

Text Mining and Analysis in Railroad Safety



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Text Mining is...



- The derivation of information from unstructured data by automated systems, specifically text documents
- Determines patterns within the text in order to discover a word or phrase's relevance to the topic.

Applications



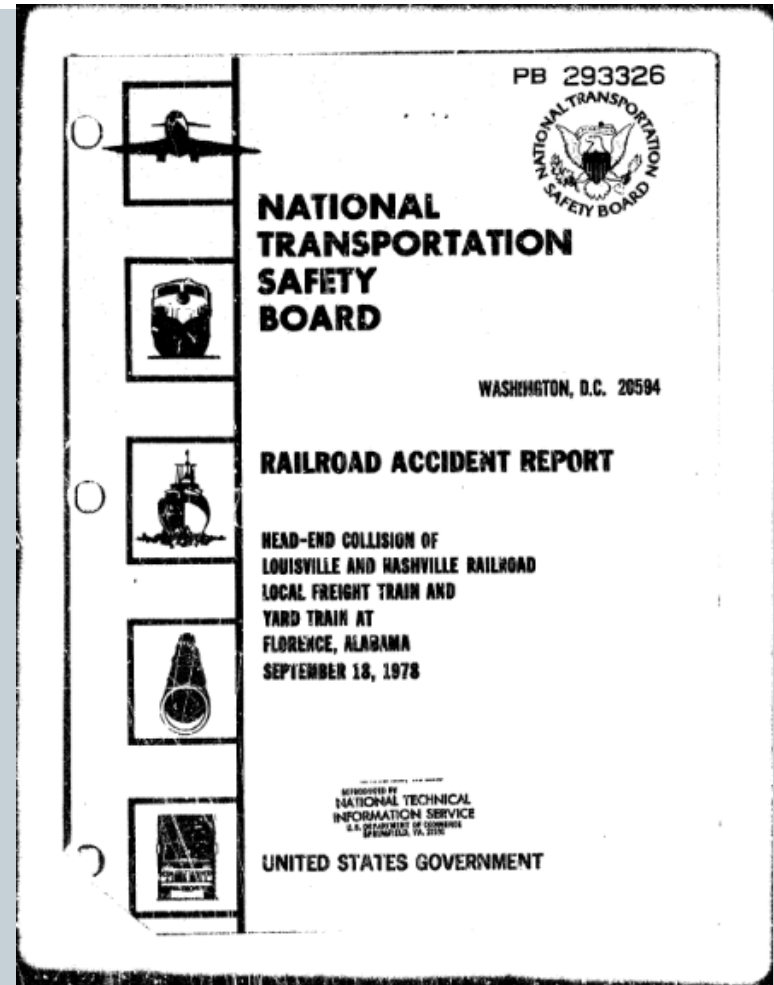
- Criminal Network Analysis
- Identifying Potential Causes of Medical Conditions
- Cost Analysis
- Identifying Common Causes of Accidents at Grade Crossings



What trends can be found in factors of
major railroad accidents?

The Data

- National Transportation Safety Board has made all accident reports available since 1996
- PDFs → Raw Text
- Processing Could Yield
 - Common Causes of Deaths
 - Common System Failures
 - Differences in Accidents based on Region



Techniques



- Latent Dirichlet Analysis
- Latent Semantic Indexing
- Clustering Analysis

Latent Dirichlet Analysis



- Returns the permeation of various words within the data collection, suggesting the most important topics

Table 2
Word stems associated with projects with large cost overruns.

Word	In documents	Total	High overrun project	Near low bid project	Under run
excav_asphalt	18	18	14	3	1
drainag	19	19	14	3	2
mobil_time	14	14	11	2	1
replac_bridg	14	14	10	4	0
cast_steel	7	10	8	2	0
steel_concret	7	10	8	2	0
overhead_structur	8	8	6	2	0
realign_roadwai	8	8	6	2	0
roadwai_bridg	8	8	6	2	0
temporari_signal	8	8	6	2	0
epoxi_coat	5	5	5	0	0
geosynthet	5	5	5	0	0
geosynthet_reinforc	5	5	5	0	0
reinforc_embank	4	5	5	0	0
temporari_fenc	4	5	5	0	0

Latent Semantic Indexing



- Patterns and association between words gives value to certain phrases more than others
- Used to derive 'meaning'
- Comparable to human cognition

Clustering Analysis



- A numeric set of values is assigned to a subject based on defining factors as determined by text mining
- Subjects are grouped based on these values, denoting similarity or dissimilarity

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