



Structure of a Data Science Research Paper

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Outline of a Research Paper

- ✓ Title
- ✓ Abstract
- ✓ Keywords
- ✓ 1. Introduction
- ✓ 2. Related Work / Literature Review
- ✓ 3. Materials and Methods / Methodology
- ✓ 4. Results and Discussion
- ✓ 5. Conclusion and Future Work
- ✓ References
- ✓ Appendix (if needed)



Title

Clear, Specific, and Reflective of the Core Contribution

***(Include keywords such as “A Novel...”, “An Efficient...”, “Explainable...”, “Federated...”, etc. as appropriate)*



Abstract (200–250 words)

- ✓ 1–2 sentences: **Problem and Motivation**

- ✓ 1–2 sentences: **Methodology**

- ✓ 1–2 sentences: **Results and Key Findings**

- ✓ 1 sentence: **Conclusion and Significance**

***(Ensure technical clarity and avoid vague claims)*



Keywords (4–6)

- ✓ Include ***method-specific*** (e.g., “Federated Learning”), ***domain-specific*** (e.g., “Medical Diagnosis”), and ***innovation-specific*** (e.g., “Explainability”, “Adversarial Robustness”) terms.



1. Introduction

1.1 Background and Problem Definition

- ✓ Present the broader context of the field.
- ✓ Define the **specific problem** your research addresses.
- ✓ Clarify its technical or practical boundaries.

1.2 Motivation and Significance

- ✓ Explain **why solving this problem is important**.
- ✓ Discuss its **real-world applications** or societal/industrial impact.
- ✓ Highlight what is at stake if the problem remains unsolved.

1.3 Challenges and Limitations of Existing Approaches

- ✓ Describe the **technical or conceptual difficulties** involved in solving the problem.
- ✓ Review **existing methods/solutions** briefly.
- ✓ Explain **why they fall short** or where they lack performance or generalizability.



1. Introduction

- ✓ 1.4 **Our Contribution and Innovation**
 - ✓ Clearly state ***what your research proposes*** that is new or better.
 - ✓ Present your main contributions as bullet points.
- ✓ A table for presenting ***notations and abbreviations*** used in the paper.
- ✓ 1.5 **Paper Organization**
 - ✓ Conclude with a short paragraph describing the ***structure of the rest of the paper***.



2. Related Work / Literature Review

******(Recent and relevant to **problem domain + methods used**).

- ✓ Organized either **chronologically, thematically, or by methodology.**
- ✓ For each paper reviewed:
 - ✓ Objective
 - ✓ Methodology/Model
 - ✓ Datasets
 - ✓ Key Results
 - ✓ ***Limitations / Research Gaps***

******(Use Scopus, Web of Science, SCI/SCIE indexed papers (preferably Q1/Q2). Summarize in a comparative table if relevant).



3. Materials and Methods / Methodology

3.1 Overview of Proposed Framework

- ✓ System Diagram / Pipeline Figure
- ✓ High-level Explanation

3.2 Dataset and Data Processing

- ✓ Source, Size, and Description
- ✓ Preprocessing Steps (missing values, scaling, balancing)
- ✓ Feature Engineering or Feature Selection



3. Materials and Methods / Methodology

3.3 Proposed Model / Algorithm

- ✓ Model Architecture (with diagram if DL, transformer, etc.)
- ✓ Mathematical Formulation or Notation
- ✓ Algorithm Steps (pseudo-code or boxed algorithm)

3.4 Training Strategy

- ✓ Hyperparameter tuning (Grid, Random, BO)
- ✓ Loss Functions, Optimizers
- ✓ Early stopping, regularization, dropout
- ✓ Cross-validation method (K-fold, LOOCV, etc.)
- ✓ Computational Resources (e.g., Tesla V100, Colab, GPU specs)



4. Results and Discussion

4.1 Experimental Setup

- ✓ Hardware, Software, Environment
- ✓ Evaluation Protocol

4.2 Performance Metrics

- ✓ Accuracy, Precision, Recall, F1-score
- ✓ ROC-AUC, PR-AUC
- ✓ Cohen's Kappa, MCC, FPR, etc.
- ✓ Training/Test Time, Model Size, Memory Usage



4. Results and Discussion

4.3 Visualizations

- ✓ Confusion Matrix
- ✓ ROC & PR Curves
- ✓ Training-Validation Curves
- ✓ t-SNE / PCA Visualizations (if interpretability relevant)

4.4 Statistical Significance Testing

- ✓ e.g., t-test, Wilcoxon, ANOVA for metric comparisons



4. Results and Discussion

✓ 4.5 Comparison with Baselines / Prior Studies

- ✓ Include a comparison table with multiple models
- ✓ Highlight improvements over SOTA

✓ 4.6 Error Analysis (optional but impactful)

- ✓ Analysis of failure cases
- ✓ Possible improvements



5. Conclusion and Future Work

- ✓ Recap the goal, approach, and findings
- ✓ Highlight significance and impact
- ✓ Summarize limitations (honest + constructive)
- ✓ Mention possible extensions (e.g., scaling, deployment, real-time use)



References

- ✓ Use **Zotero** or **Mendeley** for managing
- ✓ Follow target journal/conference style (e.g., APA, IEEE, Elsevier)



Appendix (if needed)

- ✓ Additional figures, tables, or pseudocode
- ✓ Supplementary datasets / external links
- ✓ Ethics Approval Statement (for sensitive data)



Thank You