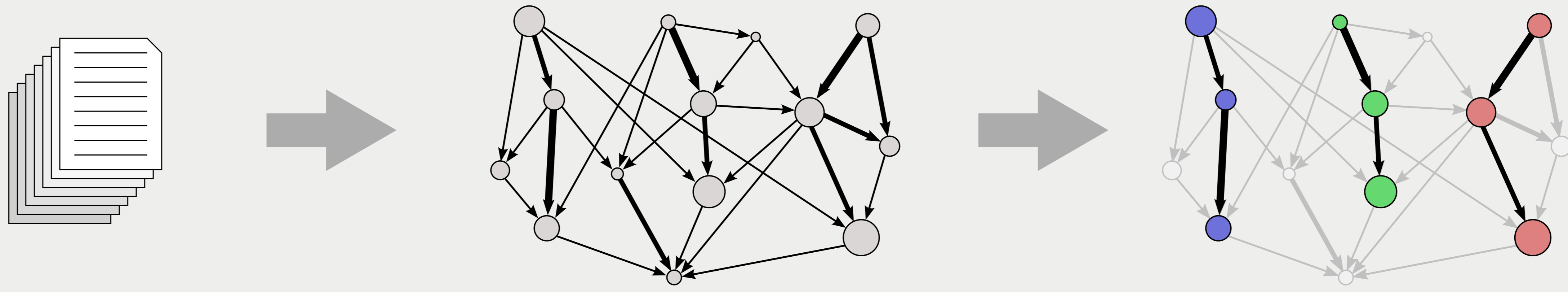


Document Collection Threading – (1) Build a graph from the collection, using measures of importance and relatedness to weight nodes (documents) and build edges (relationships). (2) From this graph, extract a diverse, salient set of threads to represent the collection.



Introduction

► Motivation: current search tools are insufficient

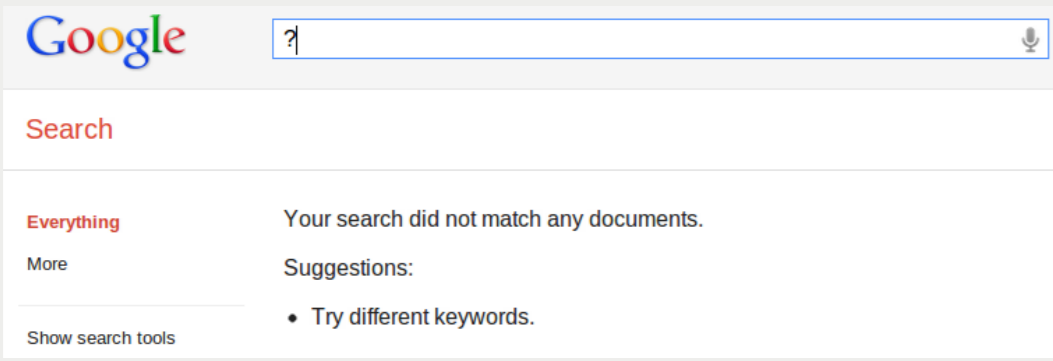


Figure: Prior knowledge of document contents is required to construct a query



Figure: Structure indicating relationships among returned documents is missing

► Related threading work

- Selecting a *single* thread (D. Shahaf and C. Guestrin, KDD 2010)
- Constructing diverse *topic* threads (A. Ahmed and E. Xing, UAI 2010)

Approach: Determinantal Point Processes

► Decompose quality and similarity of a thread $\mathbf{y}_i = (y_{i1}, \dots, y_{iT})$

$$\mathbf{q}(\mathbf{y}_i) = \mathbf{q}(y_{i1}) \prod_{t=2}^T \mathbf{q}(y_{it} | \mathbf{y}_{i(t-1)}) \quad \phi(\mathbf{y}_i) = \sum_{t=1}^T \phi(y_{it})$$

► Score a set of threads \mathbf{Y} via structured determinantal point process (SDPP)

(A. Kulesza and B. Taskar, NIPS 2010)

► SDPP: defines a distribution over sets \mathbf{Y}

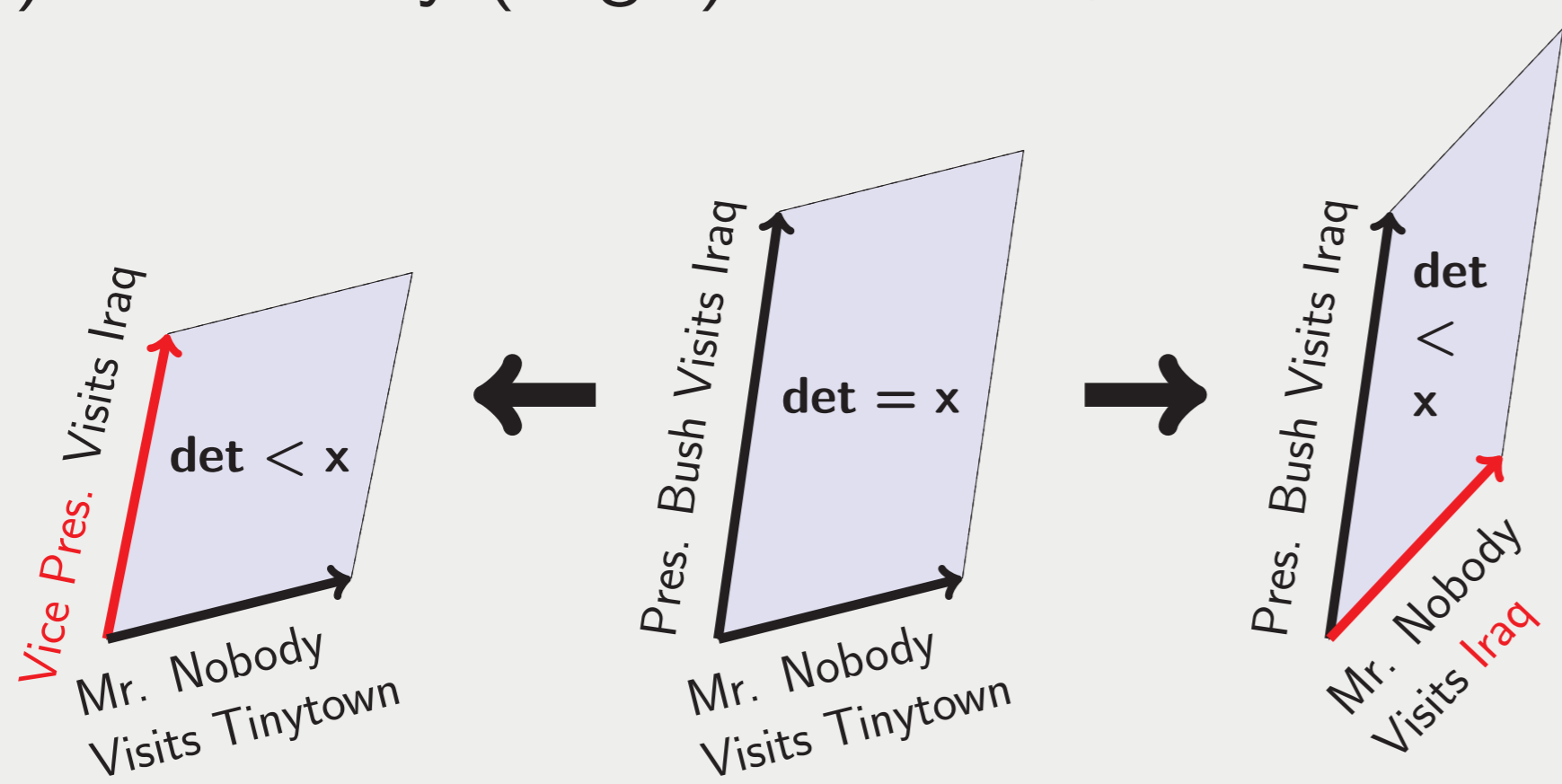
$$\mathbf{L}_{ij} = \mathbf{q}(\mathbf{y}_i) \phi(\mathbf{y}_i)^T \phi(\mathbf{y}_j) \mathbf{q}(\mathbf{y}_j)$$

$$\mathcal{P}(\mathbf{Y}) = \frac{\det(\mathbf{L}_{\mathbf{Y}})}{\sum_{\mathbf{Y}' \subseteq \{1, \dots, n\}} \det(\mathbf{L}_{\mathbf{Y}'})} = \frac{\det(\mathbf{L}_{\mathbf{Y}})}{\det(\mathbf{L} + \mathbf{I})}$$

$$\mathbf{Y} = \{i\} \rightarrow \mathcal{P}(\mathbf{Y}) \propto \mathbf{q}(\mathbf{y}_i)^2$$

$$\mathbf{Y} = \{i, j\} \rightarrow \mathcal{P}(\mathbf{Y}) \propto \mathbf{q}(\mathbf{y}_i)^2 \mathbf{q}(\mathbf{y}_j)^2 (1 - (\phi(\mathbf{y}_i)^T \phi(\mathbf{y}_j))^2)$$

► $\det(\mathbf{L}_{\mathbf{Y}})$ is proportional to volume spanned by the vectors $\mathbf{q}(\mathbf{y}_i) \phi(\mathbf{y}_i)$. As quality (length) or diversity (angle) decreases, volume decreases.



► **k**-SDPPs: fix # of points in \mathbf{Y} to \mathbf{k} (A. Kulesza and B. Taskar, ICML 2011)

► Sampling from **k**-SDPPs can be done in $\mathcal{O}(\text{TrnD}^2 + \mathbf{D}^3)$, where $\mathbf{r} = \max$ node degree, $\mathbf{n} = \#$ of nodes, $\mathbf{D} = \#$ of features

Random Projections for Tractability

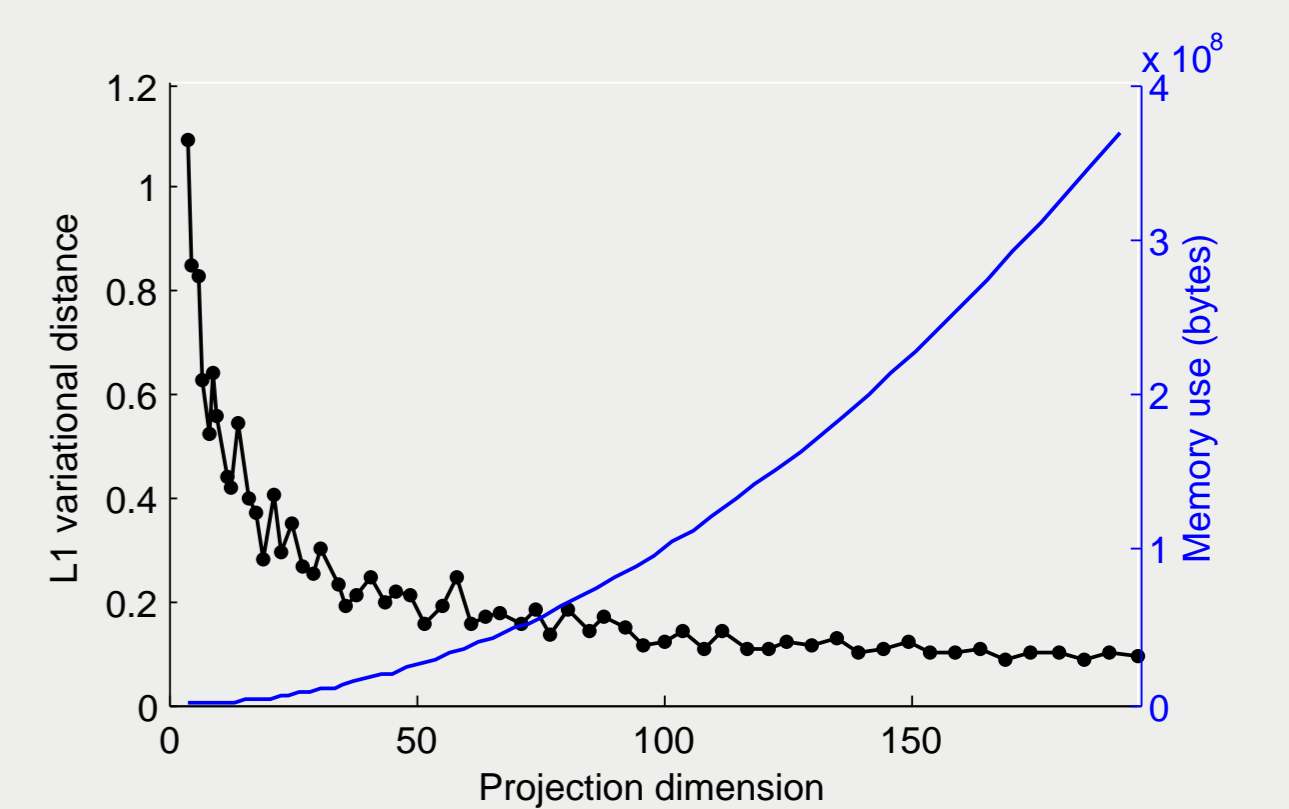
► Complexity \mathbf{D}^3 can be prohibitively large, so we project \mathbf{D} down to \mathbf{d}

► **Theorem:** Given $\tilde{\mathcal{P}}^{\mathbf{k}}(\mathbf{Y}) =$ distribution after projecting \mathbf{D} to $\mathbf{d} = \mathcal{O}(\max\{\mathbf{k}/\epsilon, (\log(1/\delta) + \log \mathbf{N})/\epsilon^2\})$, error is bounded by:

$$\|\mathcal{P}^{\mathbf{k}} - \tilde{\mathcal{P}}^{\mathbf{k}}\|_1 \leq e^{6\mathbf{k}\epsilon} - 1 \approx 6\mathbf{k}\epsilon$$

with probability at least $1 - \delta$.

► Random projections on a small threading task where the exact model is tractable: $\mathbf{n} = 600$ and $\mathbf{D} = 150$. As predicted by the theorem, fidelity to the true model increases rapidly with \mathbf{d} .



New York Times Timelines

► **Data** – six 6-month NYT article sets; **Graph** – edges are tfidf cosine scores

► **Baselines** – **k**-means clustering on time slices, dynamic topic model (DTM) (D. Blei and J. Lafferty, ICML 2006)

	ROUGE-SU4	Coherence	Interlopers	Secs
k -means	3.76	2.73	0.71	625
DTM	3.44	3.19	1.10	19,443
k -SDPP	3.98	3.31	1.15	252

Table: **ROUGE-SU4**: comparison to human summaries. **Mechanical Turk**: thread coherence rating (1-5); average # of random interloper articles identified. **Secs**: runtime.

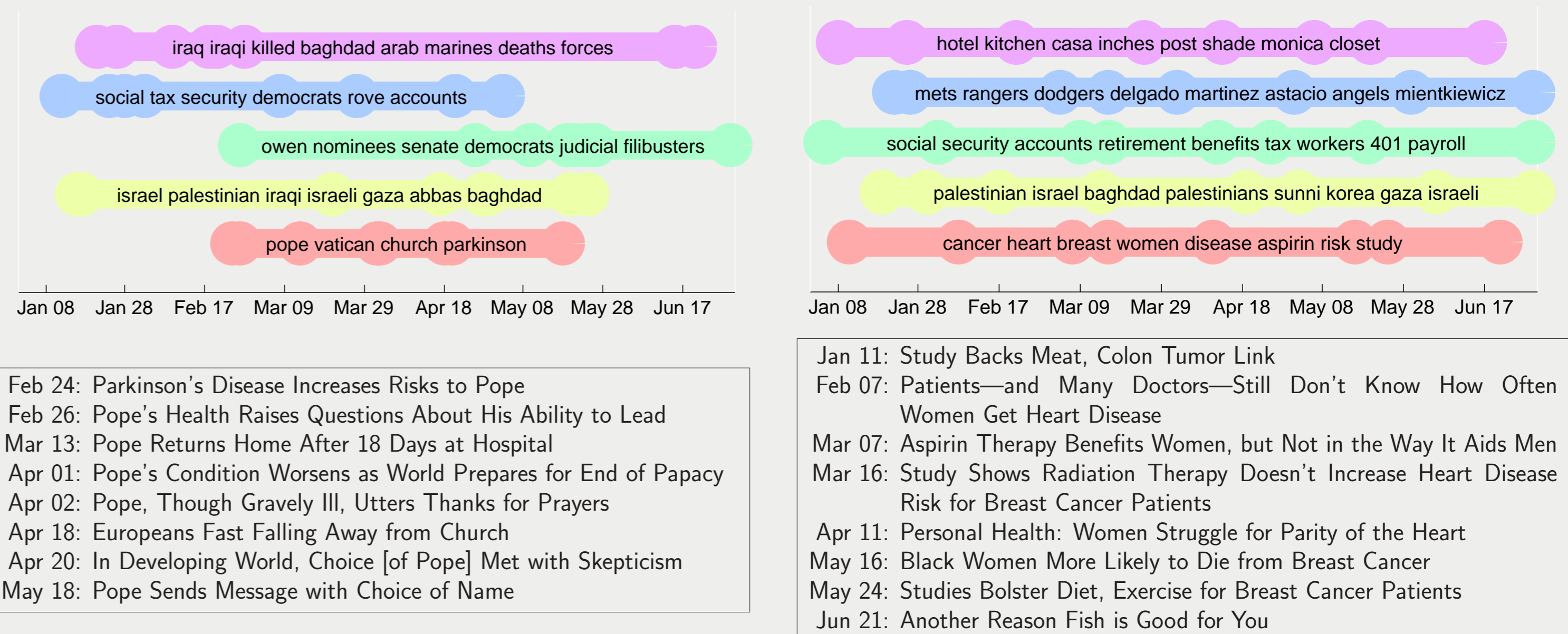


Figure: A set of threads from a **k**-SDPP (left) and a DTM (right). Above, threads are shown with the most salient words superimposed; below, headlines from the last thread are listed.

Example New York Times Graph

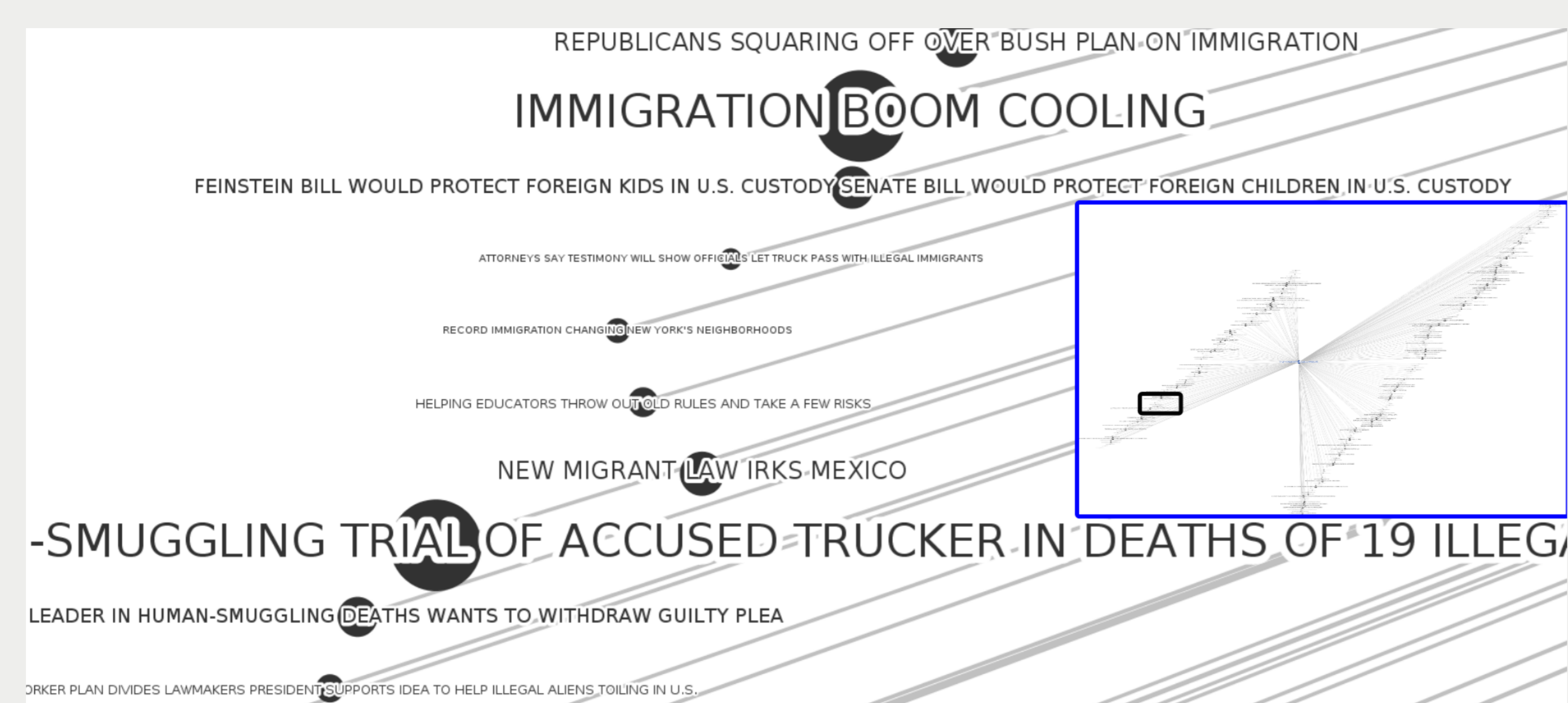


Figure: **Blue box**: Single node "Study Analyzes Data on Illegal Immigrants" with all its neighbors. **Other**: Zoom in, indicated by a black rectangle in the full image.

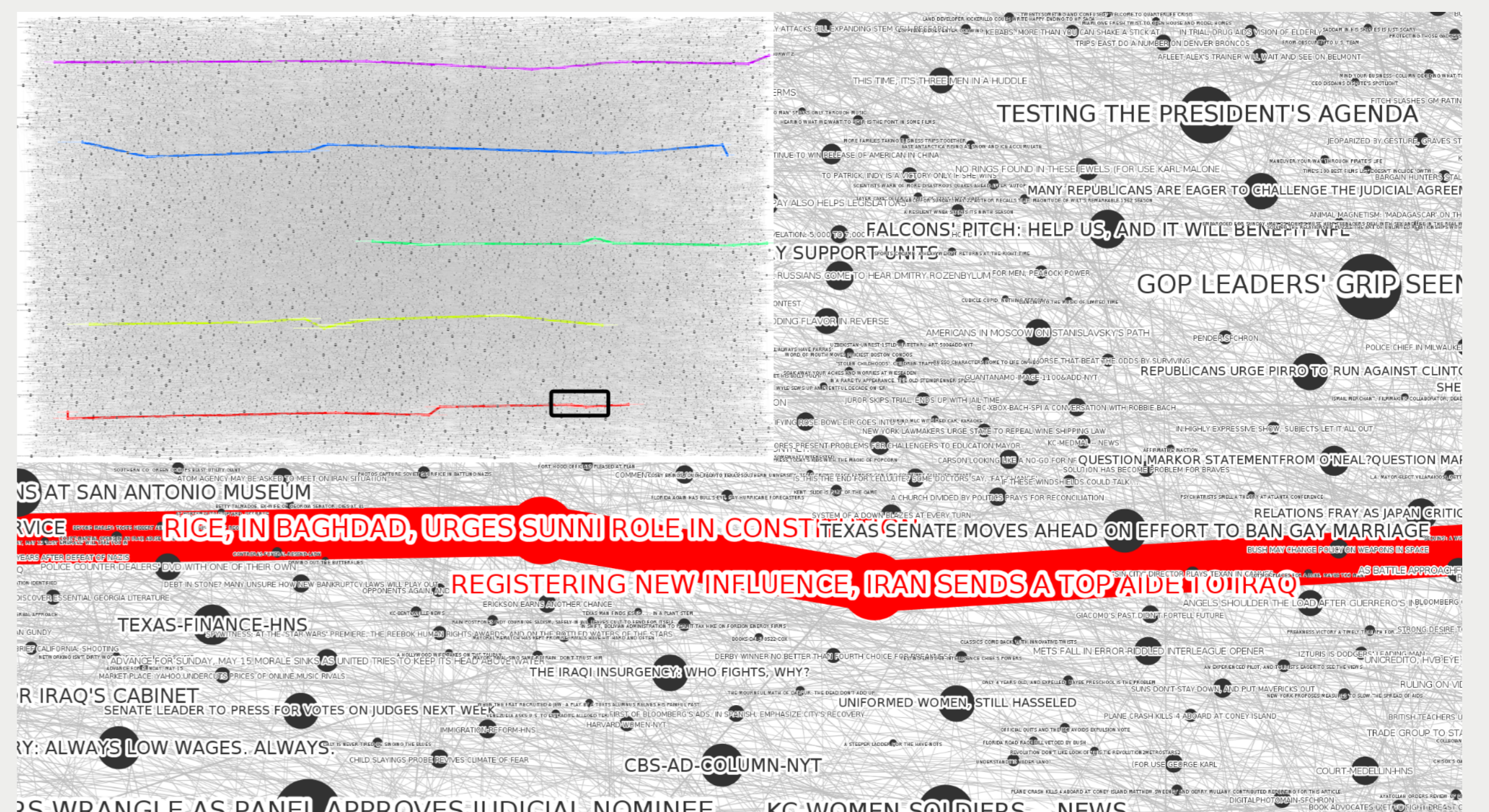


Figure: **Top left**: Full graph with 5 DPP threads. **Other**: Zoom in.