Statistical Topic Models for Computational Social Science

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Complex Social Processes



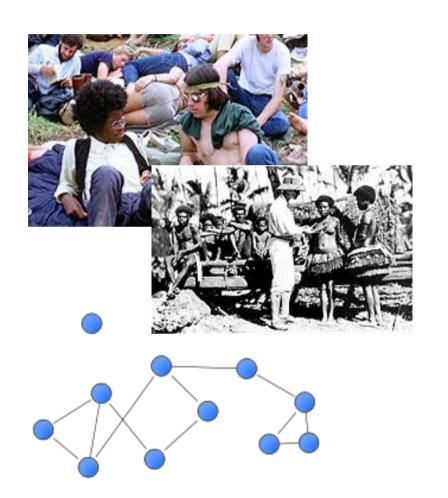






TOT. DECIMEN	21 FEB 67
2541 OUTLINE	Page
. Military actions against North Vietnam and In Laos	
A. Present program	1
B. Options for increased military programs	2
- Thermal power (7-plant grid)? - Steel and cement Marking tool plant 1. Destroy modern industry - Thermal power (7-plant grid)? - Steel and cement - Marking tool plant	3 ·
- Machine tool plant - Other By	re 4-6-93
2. Destroy dikes and levees	6

"Traditional" Social Science

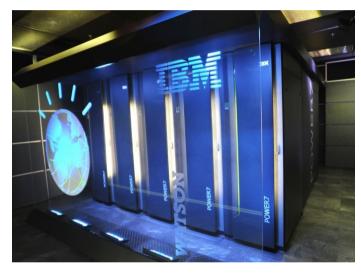


- Case studies
- Interviews
- Participant observation
- Survey research
- Social network analysis

⇒ Self-reports, one-time snapshots, small scale

The Computer "Revolution"



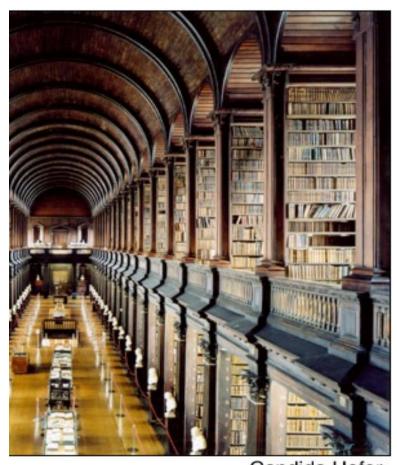






The Unreasonable Effectiveness of Data

Computational Social Science

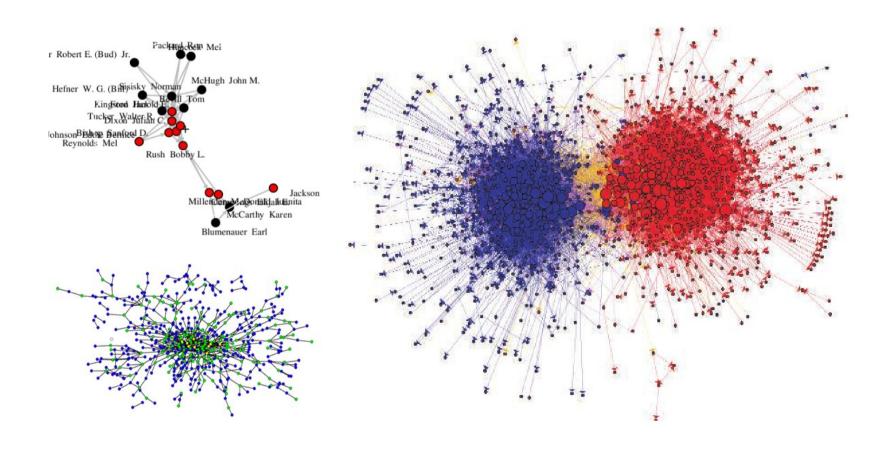


Candida Hofer

"A computational social science is emerging that leverages the capacity to collect and analyze data with an unprecedented breadth and depth and scale and may reveal patterns of individual and group behaviors."

— Lazer et al., 2009

Structure vs. Content



Products of Interactions



"Scientific information is both the basic raw material for, and one of the principal products of, scientific research [...] Scientists find out what other scientists are accomplishing through [...] journals, books, abstracts and indexes, bibliographies, reviews."

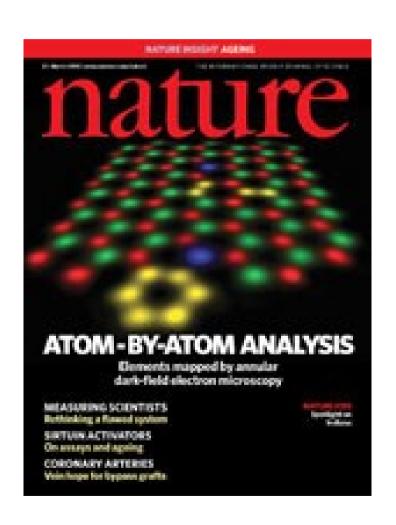
— NSF Brochure, 1962

Text as Data

United States Arnold, et a	Kerry to A		S. Policy Toward
	the Carnegie End policy in the Midd the event.	dowment for Interna	enator John Kerry, Chairman ational Peace in Washington, ⁄luasher, vice president for st.
A method, artic cryptographica key managemen	WHO:	Senator Joh	Abstract A natural evaluation metric
interface (API) i that provide an ways or with un	WHAT:	Speech on M	topic models is the probabil documents given a trained : exact computation of this pr tractable, several estimators
K	urnold; Todd urt S. (Roski		ability have been used in the ing literature, including the li- method and empirical likeliho this paper, we demonstrate that commonly-used methods accurately estimate the prob-
2541	TOP SECR	F	out documents, and propose methods that are both accura
	OUTLINE		Introduction
 Military action A. Present prog 		Vietnam and In L	aos 1
B. Options for l. Destroy m - Thermal - Steel a	increased milita odern industry power (7-plant	grid) ? SAI E.O. 12	2 NITIZED 3 356, Sec. 3.4 90-112 NARA, Date <u>4-6-</u> 93

- Structured and formal: e.g., publications, patents, press releases
- Messy and unstructured: e.g., chat logs, OCRed documents, transcripts
- ⇒ Large scale, robust methods for analyzing text

Collaborate to Study Collaboration



"There needs to be a greater focus on what these [interaction] data mean [...] This requires the input of social scientists, rather than just those more traditionally involved in data capture, such as computer scientists."

— Julia Lane, NSF, 24 March 2010

Different (But Overlapping) Roles



- Social science: specific models for specific applications, extensive post-analysis work
- Computer science:
 novel classes of models,
 mathematical and
 computational properties
 of models that extend
 across applications

This Talk

- Statistical topic models for text analysis
- "Off-the-shelf" topic models: priors, stop words
- Studying formerly-classified government documents

Statistical Modeling

- Modeling challenges:
 - Aggregating and representing large data sets
 - Handling data from sources with disparate emphases
 - Reasoning under uncertain information
 - Performing efficient inference
- Bayesian latent (hidden) variable models:
 - Powerful and flexible [Wallach et al. & Adams et al., AISTATS '10]
 - This talk: statistical topic models

Statistical Topic Modeling

- Three fundamental assumptions:
 - Documents have latent semantic structure ("topics")
 - We can infer topics from word-document co-occurrences
 - Can simulate this inference algorithmically
- Given a data set, the goal is to
 - Learn the composition of the topics that best represent it
 - Learn which topics are used in each document

Why Topic Models?

VS.

From (9) it can then be shown that (Exercise

$$\lambda = \{\mathbf{K}^{-1} - \mathbf{K}^{-1}\mathbf{M}(\mathbf{M}^{T}\mathbf{K}^{-1}\mathbf{M}) + \mathbf{K}^{-1}\mathbf{M}/\mathbf{M}^{T}\mathbf{K}^{-1}\mathbf{M})^{-1}\mathbf{n}\}$$

$$\lambda^T \mathbf{Z} = \mathbf{k}^T$$

which is identical to what w generalized least squares est

$$k_0 - \mathbf{k}^T \mathbf{F}$$

where $\gamma = \mathbf{m}(\mathbf{x}_0) - \mathbf{M}^T \mathbf{K}^-$

Best linear unbiased pred erature, named after the Sou 1951; Journel and Huijbregt process is assumed to be an prediction is called ordinary matrix more general m is known a with the mean assumed 0 is erally called objective analyst Pedder 1987 and Daley 1991

so that the resulting predict kriging covariance mean $k_0 - \mathbf{k}^T \mathbf{k}$ estimate weight random mse conditional point

linear unbiased prediction for regression moder did not explicitly consider the spatial setting. C further discussion on the history of various for As noted in 1.3, A useful characterization c **Definition 2.1** A Gaussian process is a ϵ finite number of which have a joint Gaussia

gaussian regression covariance prediction function bayesian

process prior

distribution

matrix

rocess is completely speci We define mean function process $f(\mathbf{x})$ as

$$m(\mathbf{x}) = \mathbb{E}[f(\mathbf{x})],$$

 $(\mathbf{x}, \mathbf{x}') = \mathbb{E}[(f(\mathbf{x}) - m(\mathbf{x}))]$

Gaussian process as

$$f(\mathbf{x}) \sim \mathcal{GP}(m(\mathbf{x}))$$

ional simplicity we will t I not be done, see section e random variables repres en, Gaussian processes ai andom variables is time. ere the index set X is the \mathbb{R}^{D} . For notational \mathfrak{g}

enumeration of the cases in the training se such that $f_i \triangleq f(\mathbf{x}_i)$ is the random variable as would be expected.

Topics and Words

probability

human
genome
dna
genetic
genes
sequence
gene
molecular
sequencing
map
...

evolution
evolutionary
species
organisms
life
origin
biology
groups
phylogenetic
living
...

host
bacteria
diseases
resistance
bacterial
new
strains
control
infectious

computer
models
information
data
computers
system
network
systems
model
parallel
...

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Documents and Topics

Seeking Life's Bare (Genetic) Necessities

Haemophilus

genome

Genes in common

233 genes

Mycoplasma genome 469 genes

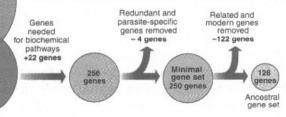
COLD SPRING HARBOR, NEW YORK—How many genes does an organism need to survive? Last week at the genome meeting here,* two genome researchers with radically different approaches presented complementary views of the basic genes needed for life. One research team, using computer analyses to compare known genomes, concluded that today's organisms can be sustained with just 250 genes, and that the earliest life forms

required a mere 128 genes. The other researcher mapped genes in a simple parasite and estimated that for this organism, 800 genes are plenty to do the job—but that anything short of 100 wouldn't be enough.

Although the numbers don't match precisely, those predictions

"are not all that far apart," especially in comparison to the 75,000 genes in the human genome, notes Siv Andersson of Uppsala University in Sweden, who arrived at the 800 number. But coming up with a consensus answer may be more than just a genetic numbers game, particularly as more and more genomes are completely mapped and sequenced. "It may be a way of organizing any newly sequenced genome," explains

Arcady Mushegian, a computational molecular biologist at the National Center for Biotechnology Information (NCBI) in Bethesda, Maryland. Comparing an



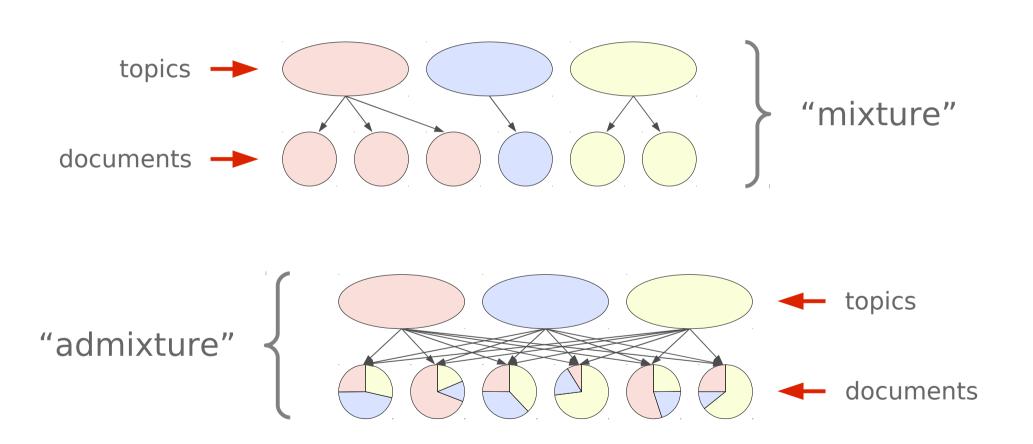
Stripping down. Computer analysis yields an estimate of the minimum modern and ancient genomes.

ing, Cold Spring Harbor, New York, May 8 to 12.

* Genome Mapping and Sequenc-

SCIENCE • VOL. 272 • 24 MAY 1996

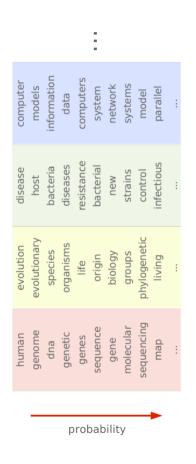
Mixtures vs. Admixtures



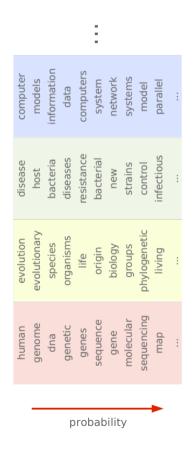
Generative Statistical Modeling

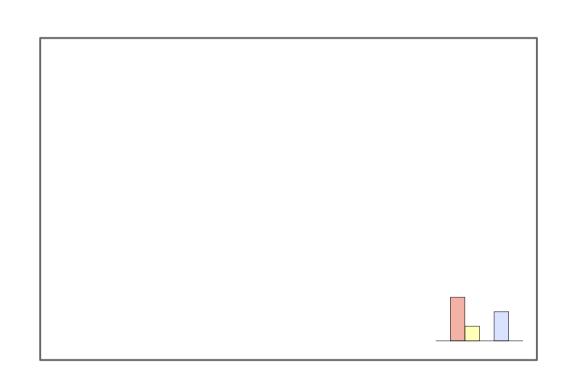
- Assume data was generated by a probabilistic model:
 - Model may have hidden structure (latent variables)
 - Model defines a joint distribution over all variables
 - Model parameters are unknown
- Infer hidden structure and model parameters from data
- Situate new data in estimated model

Generative Process



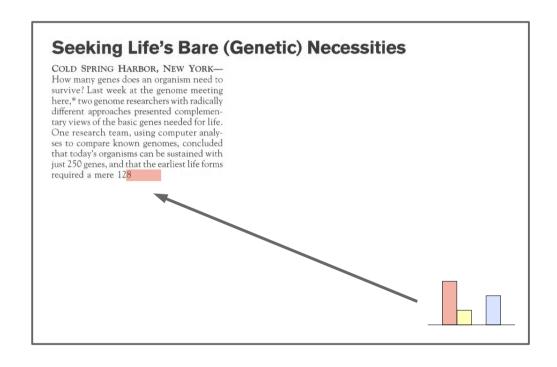
Choose a Distribution Over Topics



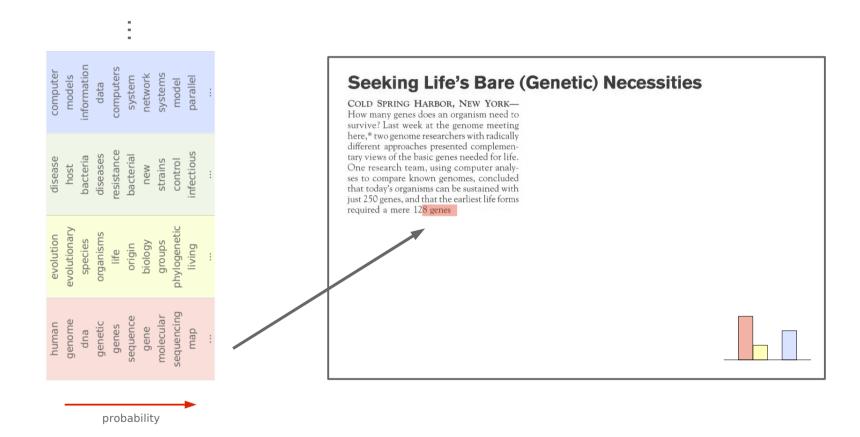


Choose a Topic

computers system network systems nformation computer parallel model data diseases resistance bacterial disease bacteria nfectious new strains control evolution evolutionary phylogenetic organisms species origin biology groups genes sequence gene molecular sequencing genetic map probability

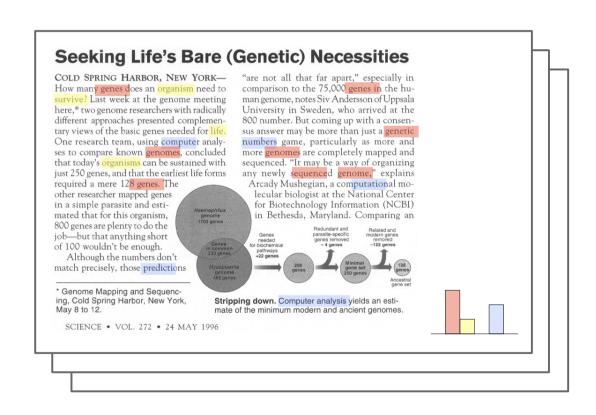


Choose a Word

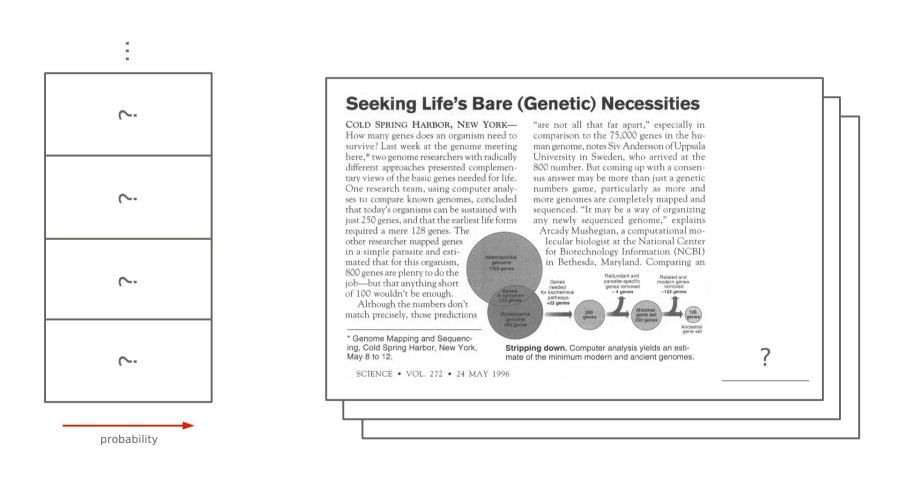


... And So On

nformation computers network systems computer models system parallel model data resistance diseases bacterial disease host bacteria nfectious strains control new evolution evolutionary phylogenetic organisms species origin biology groups living genes sequence gene molecular sequencing human genome dna genetic map probability

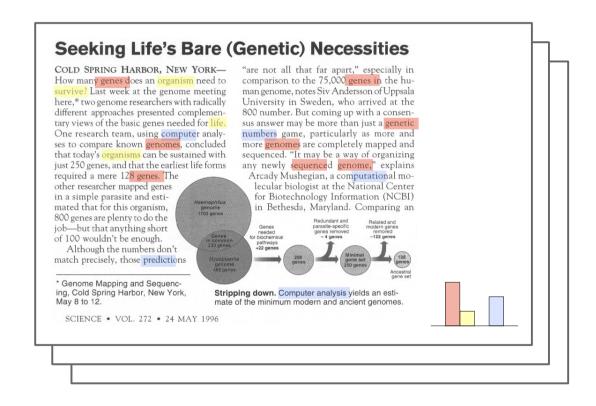


Real Data: Statistical Inference



The End Result...

nformation computers network systems computer models system parallel model data resistance diseases bacterial bacteria nfectious disease strains control new evolution evolutionary phylogenetic organisms species origin biology groups genes sequence gene molecular sequencing genetic map probability



This Talk

- Statistical topic models for text analysis
- "Off-the-shelf" topic models: priors, stop words
- Studying formerly-classified government documents

The State of The Art

- Topic models are extremely appealing
- ... but they're not always usable by non-experts
- Need to bridge this gap between producers and consumers of topic modeling technology:
 - Address problems/challenges faced by practitioners
 - Question unquestioned assumptions
 - Explore the interplay between theory and practice

"Off-the-Shelf" Topic Modeling



I want to model technology emergence by analyzing patent abstracts...

I have a statistical model that you can use...



"Off-the-Shelf" Topic Modeling



I want to model technology emergence by analyzing patent abstracts...

I have a statistical model that you can use...



а	а	the	the
field	the	of	invention
emission	carbon	а	of
an	and	to	to
electron	gas	and	present
		•••	

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"Off-the-Shelf" Topic Modeling?



Help! All my topics consist of "the, and of, to, a ..."



Now they all consist of "invention, present, thereof ..."



Wait, but how do I choose the right number of topics?

Preprocess your data to remove stop words...



Make a domain-specific list of stop words...



Evaluate the probability of unseen data for different numbers...

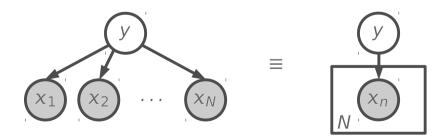


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Directed Graphical Models

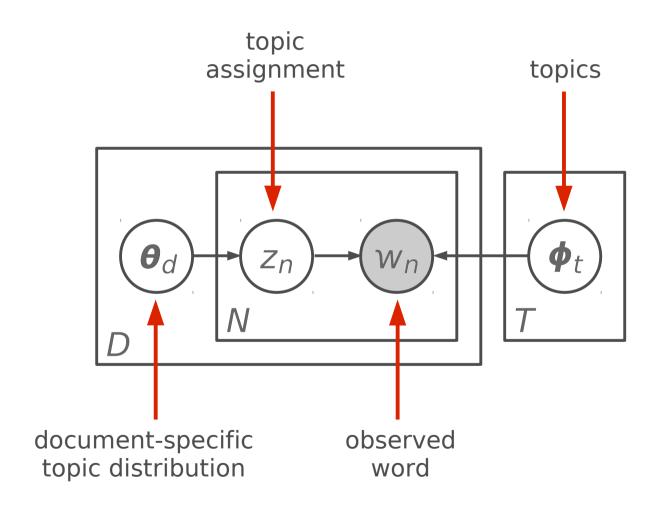
$$P(y, x_1, ..., x_N) = P(y) \prod_{n=1}^{N} P(x_n | y)$$

- Nodes: random variables (latent or observed)
- Edges: probabilistic dependencies between variables
- Plates: "macros" that allow subgraphs to be replicated



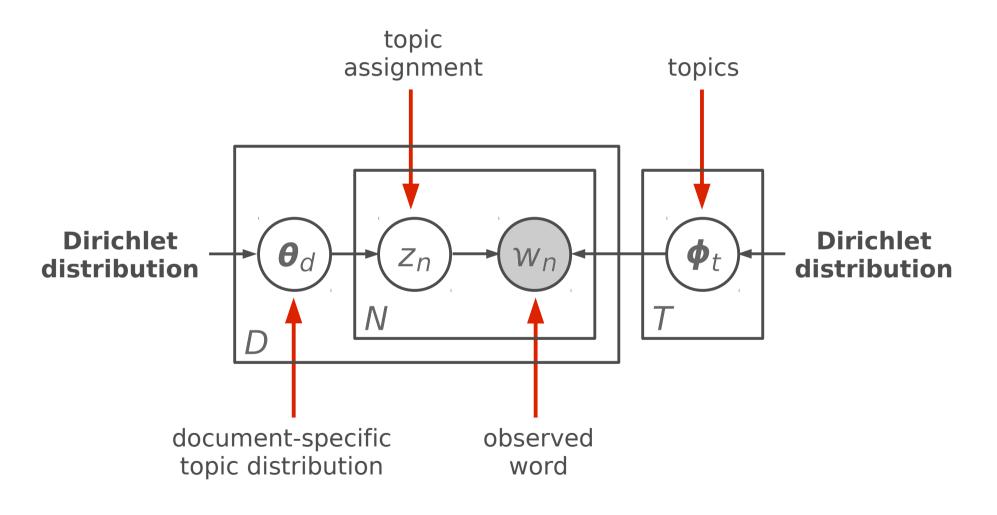
Statistical Topic Modeling

[Hofmann, '99]



Latent Dirichlet Allocation (LDA)

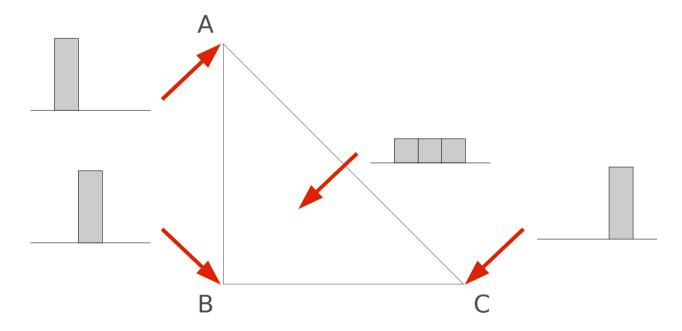
[Blei, Ng & Jordan, '03]



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Discrete Probability Distributions

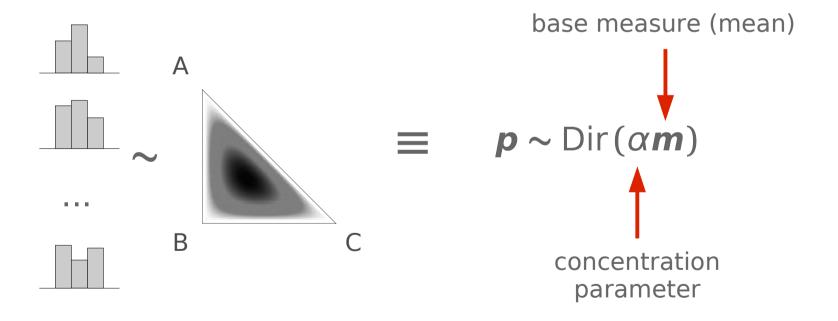
• 3-dimensional discrete probability distributions can be visually represented in 2-dimensional space:



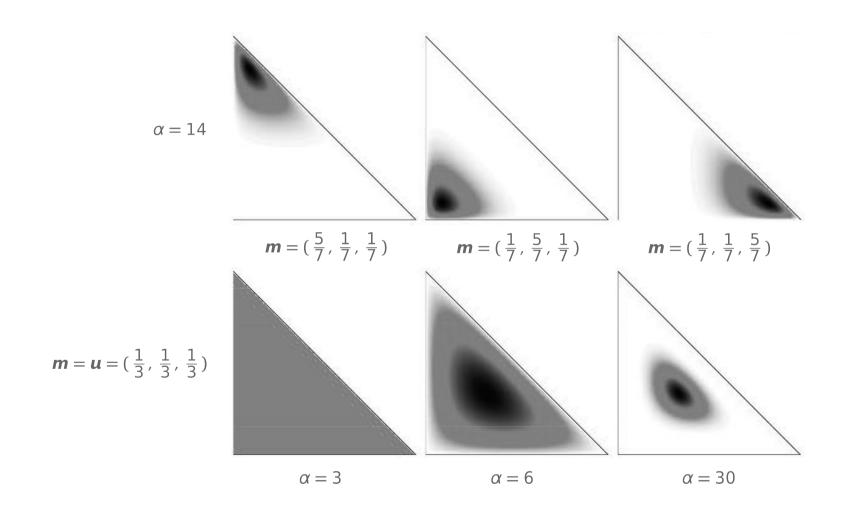
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Dirichlet Distribution

• Distribution over discrete probability distributions:

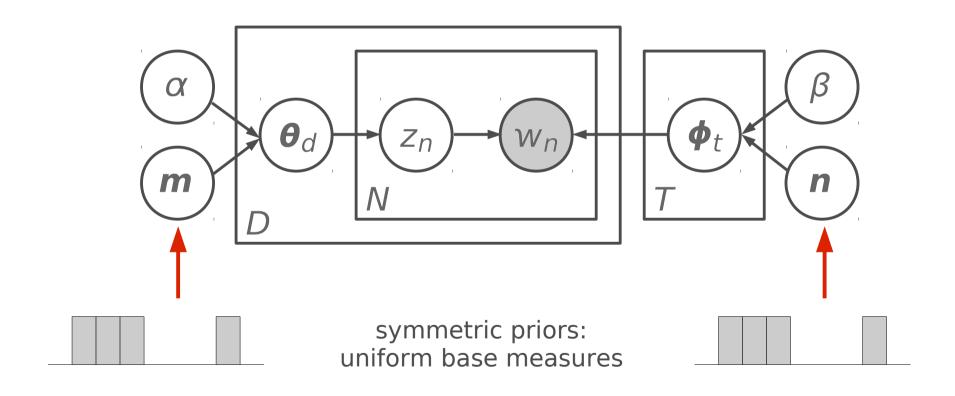


Dirichlet Parameters



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Dirichlet Priors for LDA



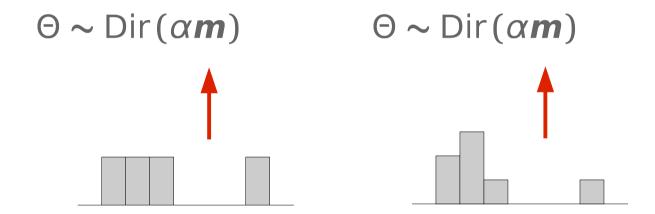
Dirichlet Priors for LDA

- Two scalar concentration parameters: α and β
- Concentration parameters are usually set heuristically
 - e.g., $\alpha = 50$ and $\beta = 0.01W$
- Some recent work on learning optimal values for the concentration parameters from data
- No rigorous study of the Dirichlet priors:
 - e.g., asymmetric vs. symmetric base measures
 - Effects of the base measures on the inferred topics

Symmetric → Asymmetric

[Wallach et al., '09]

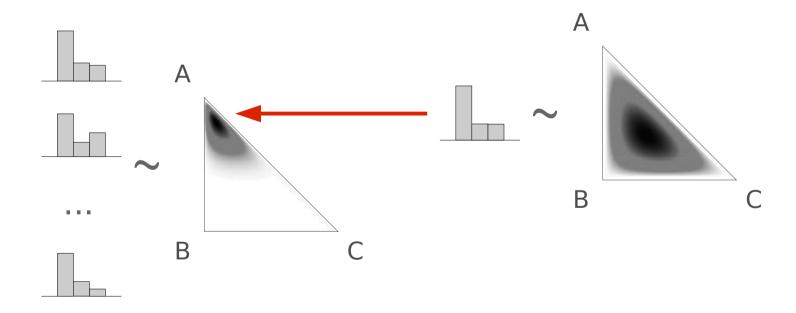
- Use prior over $\Theta = \{\theta_1, \dots, \theta_D\}$ as a running example
- Uniform base measure → nonuniform base measure



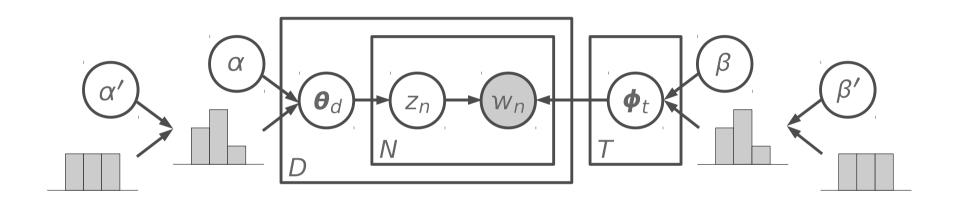
Asymmetric prior: some topics more likely a priori

Hierarchical Asymmetric Dirichlet

- Which topics should be more probable a priori?
 - Draw m from a Dirichlet distribution:



Putting Everything Together

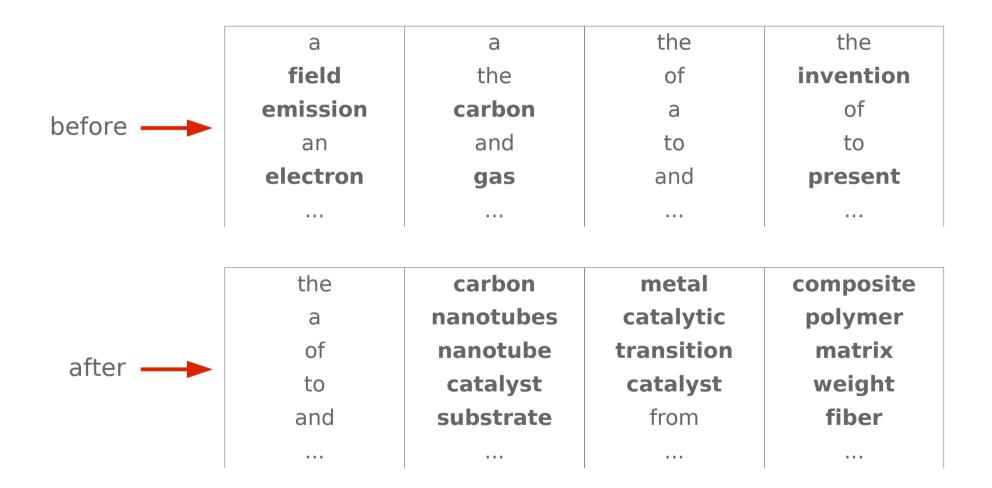


- Asymmetric hierarchical Dirichlet priors
- Integrate out Θ , Φ and base measures
- Learn z and concentration parameters from data

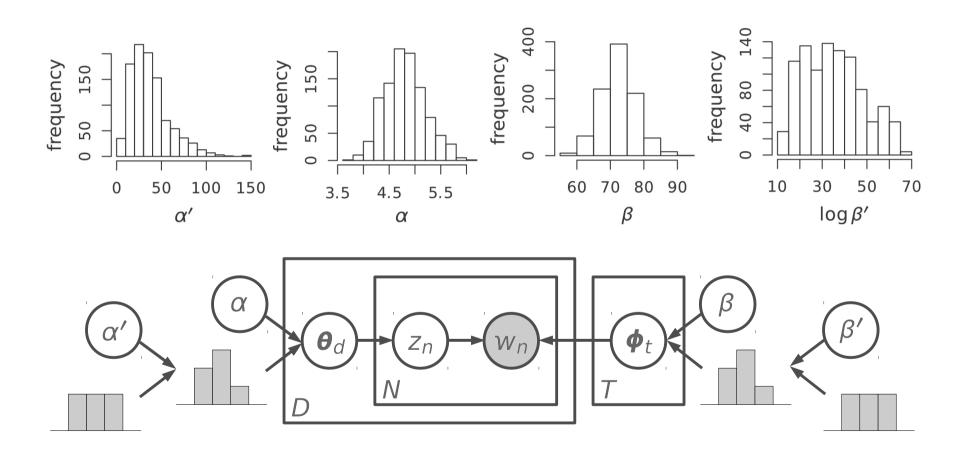
Data Sets

- Carbon nanotechnology patents:
 - Ultimate goal: track innovation and emergence
 - Fullerene and carbon nanotube patents
 - 1,016 abstracts (~100 words each)
 - 103,499 total words; 6,068 unique words
- 20 Newsgroups data (80,012 total words)
- New York Times articles (477,465 total words)

Inferred Topics

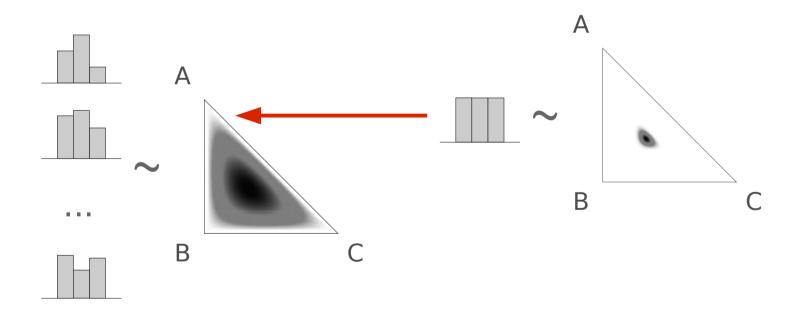


Sampled Concentration Parameters

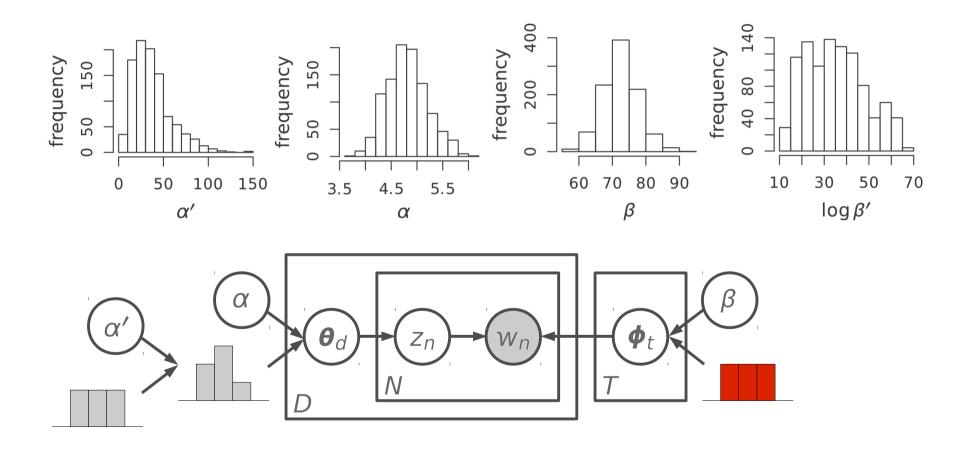


A Theoretical Observation...

 Symmetric Dirichlet is a special case of the hierarchical asymmetric Dirichlet (large concentration parameter)



Sampled Concentration Parameters



Intuition

- Topics should be distinct from each other:
 - Asymmetric prior over topics makes topics more similar to each other (and to corpus-wide word frequencies)
 - Want a symmetric prior to preserve topic "distinctness"
- Still have to account for power-law word usage:
 - Asymmetric prior over document-specific topic distributions means some topics (e.g., "the, a, of, to ...")
 can be used more often than others in all documents

"Off-the-Shelf" Topic Modeling



I can model technology emergence by analyzing patent abstracts!

Great! Let me know if you need any more help!



the	carbon	metal	composite
а	nanotubes	catalytic	polymer
of	nanotube	transition	matrix
to	catalyst	catalyst	weight
and	substrate	from	fiber

Polylingual Topics

```
CY sadwrn blaned gallair at lloeren mytholeg
DE space nasa sojus flug mission
EL διαστημικό sts nasa αγγλ small
```

EN space mission launch satellite nasa spacecraft

فضایی ماموریت ناسا مدار فضانورد ماهواره FA

FI sojuz nasa apollo ensimmäinen space lento

FR spatiale mission orbite mars satellite spatial

HE החלל הארץ חלל כדור א תוכנית

IT spaziale missione programma space sojuz stazione

PL misja kosmicznej stacji misji space nasa

RU космический союз космического спутник станции

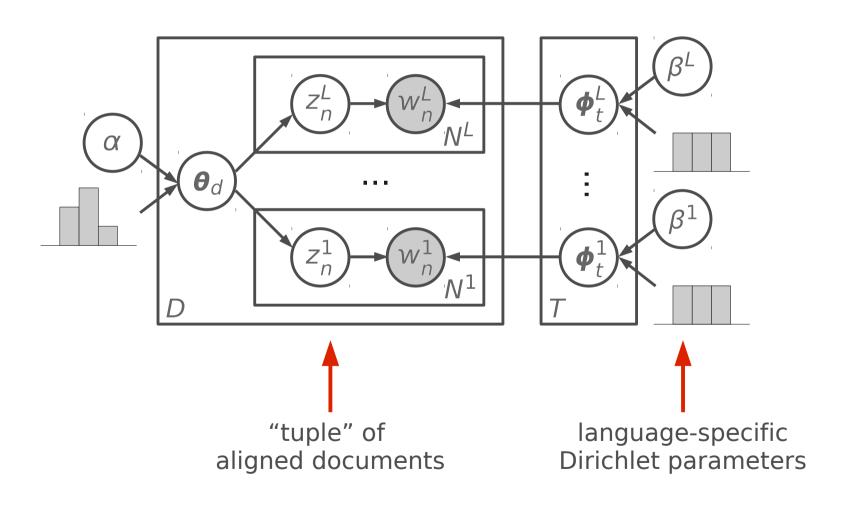
TR uzay soyuz ay uzaya salyut sovyetler

Polylingual Topics

```
bardd gerddi iaith beirdd fardd gymraeg
     dichter schriftsteller literatur gedichte gedicht werk
     ποιητής ποίηση ποιητή έργο ποιητές ποιήματα
FL
     poet poetry literature literary poems poem
ΕN
شاعر شعر ادبیات فارسی ادبی آثار FA
     runoilija kirjailija kirjallisuuden kirjoitti runo julkaisi
FI
     poète écrivain littérature poésie littéraire ses
FR
HF
     משורר ספרות שירה סופר שירים המשורר
ΙT
     poeta letteratura poesia opere versi poema
     poeta literatury poezji pisarz in jego
PL
RU
     поэт его писатель литературы поэзии драматург
TR
     şair edebiyat şiir yazar edebiyatı adlı
```

Polylingual Topic Model

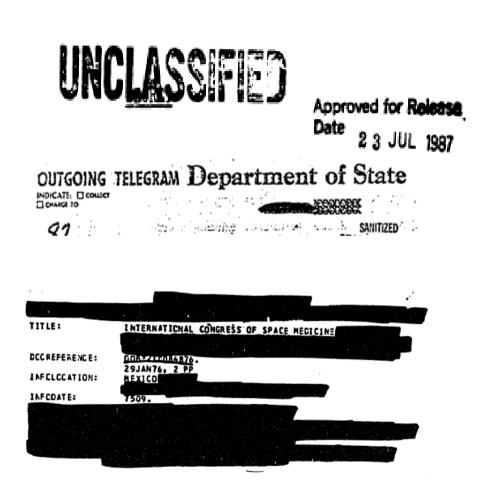
[Mimno et al., '09]



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In 2009 Alone...



- 52 million pages reviewed for declassification
- 29 million pages declassified
- \$8.8 billion spent on administration of the US government classification system

How Sensitive?

"After a 14-year legal battle by a California history professor, the FBI has released a new cache of material from a 300-page dossier on the late rock star John Lennon, and has agreed to pay \$204,000 to cover legal fees incurred in his efforts to open the file. For all the years of challenge, however, the file contains little, if any, new information about Lennon, though it does present some bizarre details, like a description of an antiwar activist trying to train a parrot to speak profanities."

- NYT, 25 September 2007

A Problematic Trade-off

```
D DIRECTOR, FEDERAL BUREAU OF INVESTIGATION

O FROM: DIRECTOR, CENTRAL INTELLIGENCE AGENCY

At no time has Acting Director Gates recommended an invasion of Libya. Moreover, any insinuation that Mr. Gates

O SUBJECT: in July 1985 encouraged such action is unfounded.
```

- The more data kept secret, the less secure the data:
 - More people need to have access to the data
 - More storage space is required

What We Are NOT Studying...





guardian.co.uk

News | Sport | Comment | Culture | Business | Money

News \rangle World news \rangle Afghanistan: The war logs

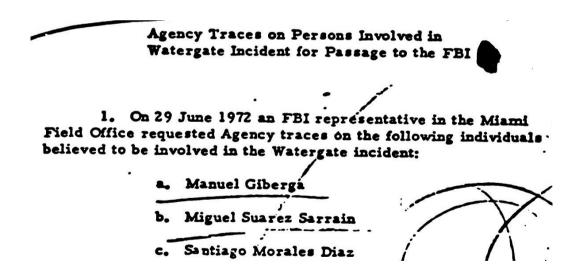
AFGHANISTAN THE WAR LOGS

Exploring Declassified Documents

- Declassification goals:
 - Recommend documents for human review
 - Match documents with human reviewers' expertise
- Transparency research goals:
 - High-level characterization of the data
 - Finding specific, known information of interest
 - Finding "interesting" or "unexpected" information

Declassified Documents: DDRS

- ~88,000 formerly-classified government documents
- Created and declassified between 1926 and 2005



CENTRAL INTELLIGENCE AGENCY WASHINGTON, D.C. 20505

29 January 1968

0006

MEMORANDUM FOR: The Honorable Walt W. Rostow Special Assistant to the President The White House

SUBJECT

: Coal and Electric Power Shortages

in Communist China

1. (Al Jenkins asked that we prepare the attached memorandum on shortages of coal and electric power in Communist China for your information. We have also included excerpts from individual reports of shortages to give you some feeling for the information available.

2. While there is no question that the shortages are widespread, it is extremely difficult to quantify the decline in industrial output caused by these shortages or by other effects of the Cultural Revolution.

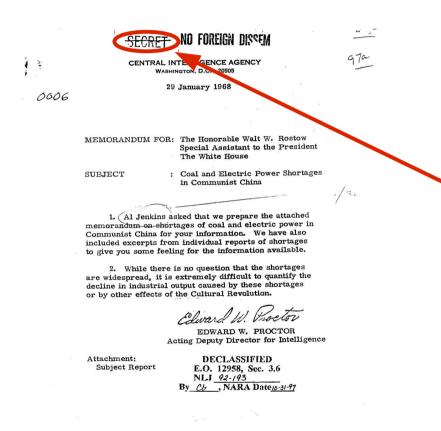
> EDWARD W. PROCTOR Acting Deputy Director for Intelligence

Attachment. Subject Report

DECLASSIFIED E.O. 12958, Sec. 3.6 NLI 92-193 By Cb , NARA Date 10-31-97 Sanitized?

- Classification level
- Issuer
- Creation date
- Document type
- Declassification date

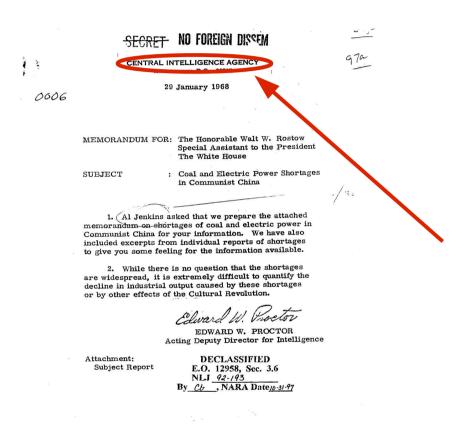
-SECRFT NO FOREIGN DISSEM



Sanitized?

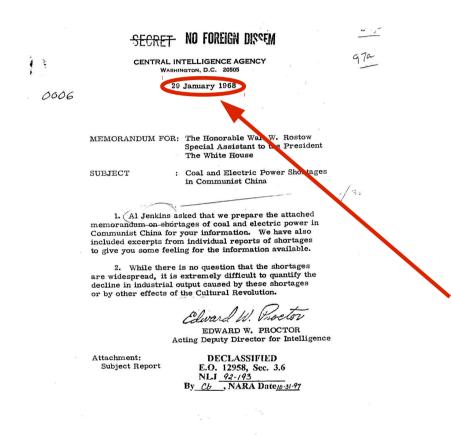
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SECRET NO FOREIGN DISCEM CENTRAL INTELLIGENCE AGENCY WASHINGTON, D.C. 20505 29 January 1968 0006 MEMORANDUM FOR: The Honorable Walt W. Rostow Special Assistant to the President The White House Coal and Electric Power Shortages SUBJECT ommunist China 1. (Al Jenkins asked that we prepare the attached memorandum-on-shortages of coal and electric power in Communist China for your information. We have also included excerpts from individual reports of shortages to give you some feeling for the information available. 2. While there is no question that the shortages are widespread, it is extremely difficult to quantify the decline in industrial output caused by these shortages or by other effects of the Cultural Revolution. EDWARD W. PROCTOR Acting Deputy Director for Intelligence Attachment. DECLASSIFIED Subject Report E.O. 12958, Sec. 3.6 NLI 92-193 By Cb , NARA Date 10-31-97

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Special Assistant to the President

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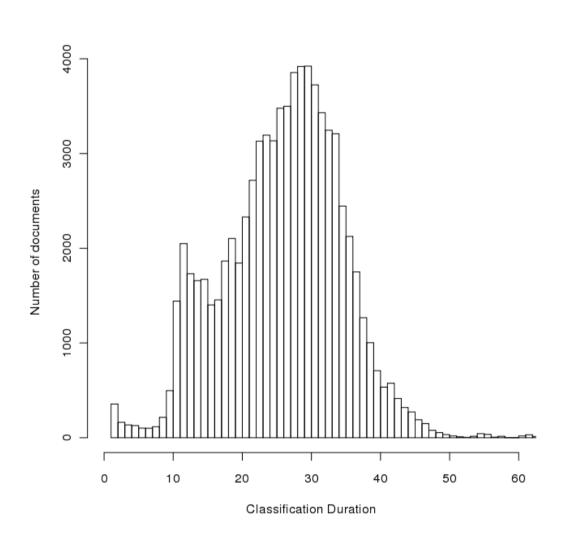
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- **Declassification date**

-SECRFT NO FOREIGN DISSEM

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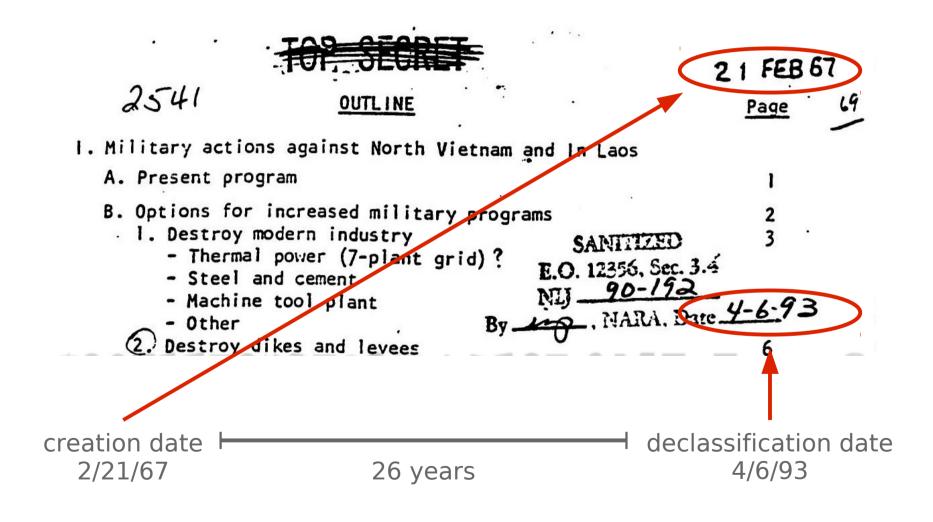
Declassification Durations



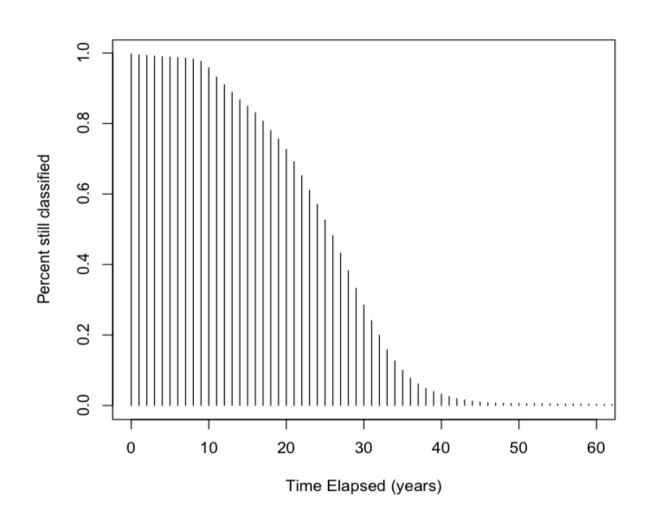
Survival Analysis

- Statistical methods for evaluating "time until death":
 - Biology/medicine: organism death
 - Engineering: component failure
 - Social science: event durations (e.g., parolee recidivism)
- Goal: model effect on survival time of covariates, e.g.,
 - Vaccine treatments
 - Temperature differences
 - Job placement or education programs

Document "Survival"



Survival Distribution of Documents



Accelerated Failure Time Models

- Survival analysis with covariates x_i
- Linear models for the log of the "duration":

$$\log(t_i) = \mathbf{x}_i^{\top} \boldsymbol{\beta} + \epsilon_i$$

- Parametric: a probability distribution is specified
 - e.g., Weibull, log-normal, gamma, log-logistic...
- Can make predictions for unseen data

Classification and Content

HIS APPROACH WAS, "WELL, OF COURSE, WE KNOW THERE ISN'T ANYTHING TO THIS ALLEGED PHENOMENON (FLYING SAUCERS), BUT ON THE OTHER HAND". DURING HIS TALK SHKLOVSKIY AND CTHEF SCVIETS JOKED AND LAUGHED AND CEVIOUSLY DID NOT TAKE THE SPEAKER'S REMARKS SERIOUSLY.



1975 to 1989

1946 to 2003

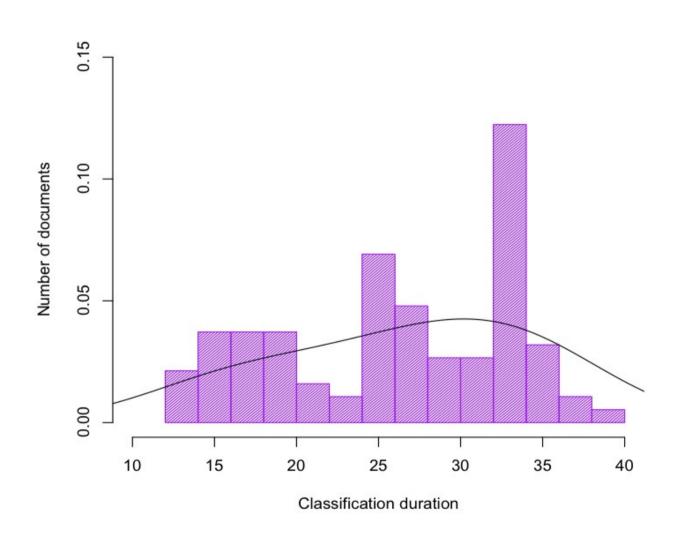


CENTRAL INTELLIGENCE GROUP

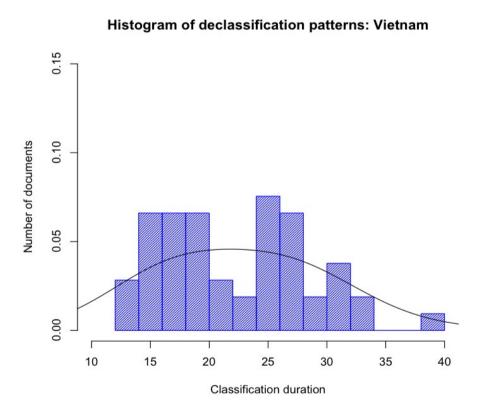
SOVIET CAPABILITIES FOR THE DEVELOPMENT AND PRODUCTION OF CERTAIN TYPES OF WEAPONS AND EQUIPMENT

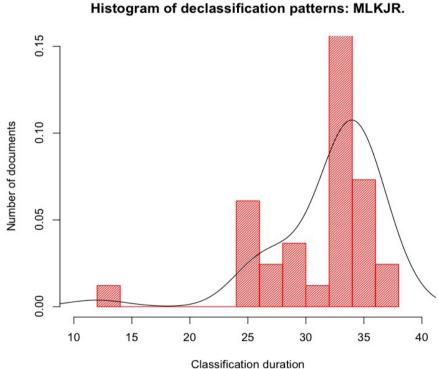
1. Herein is presented an estimate of Soviet capabilities in the development and production, during the next ten years, of certain weapons and equipment, as follows:

Classification and Content: 1960s



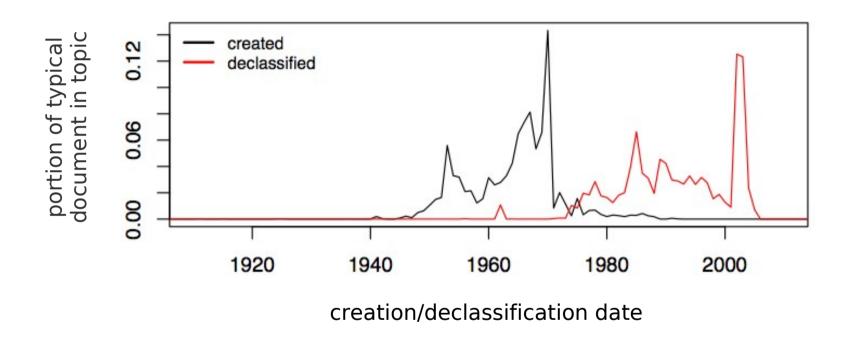
Classification and Content: 1960s



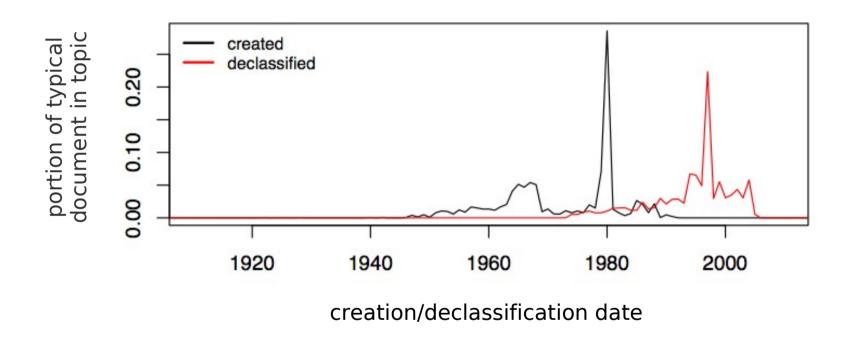


Word Frequencies?

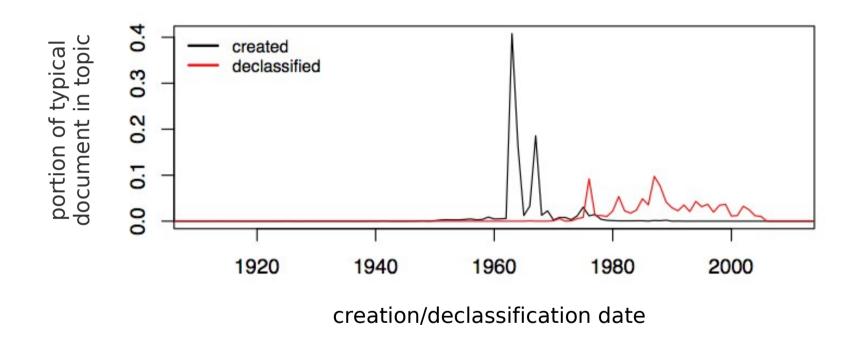
```
Bechuanaland is in effect, an enclave in the "White redoubt" of Southern
 frica, surrounded as it is by South Africa, Southern Rhodesia and South
West Africa.
           Its economy is wholly integrated with that of its white-
governed neighb
                            , the geographical and economic facts of
life make it imp rhodesia
                            erritory to insulate itself from the crises
affecting its neig africa
               southern
               khama
      Under Kl white
                            ship, Botswana has created
                            cieties in Africa and maintained
one of the mo
it in the fac african
                            turbulence in the white-ruled
states surrounding bt
                           One of only three countries in
Africa considered wholly "free" by Freedom House, its
democratic government and tolerant, multi-racial
society could well serve as a model of what the U.S.
is trying to accomplish in Rhodesia and Namibia.
```



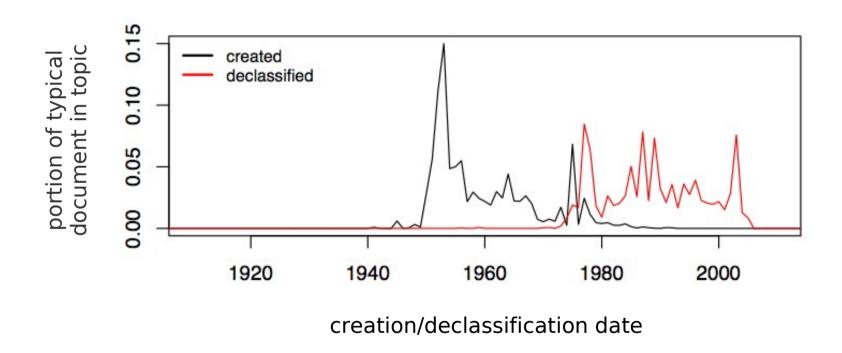
corps, service, volunteers, men, volunteer, age, draft, selective, calls, young, manpower, year, army, deferments, induction, armed, freedom, ...



package, hostages, release, hostage, khomeini, packages, ghotbzadeh, held, released, banisadr, revolutionary, debriefing, scenario, family, date, ...

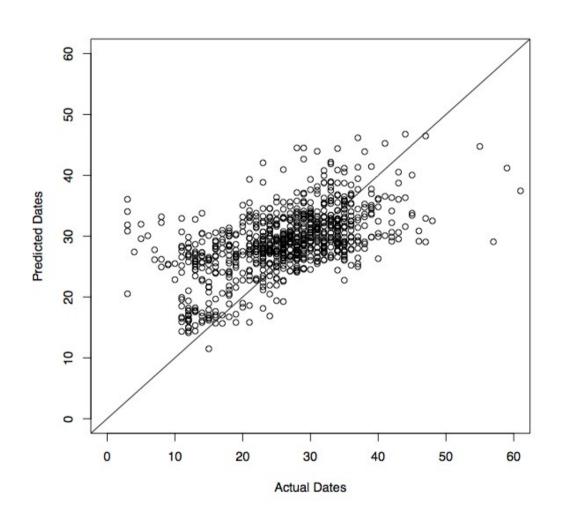


oswald, dallas, assassination, kennedy, texas, fbi, orleans, advised, lee, president, bureau, started, harvey, john, information, ruby, november, ...

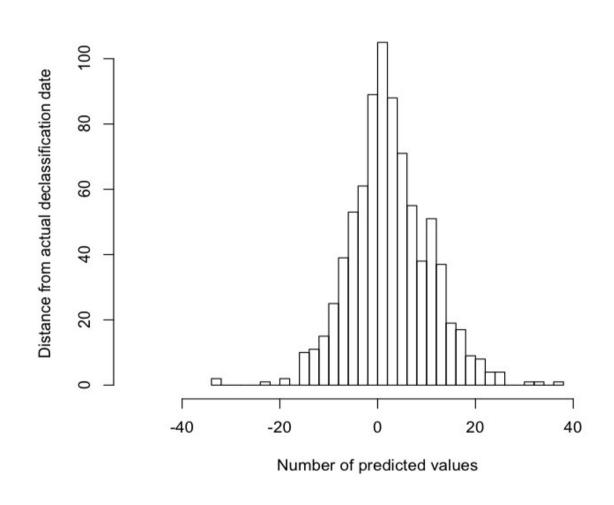


artichoke, subject, drugs, techniques, work, interrogation, writer, drug, lsd, effects, hypnosis, methods, medical, physical, subjects, human, ...

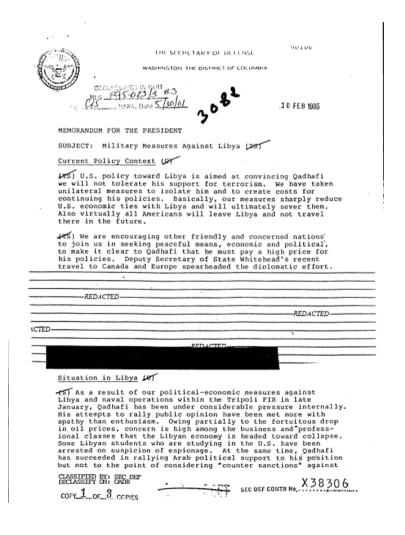
Predicting Duration Using Topics



Predicted Duration - Actual Duration



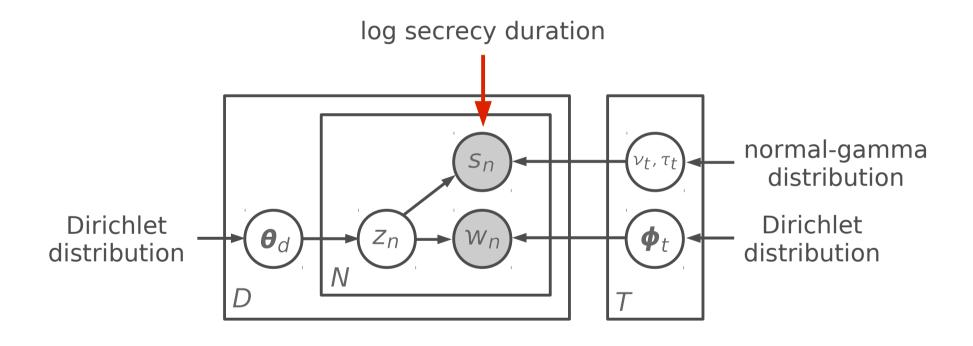
Jointly Modeling Text and Duration



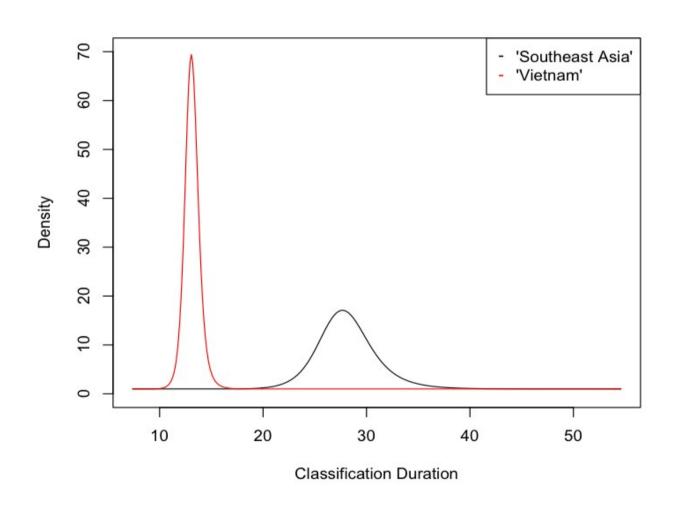
- Topics provide information about classification durations
- Goal: incorporate durations into the generative model
- Infer latent topics using both textual and temporal information

Jointly Modeling Text and Duration

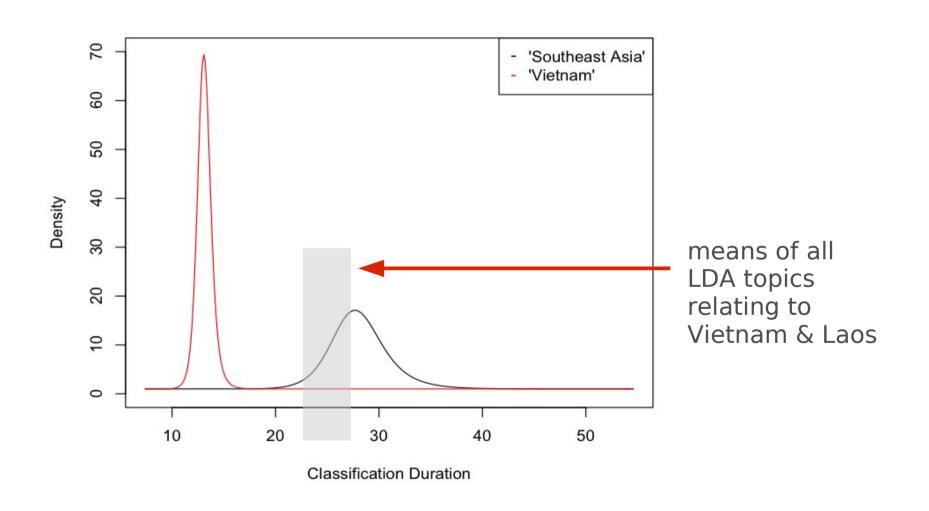
[Shorey et al., '11]



Topic-Specific Duration Distributions



Topic-Specific Duration Distributions



What's Next?

- Predict durations directly from the generative model
 - Mixture vs. admixture topics
 - Supervised topic modeling
 - Unseen content
- Subject matter experts
- Analysis and prediction of redactions

Thanks!

Acknowledgements: B. Desmarais, A. McCallum, D. Mimno, R. Shorey

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