



RESEARCH METHODS MODULE



Leveraging AI for Transparency & Reproducibility

Practical and ethical applications of AI in modern research
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AI Across the Research Lifecycle

From ideation to publication — AI enhances every phase



Idea Generation

AI brainstorming
& structured lit
exploration



Literature Review

Automated
screening &
evidence
synthesis



Data Preparation

Cleaning,
anomaly
detection,
standardization



Analysis & Modeling

ML, statistical
modeling &
prediction



Reporting

Documentation
consistency &
manuscript
drafting



AI & Research Transparency

- Document all tools, parameters, and workflows used in your research
- AI outputs must remain verifiable and reproducible by independent researchers
- Clearly disclose AI assistance in all publications and reports
- Avoid 'black-box' systems — hidden algorithms obscure decision-making
- Open documentation reduces uncertainty and strengthens public trust

Reproducibility Challenges in AI Research

Randomness

AI models rely on randomized processes — fix random seeds for consistent results.

Software Versions

Different software versions may produce varying results. Always document versions.

Proprietary Systems

Closed algorithms limit access to underlying structures, hindering independent verification.

Uncontrolled Parameters

Uncontrolled parameters introduce unintended analytical bias into findings.

✓ **Solution:** Standardized documentation + controlled settings + version tracking = reproducible AI research



AI for Literature Reviews

- Rapidly screen and categorize thousands of academic publications
- Automated extraction significantly reduces manual coding errors
- NLP identifies thematic patterns across large bodies of literature
- Always independently validate AI-generated summaries for accuracy
- Transparent, documented search strategies are essential for systematic reviews



AI in Data Cleaning & Preparation

- Automatically detect missing values and inconsistent data entries
- Standardize variables across complex, multi-source datasets
- Pattern recognition accurately identifies potential outliers
- All preprocessing steps must be fully documented for auditability
- Save scripts and use version tracking to ensure reproducible cleaning

Machine Learning & Statistical Modeling

Nonlinear Patterns

ML discovers complex relationships invisible to traditional statistics

Unstructured Data

Neural networks efficiently process text, images, and large datasets

Hyperparameter Reporting

Full transparency requires documenting all model configurations

Overfitting Risks

Over-complex models fail to generalize — validate on held-out data

Interpretability

Tools like SHAP explain algorithmic decisions to stakeholders



Version Control & Reproducible Workflows

- Version control tracks every incremental change in your research code
- Platforms like GitHub support transparent, collaborative research practices
- Tagged releases preserve stable analytical environments over time
- Shared repositories enhance accountability and enable peer verification
- Proper README documentation ensures long-term reproducibility of analyses

Ethical Considerations in AI Research

● Embedded Bias

AI systems may reflect and amplify existing societal biases in training data.

● Fairness Audits

Researchers must assess model fairness across demographic groups before deployment.

● Data Privacy

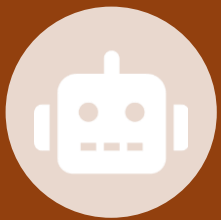
Sensitive data requires strict confidentiality protections and compliance with privacy laws.

● Human Oversight

Automated decision-making must be supervised — humans remain accountable.

● Ethics Committees

AI evaluation components should be integrated into institutional review processes.



Automation Bias & Human Responsibility

- Automation bias: the tendency to over-trust algorithmic outputs without scrutiny
- Independent verification of AI results is essential for scientific accountability
- AI should augment critical thinking and judgment — not replace it
- Researchers retain full responsibility for all published findings
- Transparency in methods strengthens accountability in AI-supported research

AI for Research Reporting

✓ BENEFITS

- Assists drafting structured manuscripts efficiently
- Automated grammar tools improve clarity and coherence
- Speeds up literature summary and methods sections

⚠️ RISKS & RULES

- Always manually verify AI-generated citations
- Hallucinated references threaten academic integrity
- Disclose AI assistance in all publications



AI Policy & Institutional Guidelines

- Universities must establish clear AI disclosure requirements for all research
- Policies should define both acceptable and prohibited AI uses explicitly
- Reproducibility statements must include full AI workflow documentation
- Training programs should promote responsible AI literacy among researchers
- Institutional leadership drives sustainable and ethical AI governance frameworks



AI Applications in Kenyan Universities

- AI can significantly strengthen systematic review capacity at local institutions
- Automated analysis reduces resource constraints for under-funded research teams
- Infrastructure limitations require strategic, phased technology investments
- Capacity building programs improve AI literacy among researchers and faculty
- Institutional collaboration accelerates responsible and sustainable AI integration

Best Practices Checklist

Responsible AI in Research

- ✓ Document AI tools, versions, and configurations thoroughly
- ✓ Fix random seeds to ensure reproducible and consistent outputs
- ✓ Archive all datasets and code repositories securely
- ✓ Disclose AI usage transparently within research manuscripts
- ✓ Conduct bias audits across demographic groups before publication

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