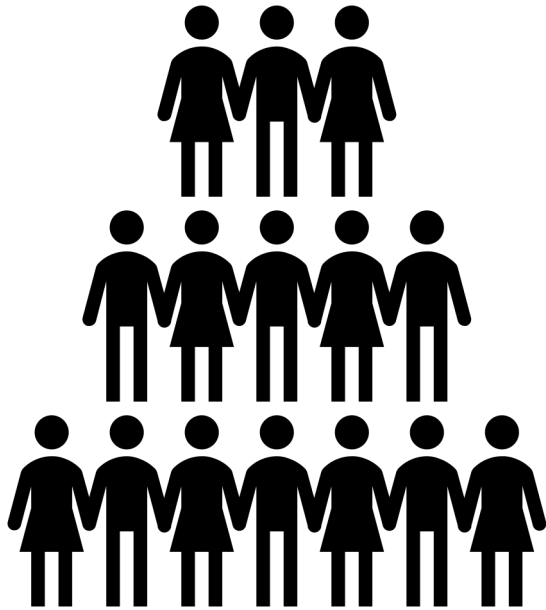


# **Creating Sentiment and Affect Lexicons**

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# How to build a lexicon?

- Expert labels
  - Very reliable 😊
  - Costly and time-consuming 😞
- Crowdsourced labels
  - Less reliable 😞
  - Inexpensive and quick 😊



# Crowdsourcing Resources

- **Amazon Mechanical Turk:**
  - <https://www.mturk.com>
- **Appen:**
  - <https://appen.com>
- **Your own online survey**

# Annotation Schemata

- What kinds of labels will be permissible for your annotators?
- How will they know which labels to select?

**Positive:** A word that evokes a happy or content emotion.

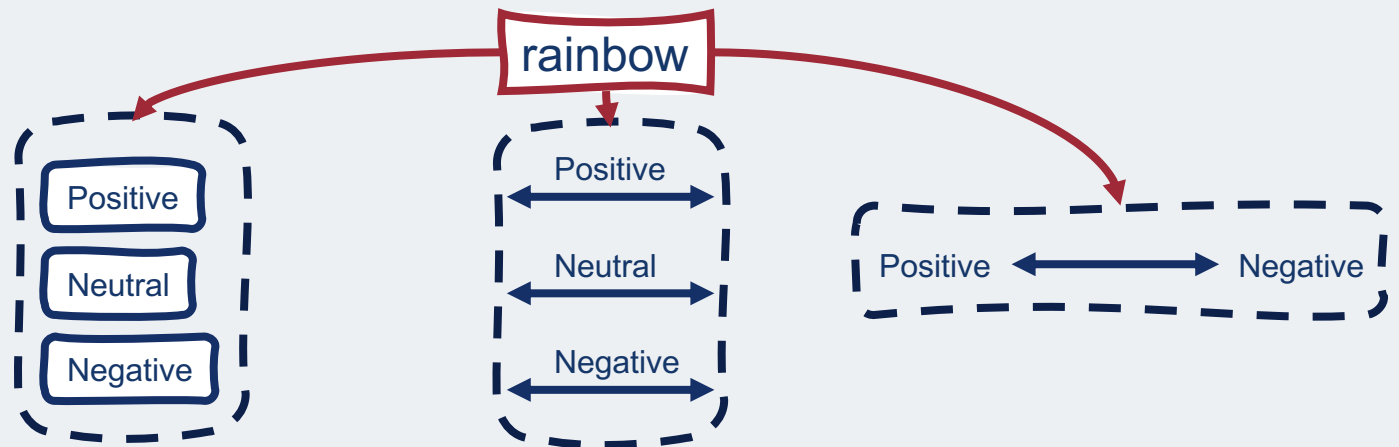
Examples: *love, great, happy*

**Neutral:** A word that does not particularly evoke any emotion.

Examples: *pencil, refrigerator, khaki*

**Negative:** A word that evokes a sad or angry emotion.

Examples: *violence, evil, upset*



# Adjudication

- Third-party adjudicator
- Majority label
- Average label



# Semi-Supervised Induction of Affect Lexicons

- **Semi-supervised label induction:** The process of labeling new, unlabeled instances based on their similarity to instances in a small, labeled seed set
- Two main families:
  - **Axis-based** induction
  - **Graph-based** induction

# Axis-Based Lexicon Induction

- Given a seed set, how similar is the instance to positive instances and how different is it from negative instances?
  - Compute an embedding for each seed word
  - Find the centroid of the embeddings for positive words, and the centroid of the embeddings for negative words
    - $V^+ = \frac{1}{n} \sum_n E(w_i^+)$
  - Compute the axis by subtracting one centroid from another
    - $V_{axis} = V^+ - V^-$
  - Compute the similarity between a given word embedding and the axis
    - $\text{score}(w) = \cos(E(w), V_{axis}) = \frac{E(w) \cdot V_{axis}}{\|E(w)\| \|V_{axis}\|}$
- Higher similarities indicate closer alignment with the positive class

# Graph-Based Lexicon Induction

- Given a graph that connects words with their nearest neighbors, how likely is it that a random walk from a positive word ends on the given word?
  - Define a graph that connects each word to its  $k$  nearest neighbors, with edges weighted by word similarity
  - Identify words in the graph belonging to a labeled seed set
  - Starting at a word from the seed set, perform an edge-weighted random walk
  - Assign an unlabeled word's score based on the probability of landing on it during a random walk from a positive seed and a random walk from a negative seed
    - $$\text{score}^+(w_i) = \frac{\text{score}^+(w_i)}{\text{score}^+(w_i) + \text{score}^-(w_i)}$$
  - Repeat multiple times using bootstrapping, and assign confidence to word scores based on their standard deviation across multiple runs