = 0.80$$

Problem: Chance Agreement

Observed agreement doesn't account for chance

Example: Binary task (Yes/No)

- Two annotators randomly guessing
- Each says “Yes” 50% of the time
- Expected agreement by chance: 50%

If we observe 70% agreement:

- Seems good, but...
- Only 20% above chance
- Is that meaningful?

Solution: Chance-corrected measures like Kappa

Cohen's Kappa: The Formula

Kappa corrects for chance agreement

$$\kappa = \frac{A_o - A_e}{1 - A_e}$$

Where:

- A_o = Observed agreement
- A_e = Expected agreement by chance
- κ = Kappa coefficient

Intuition: Agreement beyond chance, normalized by maximum possible agreement beyond chance

Computing Expected Agreement

A_e based on marginal distributions

Confusion matrix:

	Ann. 2: Yes	Ann. 2: No	Total
Ann. 1: Yes	a	b	$a + b$
Ann. 1: No	c	d	$c + d$
Total	$a + c$	$b + d$	n

$$A_e = P(\text{both Yes}) + P(\text{both No})$$

$$A_e = \frac{(a + b)(a + c)}{n^2} + \frac{(c + d)(b + d)}{n^2}$$

Step-by-Step Kappa Calculation

Example: Sentiment annotation (Pos/Neg)

	Ann. 2: Pos	Ann. 2: Neg	Total
Ann. 1: Pos	20	5	25
Ann. 1: Neg	10	15	25
Total	30	20	50

Step 1: Observed agreement

$$A_o = \frac{20 + 15}{50} = 0.70$$

Step 2: Expected agreement

$$A_e = \frac{25 \times 30}{50^2} + \frac{25 \times 20}{50^2} = 0.30 + 0.20 = 0.50$$

Step 3: Kappa

$$\kappa = \frac{0.70 - 0.50}{1 - 0.50} = \frac{0.20}{0.50} = 0.40$$

Interpreting Kappa Values

Common interpretation guidelines (Landis & Koch, 1977):

Kappa	Interpretation
< 0	Less than chance agreement
0.00 – 0.20	Slight agreement
0.21 – 0.40	Fair agreement
0.41 – 0.60	Moderate agreement
0.61 – 0.80	Substantial agreement
0.81 – 1.00	Almost perfect agreement

Caution: These are rough guidelines, not absolute thresholds

What Kappa Should You Aim For?

Depends on the task:

- **Objective tasks** (POS tagging): $\kappa > 0.80$
- **Standard annotation** (NER, sentiment): $\kappa > 0.70$
- **Subjective tasks** (quality, safety): $\kappa > 0.50$
- **Highly subjective** (sarcasm, humor): $\kappa > 0.40$

Context matters:

- Compare to published baselines for similar tasks
- Consider task difficulty
- Low Kappa may indicate unclear guidelines, not bad annotators

When Kappa Can Be Misleading

Kappa paradoxes:

① High agreement, low Kappa:

- When one category dominates (class imbalance)
- High A_e reduces Kappa

② Symmetric vs. asymmetric disagreement:

- Kappa treats all disagreements equally
- Some disagreements may be more problematic

③ Prevalence effect:

- Rare categories have outsized impact

Recommendation: Report Kappa AND confusion matrix

Kappa for Multi-Class

Same formula, larger confusion matrix

For k categories:

$$A_o = \frac{1}{n} \sum_{i=1}^k n_{ii}$$

$$A_e = \sum_{i=1}^k \frac{n_{i.} \times n_{.i}}{n^2}$$

Where n_{ii} is the count in cell (i, i) , and $n_{i.}$, $n_{.i}$ are row and column totals.

Multi-Class Example

3-class sentiment: Positive / Neutral / Negative

	Pos	Neu	Neg	Total
Pos	30	5	5	40
Neu	3	20	2	25
Neg	2	3	30	35
Total	35	28	37	100

$$A_o = \frac{30+20+30}{100} = 0.80$$

$$A_e = \frac{40 \times 35}{10000} + \frac{25 \times 28}{10000} + \frac{35 \times 37}{10000} = 0.339$$

$$\kappa = \frac{0.80 - 0.339}{1 - 0.339} = 0.697 \text{ (Substantial agreement)}$$

scikit-learn implementation:

```
from sklearn.metrics import cohen_kappa_score

ann1 = ['pos', 'neg', 'pos', 'neg', 'neu']
ann2 = ['pos', 'neg', 'neu', 'neg', 'neu']

kappa = cohen_kappa_score(ann1, ann2)
print(f"Kappa: {kappa:.3f}")
```

Output: Kappa: 0.600

What to include in your report:

- ① **Kappa value** with interpretation
- ② **Observed agreement** for context
- ③ **Confusion matrix** for detailed view
- ④ **Per-class agreement** if relevant
- ⑤ **Sample size** (number of items annotated)
- ⑥ **Number of annotators**

Example: “Two annotators labeled 500 items for sentiment. Cohen’s Kappa was 0.72 (substantial agreement) with observed agreement of 85%.”

What If Agreement Is Low?

Diagnosis and remediation:

① Unclear guidelines:

- Add examples and edge cases
- Clarify ambiguous definitions

② Insufficient training:

- More annotator calibration sessions
- Practice on sample data

③ Task too subjective:

- Simplify categories
- Accept lower agreement threshold

④ Bad annotators:

- Review individual performance
- Remove or retrain outliers

Lecture 19 (Mar 25): Inter-Annotator Agreement II

Topics:

- Fleiss' Kappa (multiple annotators)
- Krippendorff's Alpha
- Agreement for spans and rankings
- Human-LLM agreement measurement
- LLM self-consistency as quality signal

Reading: Artstein (2017) - Handbook chapter on IAA

Key Takeaways

- 1 **IAA** measures annotation quality and task clarity
- 2 **Observed agreement** doesn't account for chance
- 3 **Cohen's Kappa** corrects for chance agreement
- 4 $\kappa = \frac{A_o - A_e}{1 - A_e}$ measures agreement beyond chance
- 5 **Interpretation** depends on task type
- 6 **Low agreement** may indicate guideline or task problems

Questions?

Office Hours: Wednesdays 1-3pm, Volen 109

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