Argument Extraction Codebook

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0. Basic Definition:

Event: A paragraph in a scientific abstract that conveys one specific type of information as defined in **Event Types**. Each **Event** describes a key aspect of the research clearly and independently.

Event Types:

Background/Introduction: Briefly outlines the context, motivation, and problem being addressed. It highlights the research gap and the paper's objectives or research questions.

Method/Approach: Summarizes the methodologies, frameworks, or techniques used to conduct the study, including experimental setups, algorithms, datasets, or analytical tools.

Results/Findings: Reports the main outcomes of the research, emphasizing key data, trends, or discoveries. Focuses on what was achieved or learned.

Conclusions/Implications: Discusses the significance of the findings, their impact on the field, potential applications, and how they address the initial problem or research gap. May include recommendations or future research directions.

Summarization: A summary that best describe the **Event** within 10 words

Arguments: Combination of Event Participants and Event Attributes.

Event Participants: The entities or participants involved in the Event, categorized by their roles relating to the Main Action. (e.g., agent, object).

Event Attributes: Additional information about the **Event**. In scientific domains, this usually refers to additional scientific information such as Explanation, Results, Implication, etc. Detail definition and examples are provided in the **Event Attributes** section.

1. Annotation Guidelines

- 1. First read the paragraph and provide a **Summarization** that best describes the paragraph.
- 2. Then find an "Verb Phrase" (i.e.: "am, is, are" and Verbs and those phrases behave as a verb) from the paragraph that leads the **Event**. This is called the **Main Action**, and it is usually related to the **Summarization** in semantic meaning.
- 3. Then find the sentence the Main Action is in and in most cases, the Agent and all the Objects are in this sentence. The Agent is the "giver/doer" of the Main Action and the Base and Attached Objects are the "accepters/receivers" of the Main Action, and additionally the Base and Attached Modifier offers additional information for the Base and Attached Objects.

Note that the **Base and Attached Modifier** are always after **Base and Attached Objects** to provide additional information.

Further, in majority cases, there are only **Base Object and Modifier** in the annotation. Please refer to section 3 for examples of when **Attached Object and Modifier** are needed.

This sentence is called the **Main Sentence**, and all **Event Arguments** should be providing additional information about the **Main Sentence**.

- 4. Use the "Main action" to extract other arguments. Then use subarguments below to fit the extracted text and find its argument. Don't infer any information, use only the exact same quote from the given paragraphs. You don't need to fill all 10 categories for each paragraph, but you are encouraged to cover as much information as possible.
- 5. You should always try to extract nouns phrases only, unless one entire sentence is describing one argument, and the sentence is inseparable in meaning. When encountered abbreviation, e.g. "Chain-of-Thought (CoT) improve", you should use both original term and abbreviation: "Chain-of-Thought (CoT)" as the argument.
- 6. In total, you should annotate Event Summarization, Main action, at most 9+2 Arguments (if explicit) for each Event.

2. Event Arguments

Each **Arguments** extracted should be categorized into one of the following 9 categories (plus the **Agent** and **Object**)

- Underscored Indicator Words: Indicator words are underscored for guiding annotation.
- **Highlighted Annotations**: Where to annotated is highlighted for reference.

2.1 Context

- **Definition**: Provides foundational or situational information.
- Subarguments:
 - Historical Context (check with Professor if this format is good)

Definition: Background from prior work or historical developments.

Rationale: Historical Context provides foundational information by connecting the current study to prior developments, situating it within the broader **context** and framing its contributions and relevance.

Indicator word: "builds on," "based on," "pioneered by," "founded on," "inspired by earlier work," "influenced by foundational methods."

Examples:

(NLP): "Builds on Word2Vec by Mikolov et al. (2013), a foundational method for word representations in NLP, we propose our method." (Todo:

indicator word color, then also highlight what to annotate)

(HCI): "This <u>based on</u> Fitts' Law (1954), a key model for predicting human movement in interaction tasks."

(Health Informatics): "Our survey is designed by inspecting early electronic health records (EHRs) systems, pioneered in the 1960s."

(Digital Humanities): "This approach <u>builds on</u> early textual analysis methods developed in the mid-20th century, such as concordance creation." (Computational Biology): "The study is <u>founded on</u> BLAST, a tool pioneered in 1990 for DNA sequence alignment."

• Assumption:

Definition: Underlying ideas accepted as true without direct proof.

Rationale: Assumption identifies pre-existing conditions or untested ideas that the study relies upon, providing critical **context** for understanding the scope and setup of the research.

Indicator Words: "Assumes that", "presumes," "is based on the premise," "underlying assumption," "implicitly assumes," "accepted without validation." **Examples:**

(NLP): "The model <u>assumes</u> independence between tokens in sequence prediction tasks."

(HCI): "The study <u>presumes that</u> users have prior experience with touch-screen devices."

(Health Informatics): "The system relies on the assumption that patients accurately report symptoms in digital surveys."

(**Digital Humanities**): "This analysis <u>presumes that</u> the digitized manuscripts accurately represent the original texts."

(Computational Biology): "The algorithm is <u>based on the premise that</u> genetic variations in the sample set are representative of the larger population."

Condition:

Definition: Specific limits or situational factors present during the study that influence results and define where findings apply.

Rationale: **Condition** clarifies the specific situational scope or constraints that define the study's results, ensuring the findings are interpreted within the appropriate **context**.

Indicator Words: "Only applies to," "limited to," "restricted to," "valid for," "relevant for," "conducted under," "performed in," "with the following conditions," "in a controlled environment," "under standard conditions."

Examples:

(NLP): "The evaluation was conducted on text data <u>limited to publicly</u> available news articles."

(HCI): "Findings are valid for desktop interfaces tested <u>in environments with</u> consistent lighting conditions."

(Health Informatics): "The study is <u>restricted to</u> patient records collected from urban hospitals over a six-month period."

(**Digital Humanities**): "The analysis was <u>performed on digitized literary texts</u> from the 19th century, under standard corpus linguistics procedures."

(Computational Biology): "The algorithm's performance is <u>valid for genomic</u> data sequenced using next-generation sequencing technologies."

• Theoretical Framework:

Definition: Models underpinning the study.

Rationale: **Theoretical Framework** links the study's methods or interpretations to established models or theories, ensuring that the **context** is known to readers.

Indicator Words: "Based on," "grounded in," "applies the theory," "underpinned by," "draws on."

Examples:

(NLP): "This work is <u>based on</u> the <u>Transformer model</u>, which uses selfattention for sequence processing."

(HCI): "The study <u>applies</u> Norman's theory of affordances to evaluate interface usability."

(Health Informatics): "The intervention is grounded in the Health Belief Model, which predicts behavior based on perceived risks and benefits."

(**Digital Humanities**): "The approach <u>draws on Moretti's theory of distant reading</u> to analyze large corpora of texts."

(Computational Biology): "This algorithm is <u>underpinned by</u> the Neutral Theory of Molecular Evolution, which assumes genetic variations occur at constant rates over time."

2.2 Purpose

• **Definition**: Defines the purpose or aim.

• Subarguments:

Goal/Aim:

Definition: The main purpose or objective of the study.

Rationale: **Goal/Aim** describes the overarching intention or end result that the study seeks to achieve, providing clear **purposes** for the research by outlining what it aims to accomplish.

Indicator Words: "Aims to," "seeks to," "goal is," "objective is," "intends to," "purpose of the study."

Examples:

(NLP): "The study <u>aims to improve sentiment classification</u> for low-resource languages."

(HCI): "The <u>goal is to</u> enhance user engagement through adaptive interface designs."

(Health Informatics): "The <u>aim is to</u> evaluate the effectiveness of AI in early disease detection."

(**Digital Humanities**): "This research <u>seeks to</u> uncover patterns in 18th-century literary networks using computational methods."

(Computational Biology): "The <u>objective is to predict protein structures more accurately using deep learning models."</u>

O Hypothesis:

Definition: Statements proposed to explain a phenomenon, tested in the study. **Rationale**: **Hypothesis** specifies testable predictions that guide the study's methods, offering concrete **purposes** by explaining what phenomena the study investigates and seeks to validate.

Indicator Words: "Proposes that," "tests whether," "is expected to," "hypothesizes that," "is assumed to result in," "predicts that."

Examples:

(NLP): "Contextual embeddings are <u>expected to improve entity recognition</u> accuracy."

(HCI): "It is <u>proposed that</u> personalized feedback increases task completion rates."

(Health Informatics): "This study <u>tests whether</u> mobile health interventions reduce patient anxiety levels."

(**Digital Humanities**): "The <u>hypothesis predicts that</u> computational stylistic analysis can distinguish between authors in collaborative texts."

(Computational Biology): "It is <u>hypothesized that</u> integrating epigenetic data improves the accuracy of cancer subtype classification."

2.3 Method

- **Definition**: Techniques, tools, or frameworks used.
- Subarguments:
 - o Methodology:

Definition: General description of methods used.

Rationale: **Methodology** outlines the overall approach or framework used in the study, providing a high-level description of the **methods** applied to achieve the study's goals.

Indicator Words: "Was conducted using," "employed," "utilized," "was analyzed with," "applied methods include," "Updates based on," "refined iteratively," "adjusted using feedback," "loop includes."

Examples:

(NLP): "Transformer-based models <u>were fine-tuned</u> on annotated datasets." (HCI): "Usability testing was conducted with think-aloud protocols."

(Health Informatics): "Patient surveys were analyzed using mixed-methods approaches."

(**Digital Humanities**): "Text analysis <u>was conducted</u> using natural language processing techniques to extract recurring themes."

(Computational Biology): "Genome-wide association studies were conducted using statistical regression models."

Subsequent Methodology:

Definition: Additional or follow-up techniques and procedures applied after the primary methodology.

Rationale: Subsequent methodologies are critical for extending, validating, or refining initial findings. They establish a logical progression in the method and often bridge initial approaches with additional results or validation steps. Indicator Words: "Following this," "subsequent steps include," "in the next phase," "was further refined by," "extended through."

Examples:

(NLP): "Following this, <u>subsequent steps included</u> fine-tuning the model on domain-specific datasets to improve accuracy further."

(HCI): "Subsequent methods involved iterative prototyping and user feedback collection to refine the interface."

(Health Informatics): "In the next phase, machine learning models were retrained with additional features to enhance diagnostic precision."
(Digital Humanities): "After the initial text analysis, subsequent steps

included network analysis to map relationships between authors and works."

(Computational Biology): "The study <u>extended through</u> additional gene expression analyses to validate the primary findings from sequence alignment."

• Algorithm/Process Description:

Definition: Step-by-step technical details.

Rationale: Algorithm/Process Description provides a detailed breakdown of the procedural steps or workflows involved, explaining how the **methods** were implemented technically..

Indicator Words: "Preprocesses by," "includes steps such as," "consists of," "process involves," "workflow includes."

Examples:

(NLP): "The system <u>preprocesses text by</u> tokenizing, normalizing, and embedding inputs before classification."

(HCI): "The interface design <u>process included</u> wireframing, prototyping, and iterative user testing."

(Health Informatics): "The prediction <u>pipeline involves</u> feature extraction, model training, and validation on patient datasets."

(Digital Humanities): "The workflow includes transcribing digitized manuscripts, cleaning OCR errors, and conducting semantic analysis."

(Computational Biology): "The <u>process involves</u> sequence alignment, motif discovery, and phylogenetic tree construction to understand evolutionary relationships."

O Benchmarking:

Definition: Comparisons against standards or other methods.

Rationale: **Benchmarking** evaluates the effectiveness of the study's methods by comparing results against established baselines or alternative approaches, demonstrating the **methods**' relative performance.

Indicator Words: "Evaluated against," "compared to," "benchmark results include," "was measured relative to," "tested alongside."

Examples:

(NLP): "The model was <u>evaluated against</u> BERT and GPT baselines on standard benchmarks."

(HCI): "User performance was <u>compared to</u> <u>established metrics from prior usability studies."</u>

(Health Informatics): "Diagnostic accuracy was <u>benchmarked against results</u> from clinical experts."

(**Digital Humanities**): "The proposed algorithm was <u>compared to</u> <u>existing</u> text-mining tools to assess its efficiency in processing historical corpora." (**Computational Biology**): "The new protein-folding model was <u>tested</u> <u>alongside</u> <u>AlphaFold</u> to evaluate predictive accuracy."

• Implementation:

Definition: Practical realization of theories or tools.

Rationale: Implementation describes how the theories, methods, or frameworks were practically realized, providing concrete details on the **methods**' real-world execution.

Indicator Words: "Implemented using," "was developed as," "deployed with," "realized in," "translated into."

Examples:

(NLP): "The model was <u>implemented using PyTorch</u> and trained on a distributed GPU cluster."

(HCI): "A prototype application was <u>developed using</u> React and deployed for user testing."

(Health Informatics): "The intervention was <u>implemented as</u> a mobile app integrated with existing EHR systems."

(**Digital Humanities**): "The text analysis tool was <u>developed as</u> a <u>Python</u> package and deployed on Jupyter Notebooks for interactive use."

(Computational Biology): "The genome annotation pipeline was implemented using Snakemake and deployed on a high-performance computing cluster."

Optimization:

Definition: Refinements for better outcomes.

Rationale: **Optimization** focuses on adjustments or refinements made to improve the efficiency, accuracy, or performance of the **methods** used in the study.

Indicator Words: "Optimized by," "tuned using," "refined to," "adjusted for," "improved to maximize."

Examples:

(NLP): "Hyperparameters were <u>tuned using grid search</u> to maximize classification accuracy."

(HCI): "Interaction flows were <u>streamlined to reduce task completion time</u>." (Health Informatics): "The algorithm was <u>optimized for faster processing of large-scale patient datasets."</u>

(**Digital Humanities**): "The text-mining model was <u>refined to</u> handle OCR errors, improving the analysis of digitized historical documents."

(Computational Biology): "The protein-structure prediction pipeline was adjusted for better runtime performance on high-throughput datasets."

o Mechanism:

Definition: The step-by-step process explaining how something works. **Rationale**: **Mechanism** provides a detailed breakdown of how methods or processes function, offering a systematic **methods** of why specific approaches produce the observed outcomes.

Indicator Words: "Works by," "operates through," "step-by-step process," "mechanism involves," "functioning of."

Examples:

(NLP): "The attention mechanism <u>assigns</u> weights to tokens, prioritizing more relevant context for predictions."

(HCI): "The recommendation system <u>adjusts</u> displayed content based on realtime user interactions."

(Health Informatics): "The wearable device <u>measures</u> heart rate variability to detect early signs of stress."

(**Digital Humanities**): "The mechanism <u>involves</u> parsing historical texts, tagging named entities, and mapping relationships to create a knowledge graph."

(Computational Biology): "The algorithm works by aligning DNA sequences, identifying conserved motifs, and predicting functional domains."

Validation:

Definition: Checking to make sure the results are correct and reliable.

Rationale: Validation ensures the correctness and reliability of the research by using verification methods to confirm their accuracy and robustness. Indicator Words: "Validated through," "confirmed by," "cross-checked with," "verified using."

Examples:

(NLP): "The model's predictions were <u>validated</u> using cross-validation on three datasets."

(HCI): "Usability findings were <u>confirmed</u> through follow-up user testing sessions."

(**Health Informatics**): "The algorithm's outputs were <u>validated</u> against expertannotated clinical data."

(**Digital Humanities**): "The text classification system was <u>validated</u> through manual review by domain experts."

(Computational Biology): "Predicted protein structures were <u>verified</u> using experimentally resolved crystal structures from the PDB database."

o Metrics:

Definition: Measures used to evaluate the performance or quality of a system or method.

Rationale: Metrics show what **method** is used to measure the results by providing measurable criteria that assess the performance, quality, or effectiveness of the study's methods.

Indicator Words: "Measured by," "evaluated using," "performance was assessed with," "scores indicate."

Examples:

(NLP): "The model's performance was evaluated using accuracy, F1 score,

and BLEU for translation tasks."

(HCI): "User satisfaction was <u>measured through</u> SUS (System Usability Scale) scores and task completion time."

(Health Informatics): "The system's effectiveness was <u>assessed using</u> sensitivity, specificity, and patient adherence rates."

(**Digital Humanities**): "The system's text-mining accuracy was <u>evaluated</u> using precision, recall, and F1 scores."

(Computational Biology): "The effectiveness of the genome analysis pipeline was assessed using ROC-AUC and false discovery rates."

2.4 Results

- **Definition**: Observations or outputs.
- Subarguments:
 - Observation:

Definition: Important findings or patterns noticed during the study.

Rationale: Observation identifies key patterns or findings that emerge from the study's analysis, providing direct **results** that highlight what was discovered.

Indicator Words: "Observed that," "noticed a trend where," "found that," "results showed."

Examples:

(NLP): "The model struggled with rare words, <u>showing</u> lower accuracy for low-frequency tokens."

(HCI): "Users preferred interfaces with fewer menu options, <u>reporting</u> higher satisfaction scores."

(Health Informatics): "We <u>found that</u> patients responded more quickly to reminders sent in the morning compared to the evening."

(**Digital Humanities**): "The analysis <u>showed</u> a trend where 18th-century novels contained increasingly complex social networks over time."

(Computational Biology): "We observed that specific gene clusters were consistently expressed under high-stress conditions in bacterial samples."

• Evidence/Support:

Definition: Data or results used to back up conclusions.

Rationale: Evidence/Support provides concrete data that substantiates the study's findings, presenting measurable results to validate conclusions. Indicator Words: "Achieved," "supported by," "data shows," "results confirm."

Examples:

(NLP): "The model <u>achieved 92%</u> accuracy on the test set, outperforming all baselines."

(HCI): "80% of participants completed the task faster using the new interface design."

(Health Informatics): "Survey data showed a 30% reduction in reported symptoms after the intervention."

(**Digital Humanities**): "Topic modeling results <u>supported by</u> manual annotations revealed recurring themes across historical documents." (**Computational Biology**): "Gene expression analysis <u>confirmed</u> a 40% increase in specific proteins under experimental conditions."

• Classification/Categorization:

Definition: Organizing data into specific groups or labels.

Rationale: Classification/Categorization organizes the study's findings into meaningful groups, providing structured **results** that simplify analysis and interpretation.

Indicator Words: "Organized into," "grouped by," "categorized as," "labeled with."

Examples:

(NLP): "Text data was <u>categorized into</u> sentiment labels: positive, negative, or neutral."

(HCI): "Users were grouped based on their interaction styles: novice, intermediate, or expert."

(Health Informatics): "Patients were <u>classified by risk level</u>: low, moderate, or high."

(**Digital Humanities**): "Historical texts were <u>categorized by genre</u>: fiction, biography, and political essays."

(Computational Biology): "Samples were grouped into clusters based on gene expression patterns under different experimental conditions."

• Novelty:

Definition: New and original contributions or discoveries.

Rationale: Novelty highlights unique findings or innovations introduced by the study, presenting **results** that demonstrate its originality and contribution to the field.

Indicator Words: "Introduces," "presents a new," "first to," "novel contribution."

Examples:

(NLP): "This study <u>introduces a new</u> algorithm for real-time translation of code-mixed languages."

(HCI): "We <u>present a novel</u> interface that adapts dynamically based on user emotions."

(Health Informatics): "Our work is the first to integrate wearable device data into predictive health models."

(**Digital Humanities**): "This research presents a <u>novel</u> computational approach to trace the evolution of literary styles over centuries."

(Computational Biology): "This study is the first to use multi-omics data integration to predict cancer patient outcomes with high accuracy."

o Comparison:

Definition: Comparing of findings or approaches.

Rationale: Comparison evaluates the study's results relative to other methods or baselines, highlighting its performance, advantages, or differences. Indicator Words: "Outperforms," "compared to," "similar to," "results

indicate better than."

Examples:

(NLP): "Our model <u>outperforms</u> GPT-3 in accuracy but requires less computational resources."

(HCI): "This design resulted in faster task completion <u>compared to</u> traditional menu-based interfaces."

(Health Informatics): "The proposed algorithm shows similar precision to expert diagnoses but operates significantly <u>faster</u>."

(**Digital Humanities**): "Our method processes historical corpora more efficiently compared to existing text-mining tools."

(Computational Biology): "This pipeline performs <u>better than</u> existing models in predicting gene-disease associations with lower false-positive rates."

• Scalability:

Definition: The ability of a system to handle larger or more complex tasks effectively.

Rationale: **Scalability** evaluates how well the **results** hold up as the system is scaled to handle larger datasets, tasks, or environments.

Indicator Words: "Scales to," "adapts seamlessly," "handles large-scale," "maintains performance at scale."

Examples:

(NLP): "The architecture <u>scales efficiently to</u> datasets with billions of sentences."

(HCI): "The interface <u>adapts seamlessly to</u> multi-device environments, including tablets and smartphones."

(Health Informatics): "The platform <u>supports integration with</u> healthcare systems managing millions of patient records."

(**Digital Humanities**): "The system processes <u>large-scale</u> digitized archives, maintaining performance across collections spanning multiple languages." (**Computational Biology**): "The algorithm <u>scales to</u> genomic datasets containing millions of sequences without significant loss of speed or accuracy."

o Causation:

Definition: Direct relationships where one thing causes another.

Rationale: Causation identifies cause-effect relationships observed in the study, showing how changes in one variable directly influence **results**. Indicator Words: "Results show that," "caused by," "led to," "directly influenced by."

Examples:

(NLP): "Increasing training data size <u>directly improves</u> the model's accuracy." (HCI): "Reducing menu complexity <u>led to</u> faster task completion times." (Health Informatics): "Timely medication reminders significantly increased

(Health Informatics): "Timely medication reminders <u>significantly increased</u> patient adherence rates."

(**Digital Humanities**): "Providing historical context for documents <u>led to</u> more accurate classification in text analysis tasks."

(Computational Biology): "Mutations in this gene <u>directly caused</u> a loss of function in the associated protein under experimental conditions."

o Correlation:

Definition: Relationships observed between variables, without proving one causes the other.

Rationale: Correlation highlights relationships or associations observed in the **results**, without establishing direct causality.

Indicator Words: "Associated with," "linked to," "correlates with," "relationship between."

Examples:

(NLP): "Word frequency <u>correlates with</u> the model's confidence in its predictions."

(HCI): "Longer session times are <u>associated with</u> higher user satisfaction ratings."

(Health Informatics): "Higher social media activity is <u>linked to</u> increased patient engagement with health interventions."

(**Digital Humanities**): "The frequency of religious motifs in texts <u>is associated</u> with the political climate of the era."

(Computational Biology): "Gene expression levels <u>are correlated with</u> changes in environmental conditions such as temperature."

• Disagreement:

Definition: Unexpected differences between observed and expected results.

Rationale: Disagreement identifies inconsistencies between predictions and actual Results.

Indicator Words: "Contrary to expectations," "unexpectedly," "differences between," "observed versus expected," "discrepancies in."

Examples:

(NLP): "The model performed worse on shorter sentences than anticipated."

(HCI): "Contrary to expectations, users preferred the non-adaptive interface in time-sensitive tasks."

(Health Informatics): "Predicted recovery rates were significantly lower than clinical observations."

(**Digital Humanities**): "<u>Unexpectedly</u>, thematic patterns in the dataset did not align with established historical narratives."

(Computational Biology): "Contrary to expectations, certain gene knockouts showed no effect on cell growth under stress conditions."

2.5 Analysis

- **Definition**: Interpretation or explanation of **Methods and Results**.
- Subarguments:
 - Explanation:

Definition: Reasons given to clarify why something happens.

Rationale: Explanation provides interpretations of the study's observations by clarifying why certain methods or phenomena occur, contributing to the overall **analysis** of the study's choices and findings.

Indicator Words: "Because," "due to," "as a result of," "this happens when," "explained by."

Examples:

(NLP): "The model struggles with rare words <u>because</u> the training data lacks sufficient examples."

(HCI): "Users prefer larger buttons <u>due to</u> improved visibility and easier interaction."

(Health Informatics): "Patients responded more quickly to shorter reminders because they require less cognitive effort."

(**Digital Humanities**): "The prevalence of certain phrases in 19th-century novels is <u>explained by</u> the influence of serialized publication formats."

(Computational Biology): "The algorithm misclassified some genes <u>due to</u> insufficient training data for rare functional categories."

Examples (as a Subargument):

Definition: Specific instances or scenarios used to illustrate findings, phenomena, or explanations.

Rationale: **Examples** help contextualize and clarify abstract findings or explanations, making them more tangible and relatable. They enhance the **analysis** by providing real-world relevance or detailed case scenarios.

Indicator Words: "For instance," "such as," "an example is," "illustrated by," "demonstrated through."

Examples:

(NLP): "An example of this is the model's performance improvement on domain-specific text, such as medical records."

(HCI): "Survey shows button size has an impact of people's interaction

satisfaction, for instance, older adults demonstrated higher interaction satisfaction when larger buttons were introduced."

(Health Informatics): "This can be seen in cases where early intervention prevented adverse health outcomes, such as reduced hospital readmissions." (Digital Humanities): "An example is the application of computational text analysis to identify recurring themes in Renaissance literature." (Computational Biology): "For instance, gene mutation patterns were

demonstrated through comparative analysis of cancer and normal tissues."

Elaborations:

Definition: Detailed expansions or additional explanations that clarify findings, methods, or theoretical insights.

Rationale: Elaborations provide deeper insights or further interpret findings and observations, enhancing understanding. They enrich the analysis by breaking down complex concepts into more digestible details or exploring implications in depth.

Indicator Words: "This means that," "in more detail," "can be expanded as," "to elaborate further".

Examples:

(NLP): "This means that increasing the training corpus size directly affects performance for low-resource languages."

(HCI): "To elaborate further, user satisfaction improves when interfaces are tailored to individual preferences."

(Health Informatics): "This can be expanded as a multi-stage process where wearable devices contribute both real-time and historical data for predictions." (Digital Humanities): "In more detail, thematic patterns in 19th-century novels reveal evolving societal norms reflected in character dialogues." (Computational Biology): "To elaborate further, integrating transcriptomic and proteomic data enables a comprehensive understanding of gene regulation networks."

2.6 Challenge

• **Definition**: Constraints or weaknesses.

Subarguments:

• Limitation:

Definition: Factors that may affect the reliability or validity of the results. Rationale: Limitation highlights constraints that influence the reliability or applicability of the study's findings, making it a critical part of the overall Challenge by identifying weaknesses that frame the study's scope. Indicator Words: "Limited by," "constrained by," "restricted to," "affected by," "scope includes."

Examples:

(NLP): "The model's performance is <u>limited by</u> the lack of diverse training

data."

(HCI): "Findings are <u>constrained by</u> the use of a controlled lab environment, which may not reflect real-world conditions."

(Health Informatics): "The analysis is <u>limited to</u> data collected from a single healthcare provider."

(**Digital Humanities**): "The study is <u>constrained by</u> the <u>limited availability</u> of digitized archives for non-Western literary traditions."

(Computational Biology): "The algorithm's effectiveness is <u>restricted to</u> datasets with high-quality sequence annotations."

O Error/Artifact:

Definition: Issues or anomalies that may affect the accuracy or interpretation of results.

Rationale: Error/Artifact addresses inaccuracies or irregularities in the data or methods, contributing to the **Challenge** by identifying potential distortions that impact result interpretation.

Indicator Words: "May include errors," "affected by artifacts," "issues in," "noise in data," "anomalies from."

Examples:

(NLP): "The model occasionally misclassifies <u>due to noise in the training</u> data."

(HCI): "User testing results <u>may include</u> artifacts from inconsistent internet speeds during sessions."

(Health Informatics): "Measurement errors in wearable devices <u>could impact</u> the accuracy of health predictions."

(**Digital Humanities**): "Text analysis outputs <u>may be affected</u> by OCR errors in digitized historical manuscripts."

(Computational Biology): "Sequencing errors in raw genomic data could introduce artifacts into mutation detection algorithms."

2.7 Ethical

• **Definition**: Ethical concerns, implications, and justifications related to the study or system.

• Subarguments:

• Ethical Risks:

Definition: Potential moral or societal risks arising from the study or system. **Rationale**: **Ethical Risks** identify specific negative ethical implications, such as bias, fairness, privacy violations, or societal harm, which need to be addressed as part of **Ethical Considerations**.

Indicator Words: "Raises concerns about," "risks of," "ethical implications," "must ensure," "could perpetuate," "potential harm."

Examples:

(NLP): "Deploying biased language models <u>risks</u> reinforcing societal stereotypes and marginalizing vulnerable groups."

(HCI): "Persuasive interface designs may <u>exploit</u> users' cognitive biases without their informed consent."

(Health Informatics): "Opaque AI-driven diagnoses can <u>undermine</u> patient trust and lead to inequitable healthcare outcomes."

(Digital Humanities): "Automated text analysis tools <u>could misrepresent</u> historical narratives, perpetuating biased interpretations of cultural texts." (Computational Biology): "The use of genomic data <u>without explicit consent</u> risks violating privacy and could disproportionately impact marginalized populations."

• Ethical Justification:

Definition: Reasons given to support ethical choices in a study or system. **Rationale**: **Ethical Justification** explains the steps taken to mitigate ethical risks and ensure the study aligns with accepted moral principles, contributing to the broader **Ethical Considerations** argument.

Indicator Words: "Ensures privacy by," "justified by," "prioritizes fairness," "adheres to ethical guidelines," "promotes inclusivity," "balances utility and ethics."

Examples:

(NLP): "Training data was <u>anonymized to protect</u> user privacy while ensuring model utility."

(HCI): "The system prioritizes accessibility features to <u>promote</u> inclusivity for users with disabilities."

(Health Informatics): "Patient <u>consent was obtained</u> before using data to ensure adherence to ethical guidelines."

(**Digital Humanities**): "The digitization process <u>adheres to</u> copyright laws and respects the cultural significance of historical artifacts."

(Computational Biology): "Genomic data usage was justified by obtaining participant consent and adhering to international bioethics standards."

• Stakeholders/Beneficiaries:

Definition: Groups or individuals affected by or benefiting from the study or system.

Rationale: **Stakeholders/Beneficiaries** identify the groups impacted by the study or system, ensuring that **Ethical Considerations** account for fairness in distributing benefits and acknowledging societal impacts.

Indicator Words: "Benefits include," "stakeholders impacted are," "beneficiaries gain from," "affects groups such as," "end-users benefit from." Examples:

(NLP): "Developers of multilingual applications <u>benefit from</u> the improved translation model."

(HCI): "End-users of e-learning platforms gain from the enhanced interface usability."

(Health Informatics): "Patients with chronic conditions <u>benefit from</u> realtime monitoring and predictive alerts."

(**Digital Humanities**): "Historians and educators <u>benefit from</u> automated analysis of large text corpora, enabling new insights into historical narratives." (**Computational Biology**): "Researchers in precision medicine <u>gain from</u> improved gene-disease association predictions, advancing personalized treatment strategies."

2.8 Implications

- **Definition**: Broader applicability or significance and future research suggestion.
- Subarguments:
 - Implication:

Definition: The broader significance or potential impact of the findings. **Rationale**: **Implication** highlights the broader applicability or real-world impact of the study's findings, demonstrating the relevance and significance of the research within the **Implications** argument.

Indicator Words: "Suggests that," "could improve," "may lead to," "potential impact includes," "significant for," "enhances."

Examples:

(NLP): "<u>Improved</u> sentiment analysis <u>can</u> enhance content moderation on social media platforms."

(HCI): "Findings <u>suggest</u> that adaptive interfaces could significantly improve productivity in workplace applications."

(Health Informatics): "Real-time health monitoring <u>may reduce</u> hospital readmission rates through early intervention."

(**Digital Humanities**): "Automated analysis of historical texts <u>could broaden</u> access to previously understudied literary works, enhancing cultural understanding."

(Computational Biology): "Better protein-structure predictions <u>may lead to</u> advancements in drug discovery and personalized medicine."

• Risk/Uncertainty:

Definition: Potential challenges or unknown factors that could affect outcomes.

Rationale: Risk/Uncertainty identifies potential limitations or unknowns regarding the applicability or success of the study's findings, ensuring that the **Implications** account for real-world challenges and variability.

Indicator Words: "May fail when," "uncertainty about," "risks include," "unknown factors," "could be affected by," "introduces challenges."

Examples:

(NLP): "The model may fail when applied to languages not represented in the

training data."

(HCI): "There is <u>uncertainty</u> about how users will adapt to the new interface in high-stress environments."

(Health Informatics): "Reliance on wearable devices <u>introduces the risk</u> of inaccurate health data due to hardware malfunctions."

(**Digital Humanities**): "The automated analysis <u>could be affected</u> by inaccuracies in OCR processing of historical texts."

(Computational Biology): "The algorithm's predictions <u>may fail</u> when applied to novel species with incomplete genomic data."

• Future Work/Questions:

Definition: Suggested directions for further research or development.

Rationale: Future Work/Questions proposes extensions, refinements, or new research directions, showing how the study's findings can inspire continued exploration under the **Implications** argument.

Indicator Words: "Future work includes," "further studies are needed," "could explore," "suggested directions include," "future research could." **Examples**:

(NLP): "<u>Future work</u> could explore adapting the model for low-resource languages."

(HCI): "<u>Further studies</u> are needed to evaluate interface performance in real-world settings."

(Health Informatics): "<u>Future research</u> could investigate the integration of genomic data into predictive health models."

(**Digital Humanities**): "<u>Future studies</u> could explore applying these methods to analyze underrepresented cultural archives."

(Computational Biology): "<u>Future research</u> could include expanding the algorithm to incorporate epigenomic data for more precise predictions."

2.9 Contradictions (to current Knowledge)

- **Definition**: Points challenging existing knowledge.
- Subarguments:

o Contradiction:

Definition: Findings that conflict with or challenge prior research.

Rationale: Contradiction highlights results that directly challenge or deviate from existing knowledge, contributing to the Contradictions to Knowledge argument by questioning established assumptions or findings.

Indicator Words: "Unlike prior studies," "contrary to previous research," "our findings challenge," "in conflict with," "disagrees with."

Examples:

(NLP): "<u>Unlike prior studies</u>, our results show that pretraining on domain-specific data is not always beneficial."

(HCI): "Contrary to previous research, we found that users preferred static

menus over adaptive ones."

(Health Informatics): "Our <u>findings contradict</u> earlier claims that wearable devices always improve patient adherence."

(**Digital Humanities**): "Our analysis challenges previous interpretations by showing that key literary themes emerged earlier than previously thought." (**Computational Biology**): "Contrary to prior research, the study found that certain gene mutations do not correlate with increased disease risk in specific populations."

• Alternative/Counterargument:

Definition: Differing or opposing perspectives on an issue.

Rationale: Alternative/Counterargument introduces opposing viewpoints or perspectives that challenge the study's claims, enhancing the Contradictions to Knowledge argument by presenting alternative explanations or frameworks.

Indicator Words: "An alternative view suggests," "some argue that," "a counter perspective is," "opposing viewpoints include," "it has been proposed that."

Examples:

(NLP): "An alternative view suggests that rule-based systems are more interpretable than neural models for certain tasks."

(HCI): "Some argue that traditional desktop interfaces remain more efficient than mobile-first designs for complex workflows."

(Health Informatics): "A counterargument is that patient-reported data might be less reliable than clinical observations."

(**Digital Humanities**): "<u>It has been proposed</u> that manual annotation of historical texts provides more nuanced interpretations than computational methods."

(Computational Biology): "A counter perspective is that single-cell data might not fully capture population-level dynamics in gene expression studies."

• Rebuttal:

Definition: Responses addressing critiques or opposing views.

Rationale: Rebuttal provides responses or justifications to counter opposing views, strengthening the Contradictions to Knowledge argument by defending the study's claims against alternative perspectives.

Indicator Words: "While it is true that," "although some argue," "despite concerns," "however," "nonetheless."

Examples:

(NLP): "While rule-based systems are interpretable, they lack the flexibility and scalability of neural models."

(HCI): "<u>Although</u> desktop interfaces may be efficient, mobile-first designs better meet the needs of on-the-go users."

(Health Informatics): "Despite concerns about reliability, patient-reported data provides valuable insights when combined with clinical records."
(Digital Humanities): "Although manual annotation provides depth, computational methods enable analysis at a scale that would be impractical otherwise."

(Computational Biology): "While population-level studies capture broad trends, single-cell data provides finer granularity essential for understanding cellular heterogeneity."

3. Notes

1. Passive tense: A + Passive verb tense + B should be annotated as Agent: A, Main Action: verb in passive tense, Base Object: B.

E.g.: "Model performance can be improved by using a better prompting strategy."

Agent: Model performance Main Action: be improved

Base Object: using a better prompting strategy

E.g. "A scoping review was conducted using the methodology of Arskey."

Agent: scoping review

Main Action: was conducted

(no object here)

Method: methodology of Arskey

- 2. Choosing a proper **Main Action**: We need to think in the authors' perspective and focus on what the author is trying to focus on. E.g., in structure like "Context + However + challenge" our **Main Action** should be from "challenge" part, because the reason for author to write the paper at all is to solve this "challenge", and it is not because the author tries to explain the "context".
- 3. Object annotation: Object will follow this structure: {"Base object": "", "Base Modifier": "", "Attached Object": "", "Attached Modifier": ""} to solve the issue.

 Specification:

Base object: The primary object without any descriptor.

Attached object: The secondary object that is also directly related to the **Main Action** or has a direct relation with the **Base Object**.

Base or Attached Modifier: The descriptor of base or attached object.

Here is an example: Understanding the full spectrum of uptake factors is essential to identify ways in which policy makers and providers can facilitate the adoption of effective digital therapeutics within a health system, as well as the steps developers can take to assist in the deployment of products.

Annotation should be:

```
Main Action: "is essential to identify"

Object: {

"Base object": "ways",

"Modifier": "in which policy makers and providers can facilitate the adoption of effective digital therapeutics within a health system",

"Attached Object": "as well as the steps",

"Modifier": "developers can take to assist in the deployment of products."}
```

Structures like "Map A into B", "Send A to B" can be also solved using this formulation of object, where A is **base object**, and into B as the **attached object**. Here Both A, and B need to be **Object**s of the **Main Action**, and it need to be a noun phrase.

Important: Attached object should include the proposition before it, to indicate the relation between it and the **Base object**.

Be careful when you annotate the Objects, it requires thinking rather than simply following structures:

E.g., "We modify the model parameter for a better performance."

Even it seemingly has a structure of "modify A for B" and both are noun phrase, but B is clearly not an Object of "modify". You can check this by reconstructing: "we modify a better performance", and it clearly contradicts to original text, and "we modify the model parameter" follows the original text.

Therefore, the only object here is a **Base Object**: "the model parameter", "for a better performance" is **Purpose**.

4. We don't have an explicit **Object** sometimes, and in that case, we leave **Object** empty: Review is conducted using ... (agent + passive verb + doing) (Agent: review, Main Action: is conducted, Object: ...) Result indicates that; (verb + that + clause)(Agent: Result, Main Action: indicates, Object: ...) We conclude by doing (verb + prop +doing) (Agent: WE, Main Action: Conclude, Object:

...

)

- 5. **Main action** should include every word if it is a Predicative Infinitive Constructions or Copular Verb Phrase. E.g., be the first to do", "have the potential to do", "be essential to do", "be a deep dive into", "be important to", (example already shown in 3 above)
- 6. Overlapping issues of **Arguments**:
- Method v.s. Context, e.g., with we obtain, using we obtain. **Method** tends to discuss more detailed design, while **Context** simply mention the name of a method.
- Challenge v.s. Result and Analysis: Challenge can sometimes also fit under Results or Analysis (examples), but we enforce a more detailed annotation when possible. E.g.,

Findings reveal the high-stakes articulation labor demanded of workers to be recognized by these systems, including maintaining multiple mobile devices, repeatedly uploading requisite images, spending time and resources visiting customer-service centers, and making physical changes to their bodies and environments.

Highlighted text should be **challenge** rather than **example** (in analysis), because challenge is **more fine-grained** definition.