

On Using AI in Data Analysis

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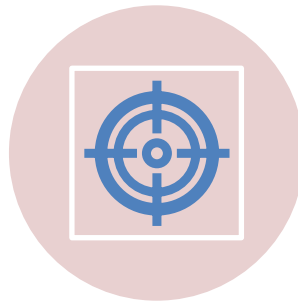
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Why This Topic Matters



DATA IS LARGER AND
MORE COMPLEX



NEED FOR FASTER,
ACCURATE ANALYSIS



AI EXPANDS WHAT
STATISTICS CAN DO

AI in data analysis refers to the application of intelligent algorithms to collect, clean, process, and interpret data



Machine Learning (ML): Enables computers to learn from data and improve their performance: regression, classification, and clustering



Deep Learning: A subset of ML that employs AI to analyze complex, high-dimensional data such as images, speech, or text



Natural Language Processing (NLP): Facilitates the analysis of textual data



Predictive Analytics: Uses AI algorithms to forecast future trends



Automated Data Preparation: AI tools can clean, normalize, and transform raw data to improve data quality

Advantages of Using AI in Data Analysis



Efficiency and Speed



Accuracy and Consistency



Automation of Repetitive Tasks



Discovery of Complex Patterns



Real-Time Analysis

AI Adoption and Usage Statistics

Metric	Statistic
Companies Using AI (at least one function)	78% (as of 2024)
AI Users Globally (est. 2025)	378 million
Professionals Using AI Daily	74% of global AI users
Data Scientists Using AI Daily	45.1%
AI Adoption as Top Priority for Companies	83%
Adults Using Generative AI Daily	Approx. 1/3

Common Statistical Techniques Requested and Used with AI



**Descriptive
Statistics**



**Hypothesis
Testing**



**Regression
Analysis**



Clustering



**Predictive
Analytics**



**Time Series
Analysis**



**Natural Language
Processing (NLP)**



**Dimensionality
Reduction (PCA)**

Challenges



OVERFITTING



BIAS



INTERPRETABILITY
ISSUES



ETHICAL & DATA
QUALITY CONCERNS

AI platforms vs. Statistical software

Aspect	Traditional Statistical Software	AI Platforms
Data Assumptions	Requires explicit assumptions about data distribution (e.g., linearity), which can be rigorously tested.	Automatically learns complex patterns without strong prior assumptions about the data structure (non-linear relationships).
Data Quality/Bias	More resilient to minor data issues ; bias introduced is often easier to trace and correct manually.	Highly sensitive to data quality and inherent biases in the training data (e.g., a hiring algorithm trained on male-dominated data excluded female candidates).
Generalizability	Models can sometimes be less effective when applied to data outside their specific context, especially with limited initial data.	Excels at generalization and making accurate, repeatable predictions on unseen data if properly trained and validated (e.g., using cross-validation).

AI platforms vs. Statistical software

Feature	Traditional Statistical Software	AI Platforms
Accuracy (Predictive Tasks)	Good for low-dimensional data	Higher, especially for complex/large data
Speed (Processing Large Data)	Slower, often manual setup required	Faster (up to 5x for analytics tasks)
Scalability	Less scalable for "big data"	Highly scalable, designed for massive datasets
Interpretability	High (transparent models)	Low to moderate ("black box" models often require extra tools like SHAP for explanation)

• Task	AI Platforms	Common Error
• Descriptive statistics (means, SD, counts, proportions)	• High concordance — ChatGPT-4 produced results matching SPSS/R in most cases reported.	• Rare rounding / formatting differences; sometimes omitted exact sample-weight handling.
• Independent / paired t-tests	• Very high agreement (identical test statistics / p-values reported in multiple studies).	• Minimal differences (rounding).
• Simple linear regression	• High concordance for coefficients, SEs, and p-values in studies.	• Occasionally, differences in default contrasts or handling of categorical variables (reference levels).
• One-way ANOVA (F statistic)	• F-values generally consistent between ChatGPT-4 and SPSS/R.	• Post-hoc (pairwise mean diffs, CIs, adjusted p-values) showed discrepancies in several reports.
• Repeated-measures ANOVA	• F statistics sometimes matched, but studies reported incorrect degrees-of-freedom or mis-handled within-subject structure.	• Wrong df, incorrect sphericity corrections or missing Greenhouse-Geisser adjustments.
• MANOVA / Multivariate tests	• Less reliable; studies report inconsistent or incorrect outputs for MANOVA.	• Incorrect dfs, missing multivariate test statistics / misreported test assumptions.

• Task	AI Platforms	Common Error
• Non-parametric tests (Mann-Whitney, Wilcoxon, Kruskal-Wallis)	Mixed results — some metrics matched, but differences reported in U/Z values, IQRs, or p-values.	Differences in ranking/tie handling and test variants (continuity correction).
• Multivariable regression (GLMs, logistic regression)	Good agreement on coefficients and p-values for many examples; careful on variable coding and reference categories.	Differences arise from link function defaults, factor coding, or omitted interactions.
• Complex models (multilevel/hierarchical models, SEM, survival analysis)	Not well-evaluated / not robust in current literature — studies warn ChatGPT-4 struggles with these.	Often lacks detailed diagnostics, misapplies model assumptions, or omits necessary steps.
• Code generation (R / Python syntax)	Strong: ChatGPT-4 can generate runnable code for many common analyses; accelerates workflow.	Code may need minor fixes; default package choices or plotting options may differ from user's standards.
• Interpretation / narrative explanation	Excellent: clear plain-language explanations of results and guidance.	Explanations can be overconfident; may omit caveats or assumption checks unless prompted.

Strengths

Traditional
Statistics:

- Strong inference

- Clear interpretation

AI Platform

- High predictive accuracy

- Handles complex, large-scale data

Weaknesses

Traditional
Statistics:

Limited with
nonlinear/high-
dimensional data

AI Platform

- Risk of
overfitting

- Less
interpretable

When to Use Each

Traditional
Statistics:

Explanation, small
samples

AI Platform

- Prediction, large
or unstructured
data

Closing Message



AI —NOT REPLACES—THE
STATISTICIAN.



FUTURE = STATISTICS +
INTELLIGENT ALGORITHMS.

Selected References

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Thank You

Questions?

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