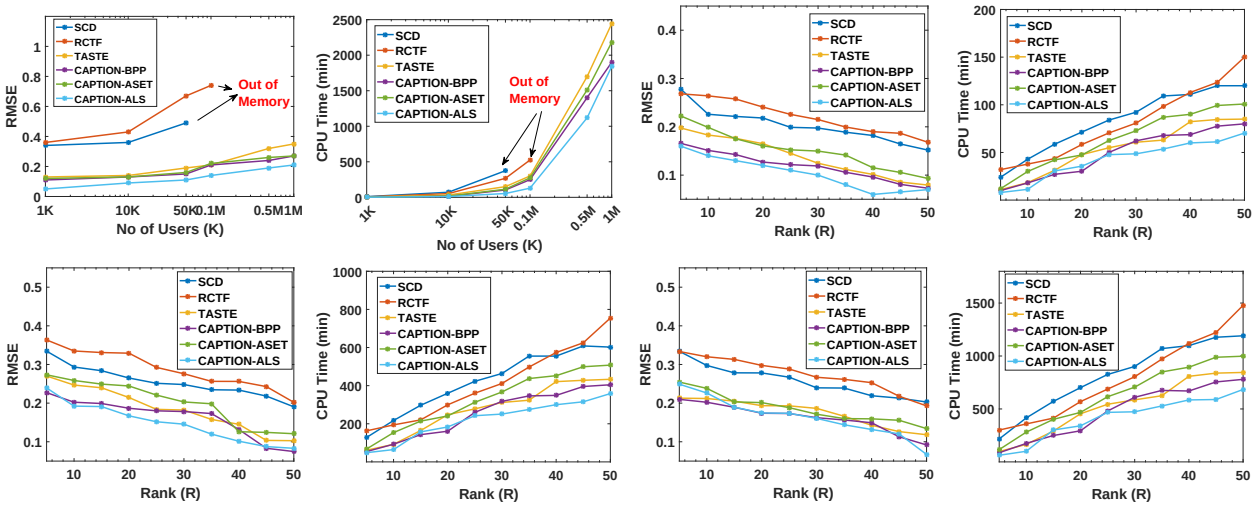

Figure. 5: Scalability analysis of C^3APTION method using synthetic and three real world datasets. (a-b): Scalability analysis of synthetic data with respect to varying number of users K, where K ranges $10^3 - 10^6$. Stable performance (RMSE) in the range $1K - 50K$, for most methods. Baseline method SCD and RCTF runs out of memory. (c-d): Scalability analysis with respect to Movielens dataset (c-d), Adobe dataset (e-f) and CMS heath record data (g-h). C^3APTION-ALS significantly outperforms the other methods even when data is very sparse.

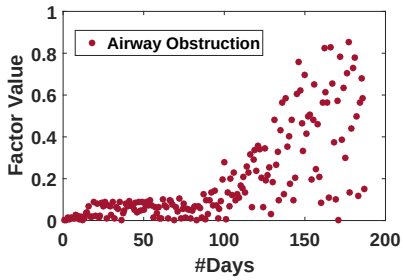

Figure. 6: Visualization of time-frame captured of the patient no. 11426 created by C^3APTION-ALS on CMS dataset.

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