

Improving Drug Name Normalization in FAERS Using RxNorm and Approximate Matching

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Introduction

- FAERS contains millions of drug names ¹
- Drug names are inconsistent (brand names, misspellings, formatting) ²
- Standardization is required for pharmacovigilance ³
- RxNorm provides a standardized drug vocabulary ⁴

Example Drug Name Normalization

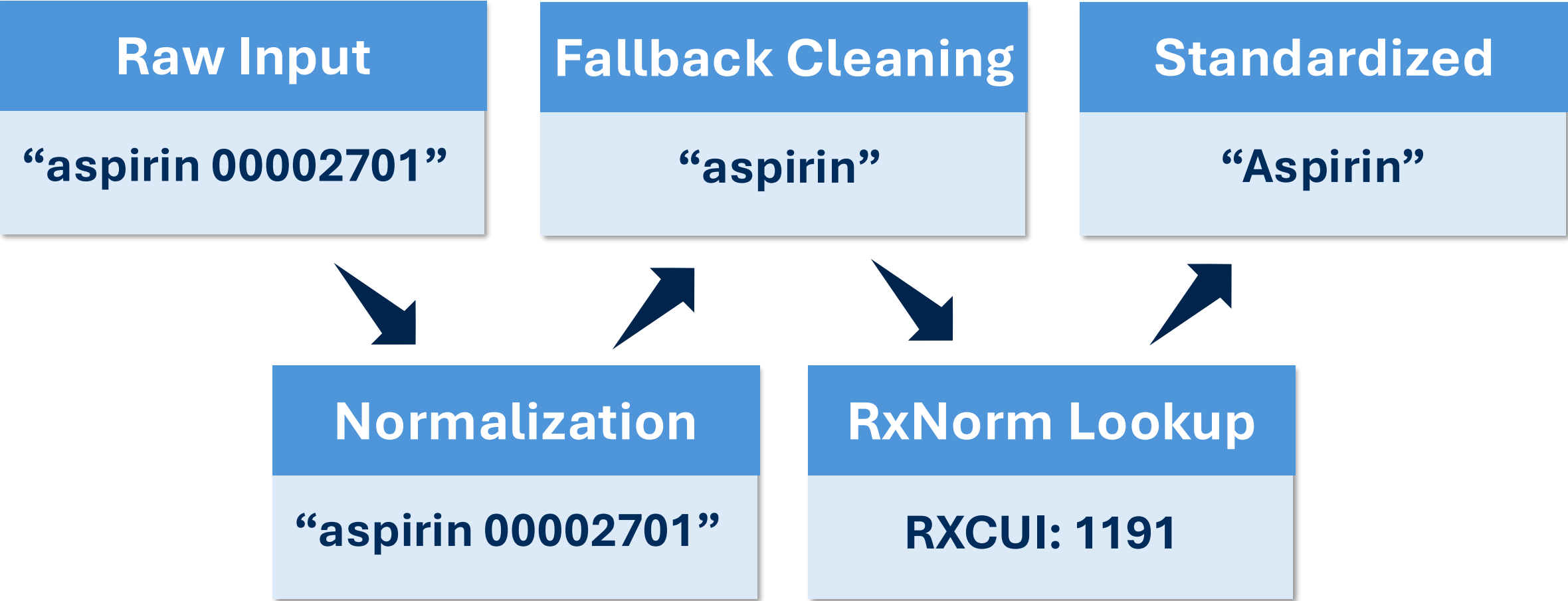


Figure 1. This example illustrates removal of numeric codes. Other common failure types included brand names, concatenated combination products, and formulation/device-specific suffixes.

Data

- Source: FDA Adverse Event Reporting System (FAERS), quarterly reports (2012–2023)
 - Total drug records: 38.9 million
 - Unique normalized drug names: 414,909
- To improve efficiency and impact, analysis focused on the top 5,000 most frequent drug names:
- Covers **35.5 million records (91.3% of total data)**
 - Captures majority of high-impact adverse event reports

Methods

- Deterministic preprocessing: Rule-based normalization (lowercasing, removal of punctuation and numeric artifacts, and whitespace normalization)
- RxNorm mapping (exact + approximateTerm): Staged lookup strategy using the RxNorm API, first attempting exact matches, followed by fallback cleaning and approximate matching (threshold ≥ 8.5)

Drug Name Normalization Pipeline

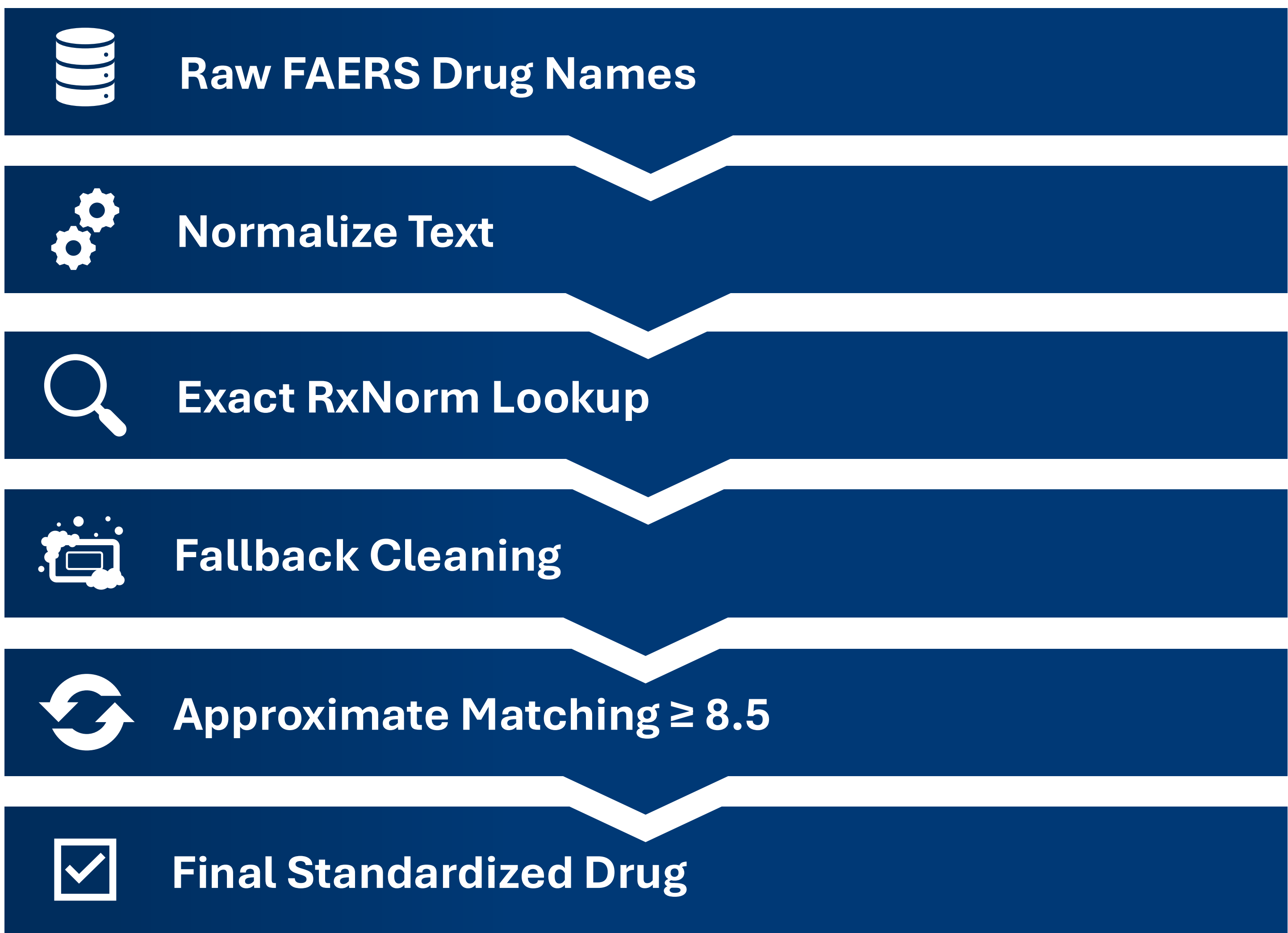


Figure 2. Deterministic pipeline for standardizing FAERS drug names using exact matching, rule-based cleaning, and approximate matching. This staged approach prioritizes precision while progressively recovering difficult-to-match drug names.

Conclusion

- Drug name normalization in FAERS requires more than exact matching due to widespread inconsistencies in real-world data.
- By combining deterministic preprocessing with approximate RxNorm matching, total row-level coverage increased from **85.16%** to **95.64%**.
- These results show that a staged, precision-first pipeline can effectively standardize large-scale pharmacovigilance data while preserving accuracy.
- Improved normalization enables more reliable downstream analysis of adverse drug events.

Future Work

- Ingredient-level normalization**
 - Map drug products to active ingredients for more consistent aggregation of adverse events
- Improved handling of complex drug names**
 - Address concatenated combination products and multi-ingredient formulations
- Expansion beyond top 5,000 drugs**
 - Extend normalization to lower-frequency drug names
- Integration into downstream analysis**
 - Apply standardized drug names to adverse event signal detection and modeling

References

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- Ogwo, M., Dhake, P., Schumaker, R., Dixit, R., & Veronin, M. (2020). A systematic approach to cleaning of drug name records data in the FAERS database: A case report. *International Journal of Big Data Management*, 1(2), 105–118. <https://doi.org/10.1504/IJBDM.2020.10034546>
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Results

Row Coverage by Method

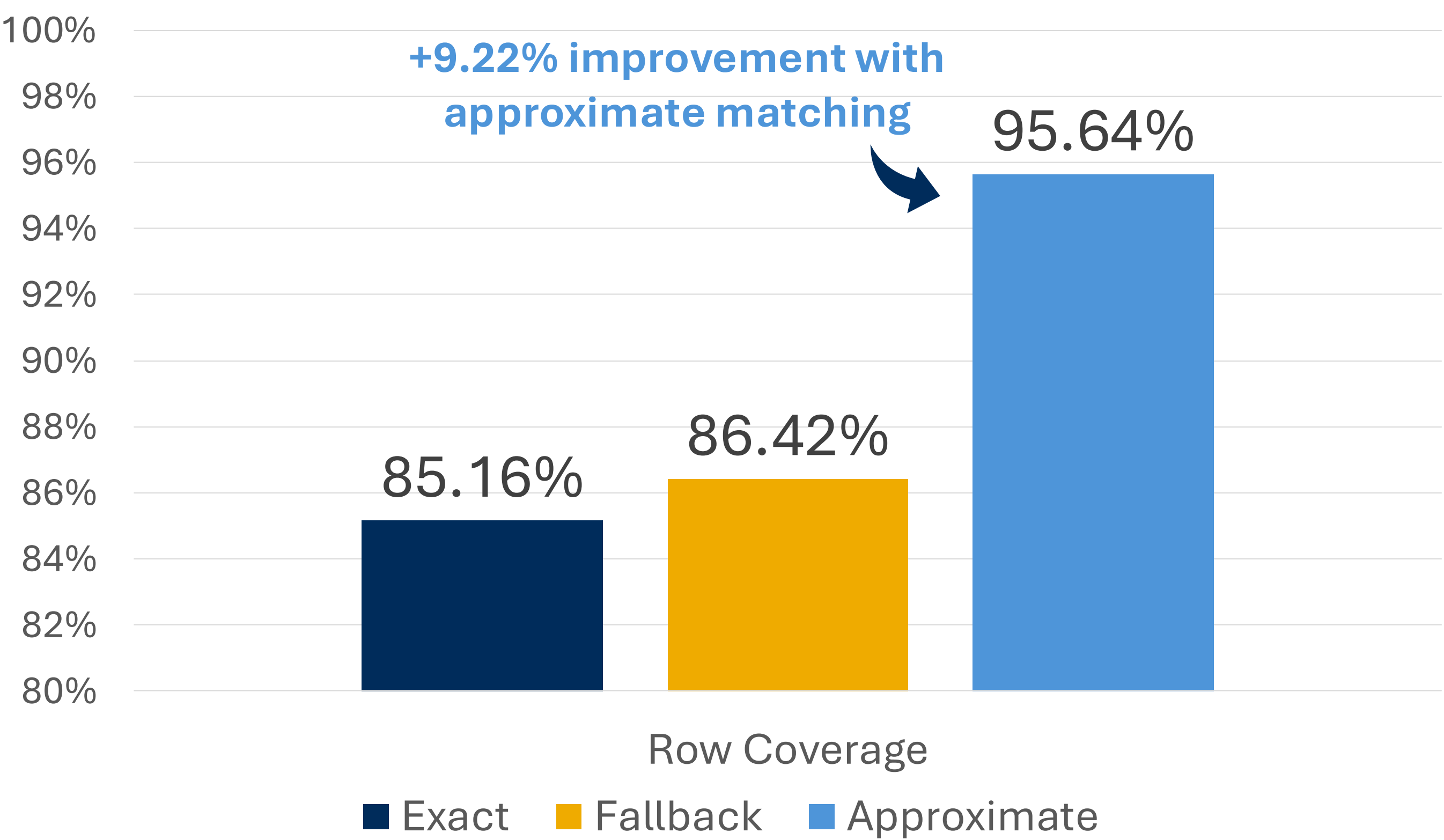


Figure 3. Comparison of row-level drug name coverage across exact, fallback, and approximate mapping methods. Approximate matching increases total coverage to 95.64%, demonstrating that many high-frequency drug names cannot be resolved through exact matching alone.

Method	% Names Mapped	Row Coverage
Exact RxNorm	56.98%	85.16%
+ Fallback	58.62%	86.42%
+ Approximate (≥ 8.5)	84.20%	95.64%

Table 1. Comparison of mapping performance across methods. While exact matching provides strong baseline coverage, fallback and approximate strategies are necessary to capture additional variation in real-world drug naming.