

Braille Auto-Correct

This document details a streamlined approach to auto-correct Braille input using only the Levenshtein distance for matching, along with optimizations and trade-offs.

1. Objective

Provide a simple yet effective real-time suggestion system for QWERTY-based Braille input that:

- Corrects substitutions, insertions, and deletions.
 - Requires minimal setup and dependencies.
 - Delivers low latency for small-to-medium dictionaries.
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2. Approach

1. Input Normalization
 - Map keys s, d, f, j, k, l \rightarrow dots 1–6.
 - Sort and join into a string pattern ("sdf" \rightarrow "123").
 2. Dictionary Preprocessing
 - Precompute pattern for each word in the dictionary once at startup.
 3. Brute-Force Matching
 - For each query, iterate all dictionary patterns:
 - Compute Levenshtein distance with early exit.
 - Collect words whose distance \leq max_dist.
 4. Ranking & Suggestion
 - Sort by (distance, word) to resolve ties.
 - Return the top-K candidates (e.g., 3).
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3. Core Algorithm Components

3.1. QWERTY-to-Pattern Conversion

- Mapping Dict: {'s':1,'d':2,'f':3,'j':4,'k':5,'l':6}
- Function: Filter valid keys, map, sort, join.

3.2. Levenshtein Distance

- DP Table: Compute in $O(n \cdot m)$ time, where n, m are string lengths.
- Space: $O(\min(n, m))$ by storing only the previous row.
- Early Exit: Abort when the current row's minimal edit cost exceeds max_dist.

4. Optimizations

- Early-Abandon: Greatly reduces work on dissimilar patterns by cutting DP loops early.
- One-Time Preprocessing: Patterns computed just once, reducing per-query overhead.
- Compact Patterns: Using short digit strings speeds up string comparisons and distance calls.
- Threshold Tuning: Adjusting `max_dist` to balance recall vs. computation (e.g., 1–2 for typical Braille patterns).

5. Trade-Offs Analysis

Choice	Benefit	Drawback
Brute-Force Search	Simplicity; no additional data structures	$O(N \cdot n \cdot m)$ per query may be slow for large N
Levenshtein	Handles all typo types uniformly	Quadratic with pattern length
Pure Python	Portable; no external deps	Single-threaded performance ceiling
Early-Termination	Cuts off costly comparisons early	Requires careful threshold setting

6. Suitability

- When to Use: Prototyping, educational demos, dictionaries up to ~50K entries.
 - Scalability: For larger datasets or microsecond latency, layer in an index (e.g., BK-tree) or compile hotspots in C/C++.
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