K-Medoids for K-Means Seeding

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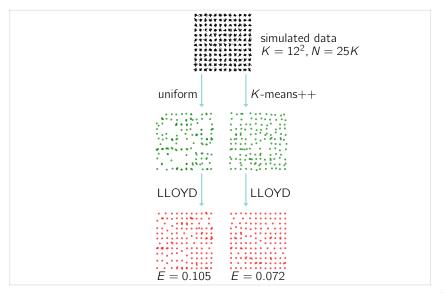




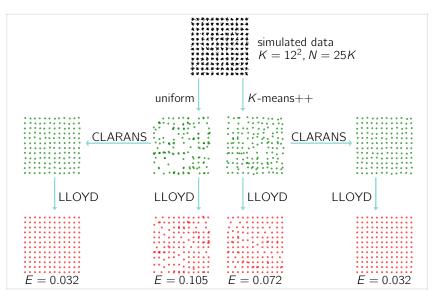


The standard K-means pipeline

First: Seeding. Second: Lloyd's (a.k.a. K-means) algorithm.



The standard K-means pipeline (+CLARANS)



CLARANS of Ng and Han (1994)

- 1: while not converged do
- 2: randomly choose 1 center and 1 non-center
- 3: **if** swapping them decreases *E* **then**
- 4: implement the swap
- 5: end if
- 6: end while

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- long-range swaps
- updating centers and samples *simultanously*.

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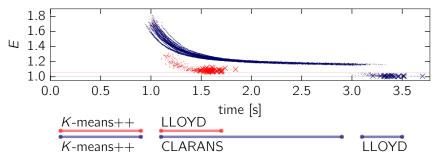
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We present algorithmic improvements, where

- computing new E is O(N/K)
- implementing swap is O(N).

Results

- RNA dataset, d = 8, $N = 16 \times 10^4$, K = 400
- 50 runs without CLARANS (red), 24 runs with (blue).



• On 16 datasets, geometric mean improvement is 3%.

CLARANS with Levenshtein metric for sequence data, $l_0, l_1, \ldots, l_\infty$ for sparse/dense vectors, many others, on github.



The end

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