

### Programming:

- Functional Programming
- Closures
- Decorators
- Object oriented Programming
- Strategy Pattern

### Developing:

- Using LLM
- Pipelines

### Architecture:

- Local LLM
- Pipeline
- Folders
- Check Tasks

### Professional Knowledge:

- Pair Programming

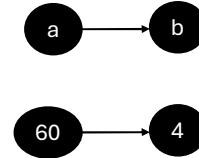


## Methods/ Functions

- A function maps an archetype to an image.

$$f: \frac{b}{15} \quad b = 60; \\ f(b) = 4$$

- Methods/functions are first class members in Python



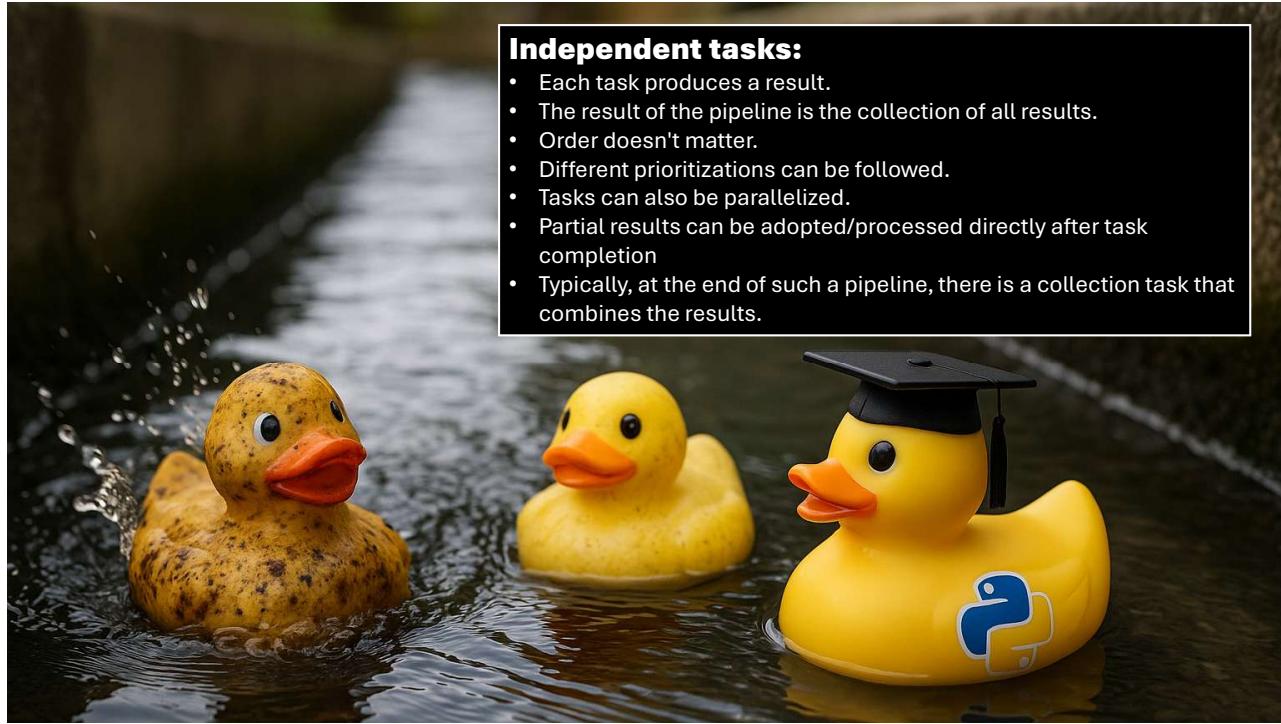
```

def pipeline_chained(s: str, funcs: list) -> None:
    """
    Applies each function on result of last
    application and prints out current state.
    """
    current = s
    for f in funcs:
        current = f(current)
        print(current)
  
```



## Pipelines:

- Collection of tasks
- One task after the other is completed.
- Each task maps an archetype to an image.
- Classic implementation:  
`for n in listTasks:`
- A function maps an archetype to an image.

**Independent tasks:**

- Each task produces a result.
- The result of the pipeline is the collection of all results.
- Order doesn't matter.
- Different prioritizations can be followed.
- Tasks can also be parallelized.
- Partial results can be adopted/processed directly after task completion
- Typically, at the end of such a pipeline, there is a collection task that combines the results.

**Dependent tasks:**

- The overall pipeline produces a result, individual pipeline steps produce an intermediate result.
- Pipelines can be interrupted and continued, but the result can only be adopted or further processed after a complete run
- Changes in the order have an impact on the overall result
- Pipelines, but not tasks, can be parallelized.

Example:

- A text is translated into 4 languages

Example:

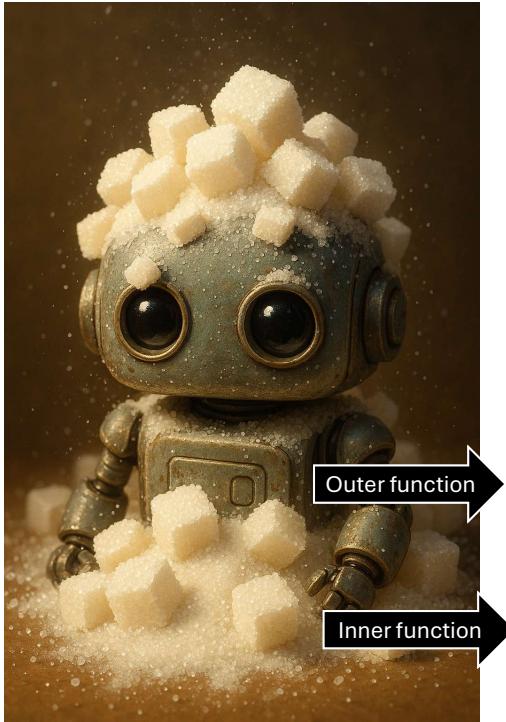
- A manuscript page is scanned
- The text is read via OCR
- The OCR-text is annotated in TEI,
- The TEI is transformed to HTML
- The HTML is published on a web-page

**Closure:**

- A method that returns a method.
- The outer function returns an inner function.
- Inner function keeps outer function's variables alive (Scope).
- After the outer function returns, that state persists.
- Re-calling the inner function accesses or updates it.
- Passing/returning the inner function moves its closed-over scope.
- Enables encapsulation, memoization, and function factories.

```
def make_appender(letter: str) -> Callable[[str], str]:
    """
    Returns a function that will append
    the letter at the end of a string.
    """
    def appender(s: str) -> p:
        return s + letter
    return appender
```





## Syntactic Sugar:

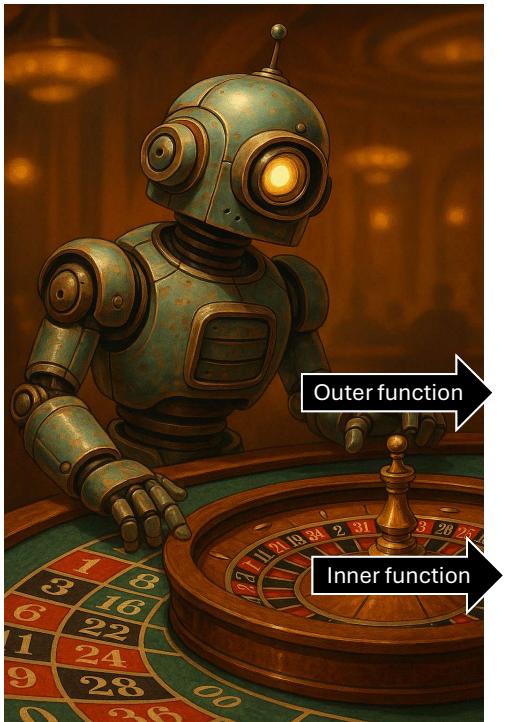
- A decorator is similar to a closure.
- It wraps a function inside another function.
- It can be used by 'decorating a function' by writing '@' followed by the name of the decorator function above the method-statement e.g. @make\_double\_run\_method

## Debugging with assert

The result of a pipeline (or a pipeline-task) can be checked with assert.

```
def make_double_run_method(func: Callable[[str], str]) -> str:
    """Apply func twice on string.

    def wrapper(s: str) -> p:
        result = func(s)
        result = func(result)
        return result
    return wrapper
```



## Probabilistic Steps

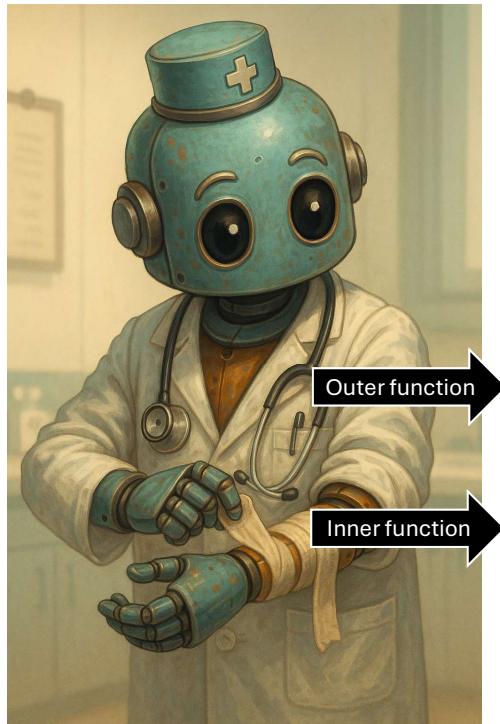
- Not all tasks have deterministic outcomes.
- Input data can lead to different results, depending on the values provided.
- Nondeterministic methods (e.g., LLMs) will produce slightly different results each time they are called.
- This behaviour can be simulated by introducing a random element.



```
def make_appender(letter: str, probability: float = 0.7) -> Callable[[str], str]:
    """Returns a function that will append the letter at the end of a string with a certain probability.

    alternatives =
    "abcdefghijklmnopqrstuvwxyzABCDEFGHIJKLMNOPQRSTUVWXYZ".replace(letter.lower(), "")  

    def appender(s: str) -> p:
        # Dice roll
        if random.random() < probability:
            return s + letter
        else:
            return s + random.choice(alternatives)
    return appender
```

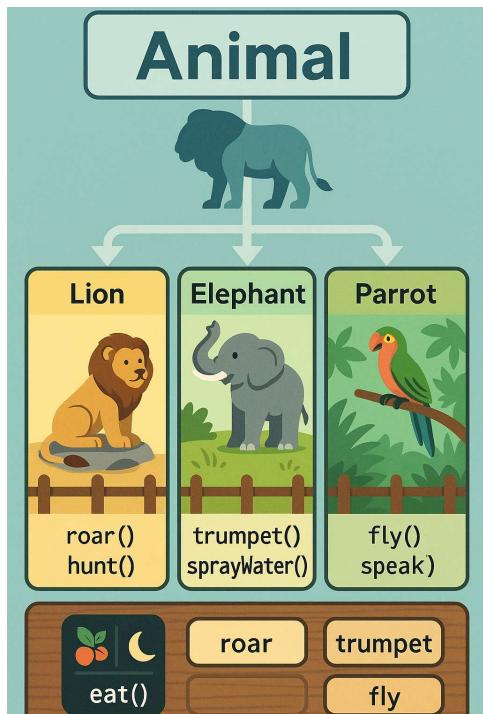


## Self healing loop

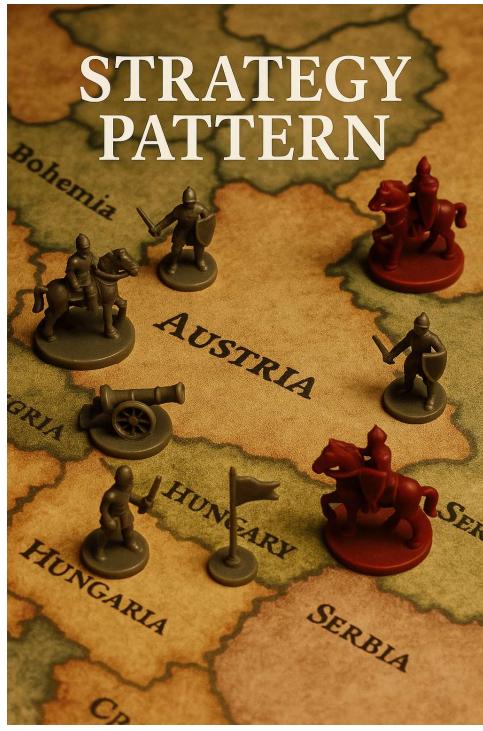
- If a task occasionally produces incorrect output, the simplest fix can be to try again until it succeeds.
- Sometimes, though, you also need to tweak certain parameters with each attempt; otherwise the result will stay the same and repeating the task would be pointless.



```
def repeat_until_condition_met(func: Callable[[str], str], condition: Callable[[str], bool]) -> Callable[[str], str]:
    """
    Returns a function that calls 'func' until the condition
    'condition' is met.
    """
    def wrapper(s: str) -> p:
        result = func(s)
        while not condition(result):
            print(f"Condition not met, repeating:
{result}")
        result = func(s)
    return wrapper
```



- Instances of Classes** – Each object (e.g., `ToLower()`) is a concrete instance of a class, bundling its specific data and the process() behavior it implements.
- Encapsulation of State and Behavior** – Encapsulate the algorithm's code inside their process(`input: str`) → `str` method, hiding implementation details from the client.
- Polymorphic Interaction** – The Pipeline object holds a list of PipelineStep objects and invokes each one via the same process() interface, regardless of the concrete class.
- Collaborating Objects** – The Pipeline instance (the context) and the various PipelineStep instances collaborate: the pipeline delegates work to each step object rather than containing the algorithms itself.
- Dynamic Composition** – Client code creates and assembles objects (e.g., `steps = [ToLower(), ReverseString(), ...]`) at runtime, showing how objects can be composed and reconfigured without changing their classes.



- **Abstract Strategy:** The PipelineStep abstract base class defines the process method.
- **Concrete Strategies:** Classes like ToLower each override process() in a specific way.
- **Context:** The Pipeline class holds a list of PipelineStep instances and delegates each input string to them.
- **Algorithm Encapsulation:**
  - No large conditional Statements
  - Self contained algorithms
- **Open/Closed Principle:** Adding a new transformation makes no changes to existing pipeline or client code.
- **Runtime Flexibility:** Steps in Pipeline can be changed during runtime..

`process(input: str) → str`



- **Shuffle-Based Exploration:** In `train_chain()`, the pipeline repeatedly calls `random.shuffle` on the *original* step list to explore different execution orders .
- **Brute-Force Permutation Search:** Each shuffled sequence is tested against the target until the correct transformation order is found .
- **Isolation of Attempts:** Every trial uses fresh shuffles of the *original* step list, ensuring no cumulative side-effects influence future attempts .
- **Logging Feedback:** Each attempt logs its sequence number and the resulting string, providing visibility into the convergence process .
- **Dynamic Strategy Update:** Once a successful permutation is discovered, `self.steps` is updated to that winning order, fixing the learned strategy for subsequent runs .





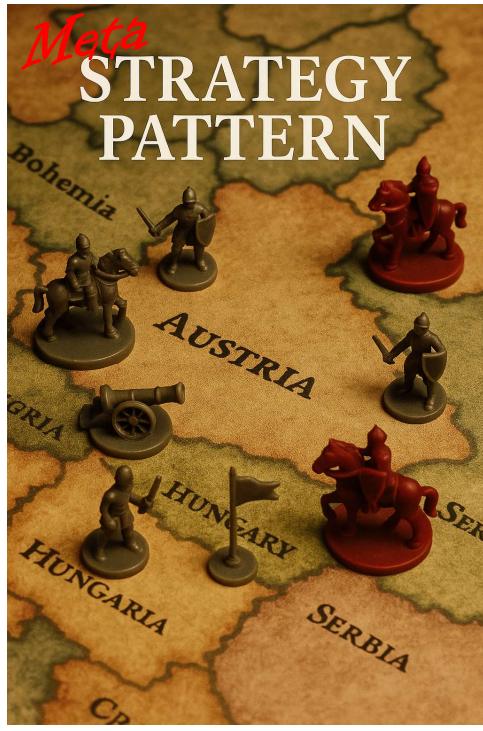
- **Random Permutation Generation:** Inside an infinite loop, `random.shuffle` creates a new ordering of the step signatures each iteration .
- **Signature-Based Deduplication:** A tested set of signature tuples prevents retesting identical sequences, reducing redundant work .
- **Early Exit on Success:** The loop breaks as soon as the shuffled order produces the target output, minimizing wasted attempts .
- **Fine-Grained Logging:** Each attempt prints both the intermediate string result and the specific signature used, making each trial's behavior traceable .
- **Probabilistic Convergence:** Success depends on chance orderings; the algorithm may find a solution quickly or require many random trials, modeling a learning process with unpredictable convergence time.



1	2	3	4	5	6	7	8
9	10	11	12	13	14	15	
16	17	18	19	20	21	22	23
24	25	26	27	28	29	30	31
32	33	34	35	36	37	38	39
40	41	42	43	43	44	45	46
47	48	49	50	73	52	53	54
55	56	57	58	59	60	61	62
63	64	65	66	67	68	69	70
71	72	73	74	75	76	77	78
79	80	81	82	83	84	85	86
87	88	89	90	91	92	93	94
95	96	97	98	99	100	111	114

- **Systematic Permutation Enumeration:** Uses `itertools.permutations` to generate every possible ordering of the steps exactly once .
- **No Redundancy Handling Needed:** Because permutations are inherently unique, there's no need for a separate deduplication structure .
- **Guaranteed Termination:** The search will always find the correct sequence within at most  $n!$  attempts, providing a deterministic bound on run time .
- **Concise Progress Logging:** Each attempt logs only the attempt count and resulting string, offering a clear, minimal trace of the exhaustive search .
- **Reproducible Strategy Discovery:** Because the search order is fixed, the same run will always test permutations in the same sequence and find the solution after the same number of steps.





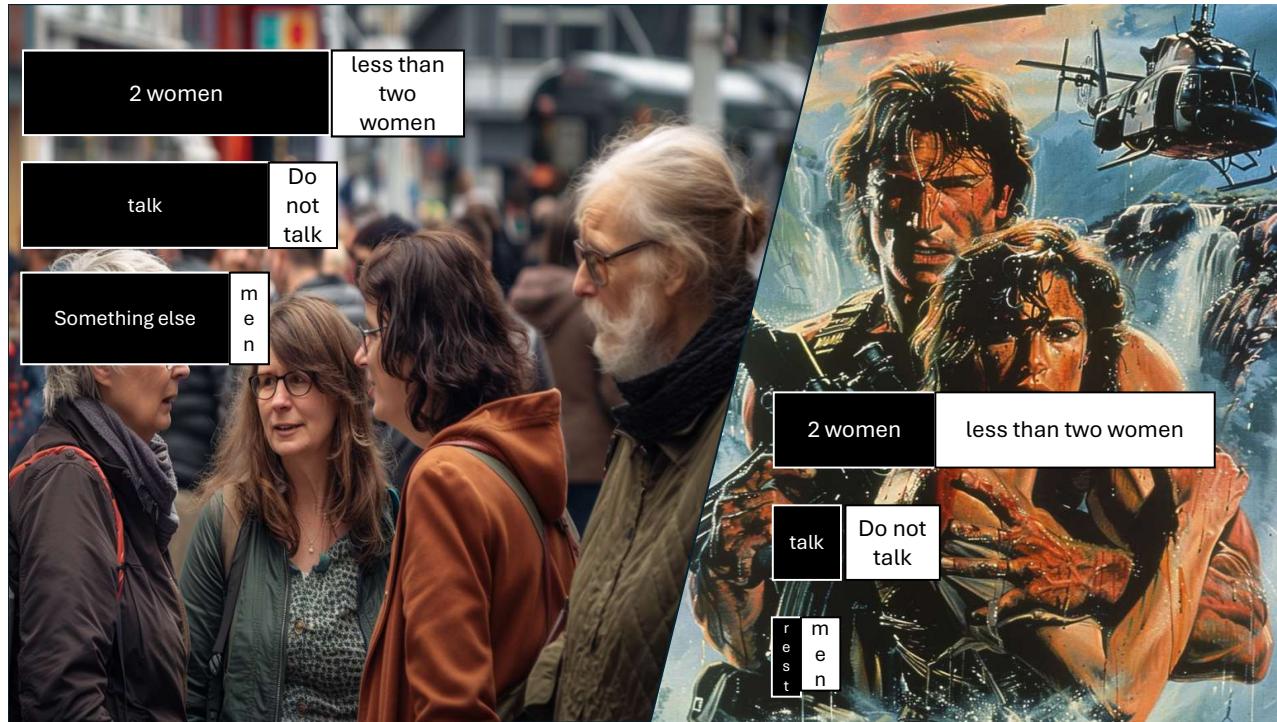
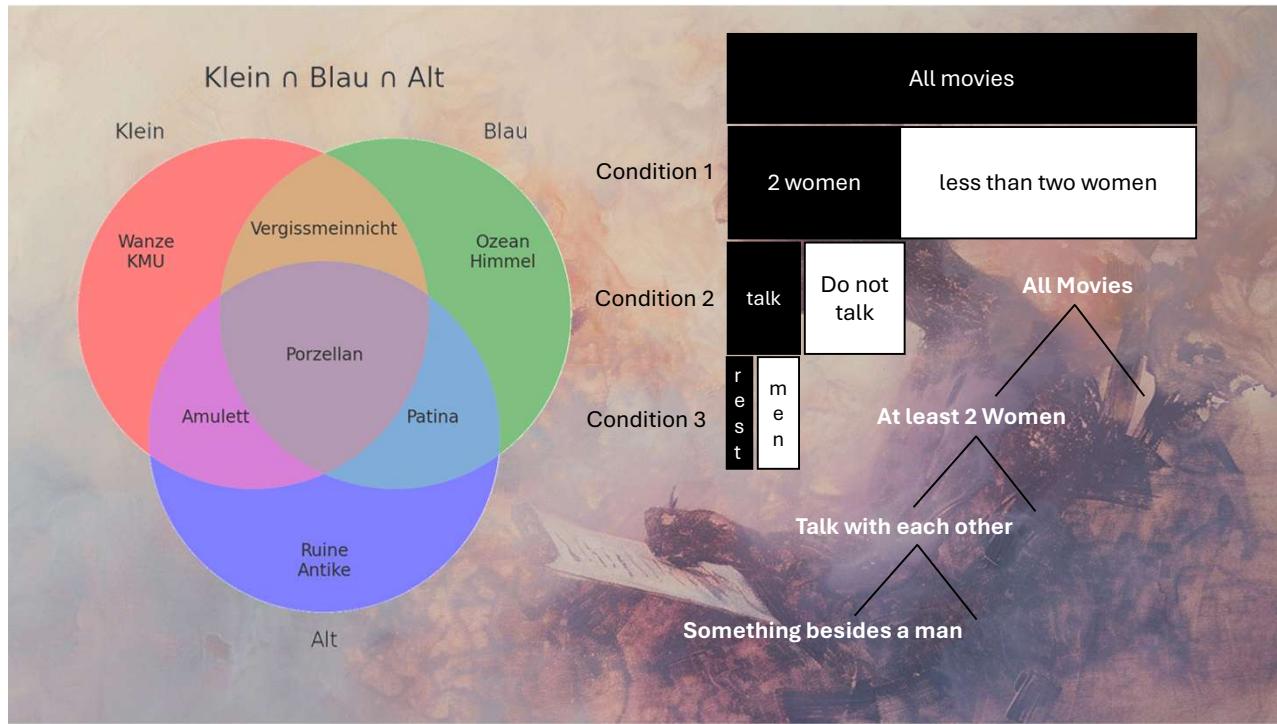
- **Strategies as First-Class Steps:** Wrappers like RepeatUntilConditionMet and DoubleRunMethod implement PipelineStep. Learning policies are just like any other transformation.
- **Probabilistic Append:** UnreliableMakeAppender injects randomness into process().
- **Looping Wrapper:** RepeatUntilConditionMet repeatedly calls its inner step on the original input until the success predicate holds, encapsulating the retry logic.
- **Decorator for Insight:** DoubleRunMethod runs a step twice in a row and logs both outcomes, letting you observe the learning variability side by side.
- **Composable Learning Pipeline:** Meta-strategies are plugged into Pipeline. Transformations and learning behaviors in one cohesive sequence.



- **Pipelines are similar to Tasks** – a task and a pipeline take a string and return a string.  
**Pipelines are similar to Tasks.**
- **Nested Pipelines as Steps** – Pipeline instance (pipeline\_Double\_e) can itself included in the top-level steps list.  
**Hierarchical Composition** – The top-level pipeline composes simple steps (DoubleLastChar, LastBecomesFirst, etc.) alongside nested pipelines.
- **Chained Execution with Sub-Pipelines** – When run\_chained() executes, it delegates into the nested pipeline at that point.
- **Complexity** – Objects store states and can provide multiple methods. E.g. Visualization.
- **Uniform Interface** – Both individual strategy steps and entire pipelines share the PipelineStep interface (process() and \_\_repr\_\_()), ensuring clients and visualizers treat them interchangeably.
- **Graphical Visualization** – Nested execution works for display.









## Double Keying

- Bearbeiter:innen machen unabhängig voneinander Fehler.
- Fehler sind selten.
- Die Wahrscheinlichkeit, dass beide Bearbeiter:innen an gleicher Stelle einen Fehler machen ist verschwindend gering

- Fehlerwahrscheinlichkeit pro Bearbeiter:  $p$
- Übereinstimmung

$$P_{\text{agree}} = (1 - p)^2 + p^2$$

- Uneinigkeit (Prüffall)

$$P_{\text{disagree}} = 2p(1 - p)$$

- Korrektheits-Wahrscheinlichkeit bei Übereinstimmung

$$P(\text{korrekt} \mid \text{agree}) = \frac{(1 - p)^2}{(1 - p)^2 + p^2}$$

- Beispiel  $p = 0,1$ :

$$P_{\text{agree}} = 0,9^2 + 0,1^2 = 82\%, \quad P(\text{korrekt} \mid \text{agree}) \approx 98,8\%.$$

**Obvious cases**

99% interannotator agreement –  
1% Error rate

**Difficult cases:**

Answers depend on:

- Priming,
- Character
- Obscure rules

*Tendency to go for expected mainstream answer*

### Reasoning

```

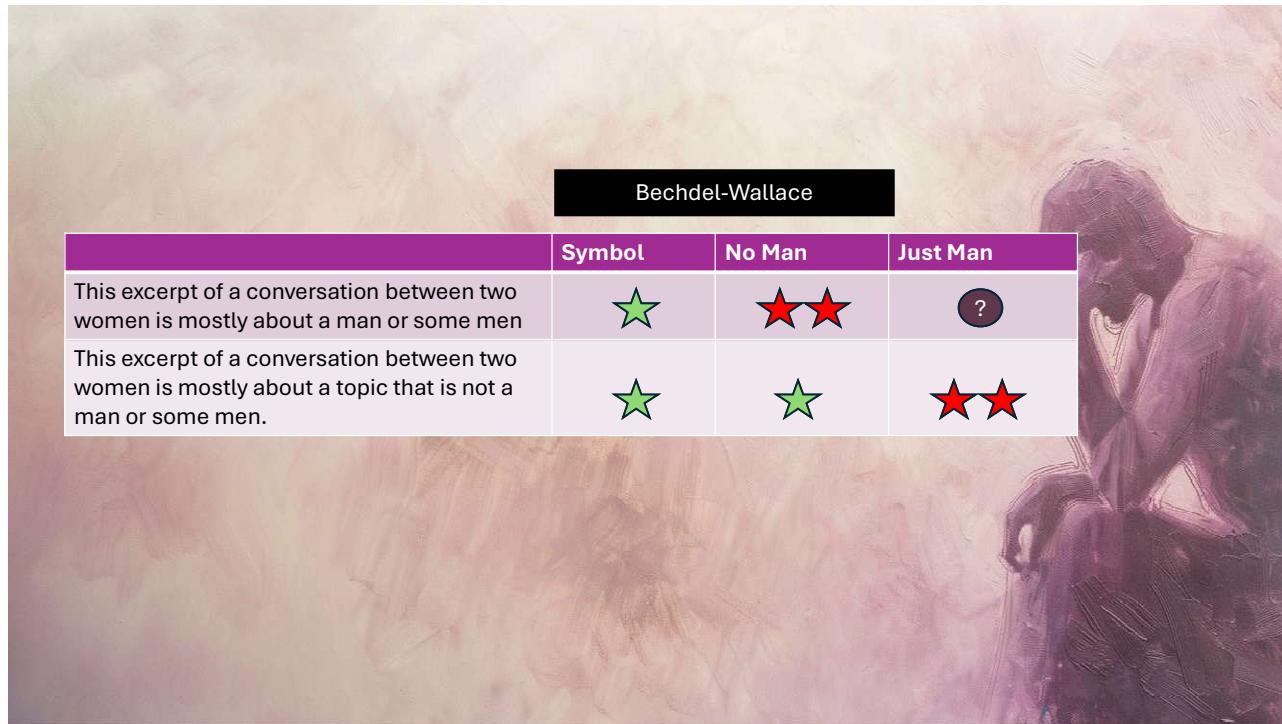
1 {
  "role": "user",
  "content": [
    {"type": "text", "text": "
      f\"Bitte lies diesen Text und notiere, welche Informationen dir zur Verfügung stehen, \"\n
      f\"um die Frage '{self.criterion}' zu beantworten, \"\n
      f\"und welche Informationen fehlen oder erschlossen werden müssen.\n\n"
    }
  ]
}
  
```

```

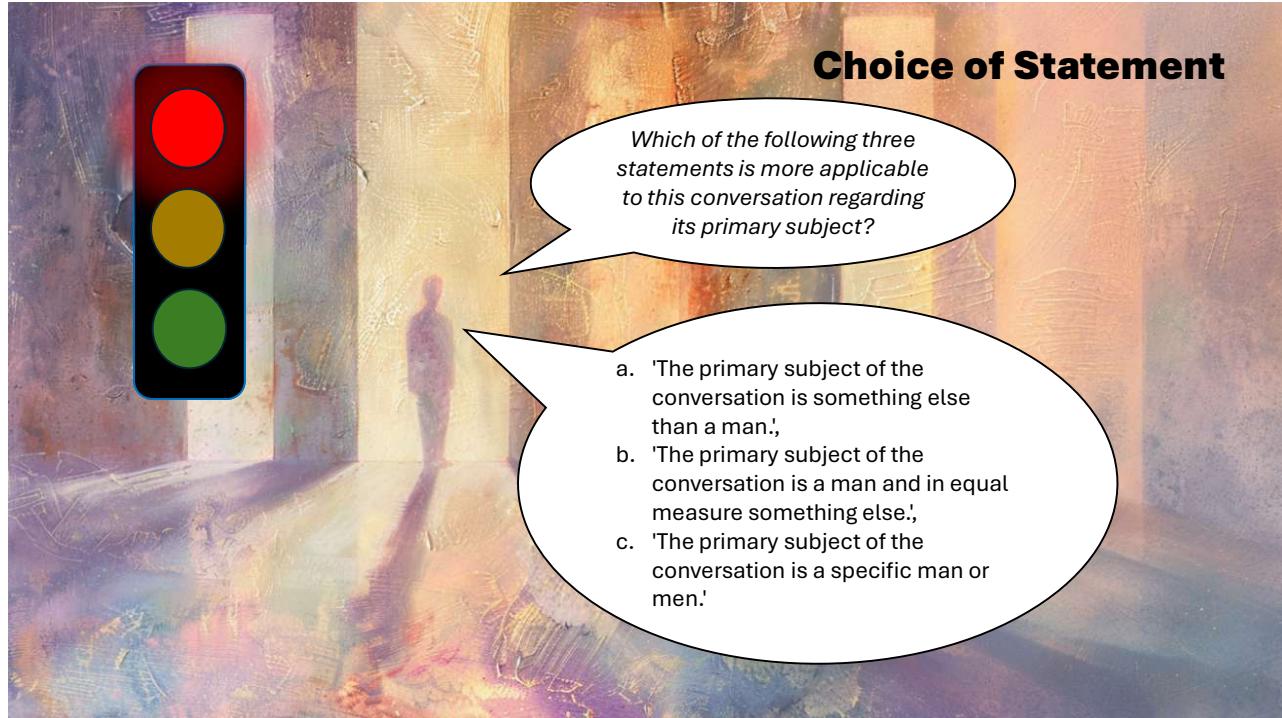
2 {
  "role": "user",
  "content": [
    {"type": "text", "text": f"Originaltext:\n{(conversation)}},
    {"type": "text", "text": f"Analyse des Informationsstands:\n{analysis.strip()}},
    {"type": "text", "text": "
      Formuliere für jede der folgenden Kategorien je eine Aussage im Muster \"\n
      \"\\"<Label>: ...\\n\" + \"\\n\".join(f\"- {lab}\" for lab in label_list)\""
    }
  ]
}
  
```

```

3 {
  "role": "user",
  "content": [
    {"type": "text", "text": f"Originaltext:\n{(conversation)}},
    {"type": "text", "text": f"Analyse der verfügbaren Informationen:\n{analysis.strip()}},
    {"type": "text", "text": f"Statements zu den Kategorien:\n{statements.strip()}},
    {"type": "text", "text": "
      Basierend auf obiger Analyse und den Statements, \"\n
      gib genau **eine** der folgenden Optionen zurück: \"\n
      + \" .join(label_list)
    }
  ]
}
  
```



Bechdel-Wallace			
	Symbol	No Man	Just Man
This excerpt of a conversation between two women is mostly about a man or some men	★	★★	?
This excerpt of a conversation between two women is mostly about a topic that is not a man or some men.	★	★	★★



## Rank texts

(if your context window is large enough)

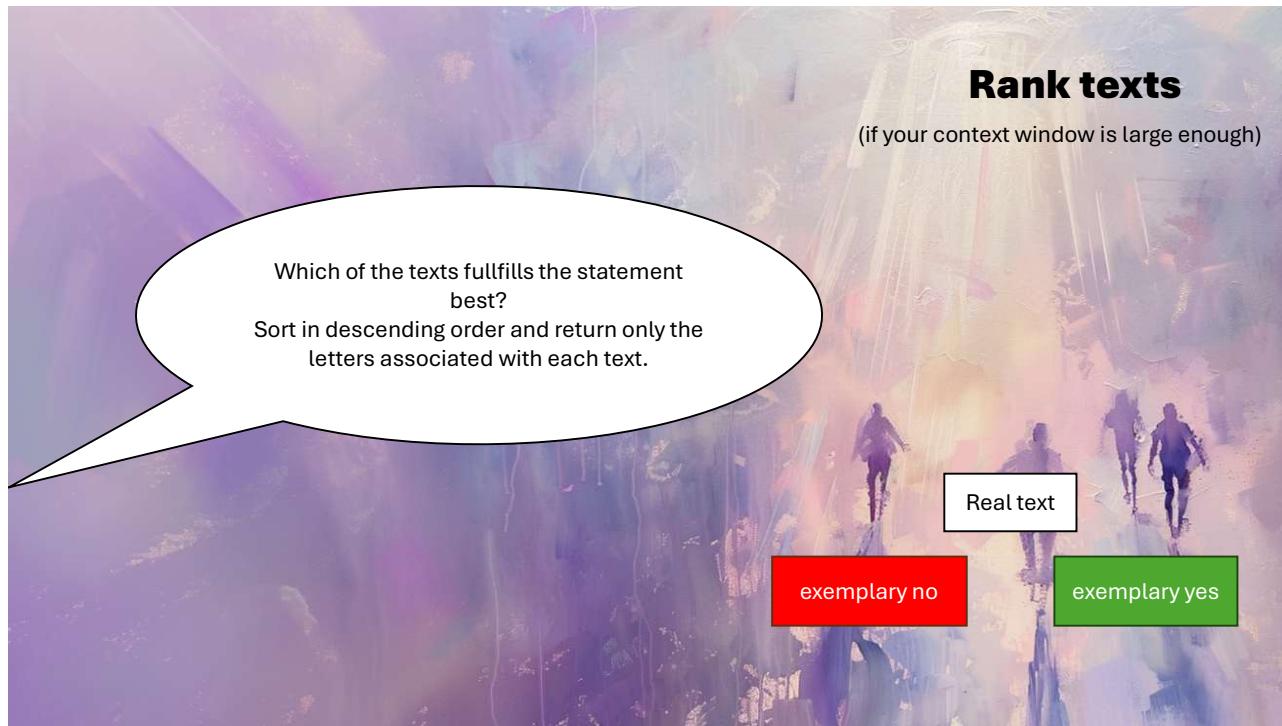
Which of the texts fulfills the statement best?

Sort in descending order and return only the letters associated with each text.

Real text

exemplary no

exemplary yes



## Tournament

Tournament mit Mustertexten



