

Practical Correlated Topic Modeling and Analysis via the Rectified Anchor Word Algorithm (Supplementary Material)

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A Implementation

For reproducibility, we provide pseudocode for algorithms integrated in different parts of JSMF.

A.1 Rectification

Algorithm 1 Alternating Projection

def RECTIFY-C-AP(C, K)

- 1: $C_{NN} \leftarrow C$
 - 2: **repeat**
 - 3: $(U, \Lambda_K) = \text{TRUNCATED-EIG}(C_{NN}, K)$
 - 4: $\Lambda_K^+ \leftarrow \text{diag}(\max\{\text{diag}(\Lambda_K), 0\})$
 - 5: $C_{PSD} \leftarrow U \Lambda_K^+ U^T$
 - 6: $C_{NOR} \leftarrow C_{PSD} + \frac{1 - \sum_{i,j} C_{PSD}(i,j)}{N^2} \mathbf{1}\mathbf{1}^T$
 - 7: $C_{NN} \leftarrow \max\{C_{NOR}, 0\}$
 - 8: **until** the convergence of C_{NN}
 - 9: **return** $C \leftarrow C_{NN} / (\sum_{i,j} C_{NN}(i,j))$
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$\text{diag}(\cdot)$ is the Matlab-style operation that maps the input vector into the diagonal matrix or extracts the diagonal vector from the input matrix.

Algorithm 2 Cyclic Douglas-Rachford Iteration

def RECTIFY-C-DR(C, K)

- 1: $C_3 \leftarrow C$
 - 2: **repeat**
 - 3: $C_1 \leftarrow \frac{I + R_{NOR} R_{PSD}}{2} C_3$
 - 4: $C_2 \leftarrow \frac{I + R_{NN} R_{NOR}}{2} C_1$
 - 5: $C_3 \leftarrow \frac{I + R_{PSD} R_{NN}}{2} C_2$
 - 6: **until** the convergence of C_3
 - 7: **return** $C \leftarrow C_3 / (\sum_{i,j} C_3(i,j))$
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I is the identity mapping, $R_{NOR} R_{PSD}$ denotes the composition of two reflection operators: the matrix is first reflected onto the PSD space, then onto the NOR space. $R_{PSD} C = 2C_{PSD} - C$, and similarly for NOR and NN .

A.2 Topic Inference

Algorithm 3 Sparse Implicit Column-pivoted QR

def FIND-S(\bar{C}, K)

- 1: $(P, Q, S, r) \leftarrow (\bar{C}^T, \mathbf{0}^{N \times K}, \emptyset, \mathbf{0}^K)$
 - 2: $u \leftarrow (\|p_1\|_2^2, \dots, \|p_N\|_2^2) \in \mathbb{R}^{1 \times N}$
 - 3: **for** $k = 1$ to K **do**
 - 4: $n \leftarrow \arg\max_{1 \leq i \leq N} u_i$
 - 5: $(S, q_k, r_k) \leftarrow (S \cup \{n\}, p_n, \sqrt{u_n})$
 - 6: $q_k \leftarrow (q_k - \sum_{l=1}^{k-1} \langle q_l, p_n \rangle q_l) / r_k$
 - 7: $u \leftarrow u - (q_k^T P) \circ (q_k^T P)$
 - 8: **end for**
 - 9: **return** (S, r)
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$\circ : \mathbb{R}^N \times \mathbb{R}^N \rightarrow \mathbb{R}^N$ denotes entry-wise multiplication of two vectors.

Algorithm 4 ADMM

def RECOVER-B($\bar{C}, c, S, \lambda, \gamma$)

- 1: $(U, \check{B}, B) \leftarrow ((\bar{C}_{S^*})^T, \mathbf{0}^{K \times N}, \mathbf{0}^{N \times K})$
 - 2: $\check{B}_{*S} \leftarrow I_K$ ($I_K = K \times K$ identity matrix)
 - 3: $F \leftarrow (\gamma U^T U + I_K)^{-1}$
 - 4: **for each** $i \in \{1, \dots, N\} \setminus S$ (in parallel) **do**
 - 5: $(v, f) \leftarrow ((\bar{C}_{i^*})^T, \gamma U^T v)$
 - 6: $y^{(0)} \leftarrow \Pi_{\Delta^{K-1}}((U^T U)^{-1}(f/\gamma))$
 - 7: $q^{(0)} \leftarrow y^{(0)}$
 - 8: **repeat**
 - 9: $p^{(t)} \leftarrow F(2y^{(t-1)} - q^{(t-1)} + f)$
 - 10: $q^{(t)} \leftarrow q^{(t-1)} + \lambda(p^{(t)} - y^{(t-1)})$
 - 11: $y^{(t)} \leftarrow \Pi_{\Delta^{K-1}}(q^{(t)})$
 - 12: **until** the convergence of $y^{(t)}$
 - 13: $\check{B}_{*i} \leftarrow y^{(t)}$
 - 14: **end for**
 - 15: **for** $(i, k) \in \{1, \dots, N\} \times \{1, \dots, K\}$ **do**
 - 16: $B_{ik} \leftarrow (\check{B}_{ki} c_i) / (\sum_{i'=1}^N \check{B}_{ki'} c_{i'})$
 - 17: **end for**
 - 18: **return** B
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$\Pi_{\Delta^{K-1}}(\cdot)$ is the orthogonal projection to the $K - 1$ simplex.

Algorithm 5 Diagonal Recovery

def RECOVER-A(C, B, S)

- 1: $(C_{SS}, D) \leftarrow (C(S, S), B(S, *))$
 - 2: $A \leftarrow D^{-1}C_{SS}D^{-1}$
 - 3: **return** A
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Set indexing extracts a principle submatrix whose rows and columns correspond to the arguments. Since $B(S, *)$ is diagonal, we use the element-wise reciprocal of D as D^{-1} .
