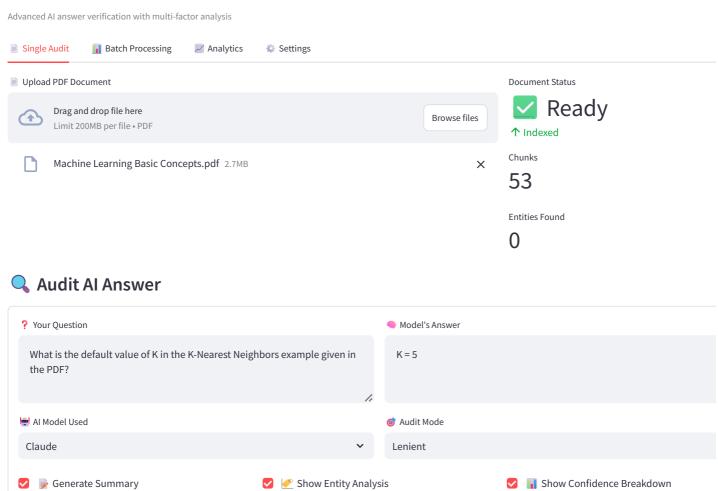
Al Knowledge Auditor Pro

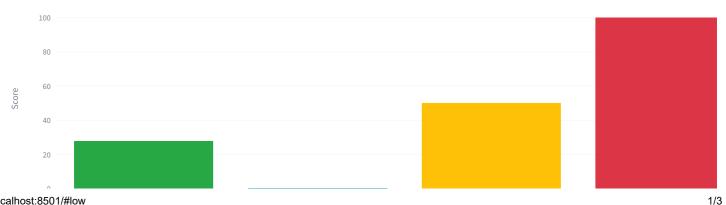


Audit Results



Audit Answer

Trust Score Breakdown



localhost:8501/#low

semantic_similarity textual_overlap entity_matching factual_consistency

Most Relevant Content

• Assumes all examples (instances) are points in the d dimen- sional space Rd. K-nearest neighbors • KNN uses the standard Euclidian distance to define nearest neighbors. Given two examples xi and xj: d(xi, xj) = v u u u t d X k=1 (xik -xjk)2

K-nearest neighbors Training algorithm: Add each training example (x, y) to the dataset D. $x \in Rd$, $y \in \{+1, -1\}$. K-nearest neighbors Training algorithm: Add each traexample (x, y) to the dataset D. $x \in Rd$, $y \in \{+1, -1\}$.



The answer appears to have little support from the document. Low semantic similarity with document content Limited textual overlap with source material Scientify alignment with document



Audit Report

Overall Assessment

- Trust Score: 34%
- Confidence Level: Low
- Similarity Score: 61.7%

Score Breakdown

- Semantic Similarity: 28.0%
- Textual Overlap: 0.2%
- Entity Matching: 50.0%
- Factual Consistency: 100.0%

Analysis

The answer appears to have little support from the document. Low semantic similarity with document content Limited textual overlap with source material Scientity alignment with document

Most Relevant Content

• Assumes all examples (instances) are points in the d dimen- sional space Rd. K-nearest neighbors • KNN uses the standard Euclidian distance to define neares neighbors.. Given two examples xi and xj: d(xi, xj) = v u u u t d X k=1 (xik -xjk)2

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localhost:8501/#low 2/3

localhost:8501/#low