

Polylingual Topic Models

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Statistical Topic Models

- ▶ Useful for analyzing large, unstructured text collections

bounds	units	policy	data	neurons
bound	hidden	action	space	neuron
loss	network	reinforcement	clustering	spike
functions	layer	learning	points	synaptic
error	unit	actions	distance	firing

- ▶ Topic-based search interfaces (<http://rexa.info>)
- ▶ Analysis of scientific progress over time (Blei & Lafferty, '07)
- ▶ Information retrieval (Wei & Croft, '06)

Automated Analysis of Text

- ▶ Previously: analyzing trends in text collections (Hall et al., '08)
- ▶ Monolingual models often work well: collections in English only
- ▶ Multilingual text collections are increasingly common
- ▶ Automated tools are most important for multilingual collections:
 - ▶ Don't know the language → cannot eyeball the data
 - ▶ New documents will appear in other languages
 - ▶ People typically only know a few languages
- ▶ Simultaneously analyze document content in many languages

Multiple Languages

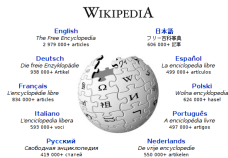
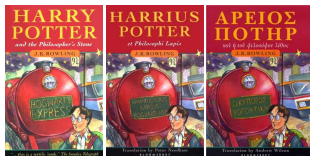
- ▶ Why model multiple languages explicitly?
- ▶ Most statistical topic models are language-agnostic

graph	problem	rendering	algebra	und	la
graphs	problems	graphics	algebras	von	des
edge	optimization	image	ring	die	le
vertices	algorithm	texture	rings	der	du
edges	programming	scene	modules	im	les

- ▶ Hodgepodge of English, German, French topics
- ▶ Imbalanced corpus: maybe only one or two French topics

Parallel vs. Comparable Corpora

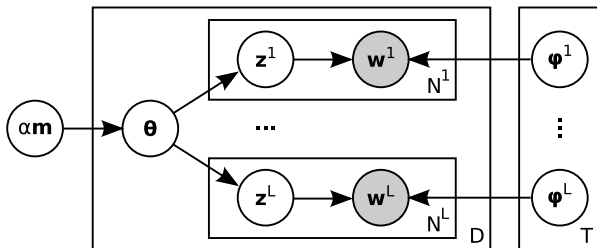
- ▶ A set of aligned documents is a “document tuple”



- ▶ Fully parallel corpora: documents are direct translations
- ▶ Corpora with a few parallel “glue” document tuples
- ▶ Comparable corpora: documents have similar semantic content

Polylingual Topic Model

- ▶ Generates a document tuple $\mathbf{w} = \mathbf{w}^1, \dots, \mathbf{w}^L$ by drawing...



- ▶ For real-world data, only the word tokens are observed

Key Characteristics

- ▶ Learning a model of *all* languages simultaneously
- ▶ A topic is a *set* of distributions over words, e.g., $\phi_t = \phi_t^1, \dots, \phi_t^L$
- ▶ Works on tuples of aligned documents, rather than documents, but each tuple can be comprised of only a subset of languages
- ▶ Tuple-specific topic distributions ensure cross-language consistency: e.g., topic 13 in French is semantically similar to topic 13 in English
- ▶ Simple, Gibbs sampling inference algorithm
 - ▶ Inference is linear in # of languages, not # of language pairs

EuroParl: Example Topics ($T = 400$)

DA centralbank europæiske ecb s lån centralbanks
DE zentralbank ezb bank europäischen investitionsbank darlehen
EL τράπεζα τράπεζας κεντρική εκτ κεντρικής τράπεζες
EN **bank central ecb banks european monetary**
ES banco central europeo bce bancos centrales
FI keskuspankin ekp n euroopan keskuspankki eip
FR banque centrale bce européenne banques monétaire
IT banca centrale bce europea banche prestiti
NL bank centrale ecb europese banken leningen
PT banco central europeu bce bancos empréstimos
SV centralbanken europeiska ecb centralbankens s lån

EuroParl: Example Topics ($T = 400$)

DA mål nå målsætninger målet målsætning opnå
DE ziel ziele erreichen zielen erreicht zielsetzungen
EL στόχους στόχο στόχος στόχων στόχοι επίτευξη
EN **objective objectives achieve aim ambitious set**
ES objetivo objetivos alcanzar conseguir lograr estos
FI tavoite tavoitteet tavoitteena tavoitteiden tavoitteita tavoitteen
FR objectif objectifs atteindre but cet ambitieux
IT obiettivo obiettivi raggiungere degli scopo quello
NL doelstellingen doel doelstelling bereiken bereikt doelen
PT objetivo objetivos alcançar atingir ambicioso conseguir
SV mål målet uppnå målen målsättningar målsättning

EuroParl: Example Topics ($T = 400$)

DA andre anden side ene andet øvrige
DE anderen andere einen wie andererseits anderer
EL άλλες άλλα άλλη άλλων άλλους όπως
EN **other one hand others another there**
ES otros otras otro otra parte demás
FI muiden toisaalta muita muut muihin muun
FR autres autre part côté ailleurs même
IT altri altre altro altra dall parte
NL andere anderzijds anderen ander als kant
PT outros outras outro lado outra noutros
SV andra sidan å annat ena annan

Parallel Corpora: “Glue” Tuples

- ▶ How many aligned documents are needed to get aligned topics?

1% “glue” document tuples

DE	rußland russland russischen tschetschenien sicherheit
EN	china rights human country s burma
IT	ho presidente mi perché relazione votato

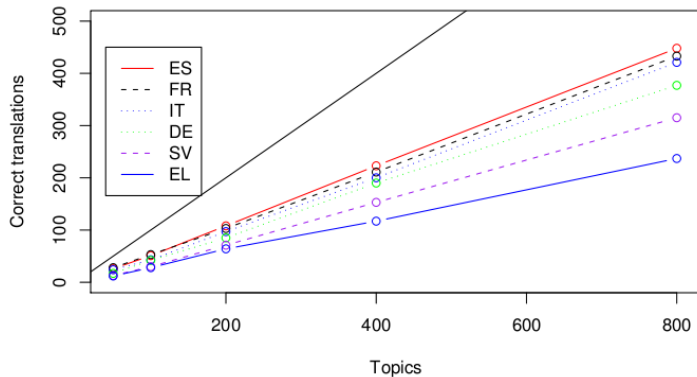
25% “glue” document tuples

DE	rußland russland russischen tschetschenien ukraine
EN	russia russian chechnya cooperation region belarus
IT	russia unione cooperazione cecenia regione russa

Generating Bilingual Lexica

- ▶ Bilingual lexicon: word pairs (e.g., English word, translation)
- ▶ High probability words in different languages for a topic are likely to include translations – can use these to generate lexica
- ▶ Advantages: unsupervised; all kinds of words, not just nouns
- ▶ Form candidate translations: Cartesian product of most probable K words in English and in each translation language
- ▶ Count # of lexicon pairs that are in the candidate set
- ▶ No morphological variants: e.g., rules/vorschriften, rule/vorschrift

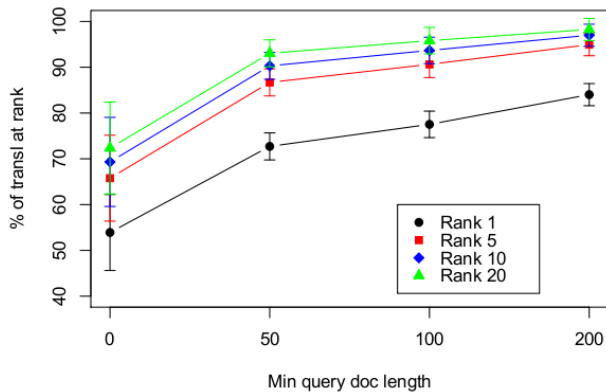
Generating Bilingual Lexica ($K = 1$)



Finding Translations

- ▶ Train model on aligned document tuples
- ▶ Output: set of polylingual topics, e.g., $\phi_t = \phi_t^1, \dots, \phi_t^L$
- ▶ Map each test document to the low-dimensional space defined by the polylingual topics \rightarrow document-topic distributions
- ▶ For each query/target language pair:
 - ▶ Compute similarities for all query/target document pairs
 - ▶ For each query document, rank target documents by similarity
- ▶ Jensen-Shannon divergence, cosine distance

Finding Translations (Jensen-Shannon)



Comparable Corpora

- ▶ Directly parallel translations are rare, expensive to produce
- ▶ Comparable corpora more common: e.g., Wikipedia, web pages
 - ▶ Our data set: all Wikipedia articles in English, Farsi, Finnish, French, German, Greek, Hebrew, Italian, Polish, Russian, Turkish, Welsh
- ▶ Documents are topically similar but not direct translations
- ▶ More interesting questions, more real-world applications:
 - ▶ Do comparable document tuples support alignment of topics?
 - ▶ Do different languages have different perspectives?
 - ▶ Which topics do particular languages focus on?

Wikipedia: Example Topics ($T = 400$)

CY	sadwm blaned gallair at lloeren mytholeg
DE	space nasa sojus flug mission
EL	διαστημικό sts nasa αγγλ small
EN	space mission launch satellite nasa spacecraft
FA	فضایی ماموریت ناسا مدار فضاورد ماهواره
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 sojus stazione
PL	misja kosmicznej stacji misji space nasa
RU	космический союз космического спутник станции
TR	uzay soyuz ay uzaya salyut sovyetler

Wikipedia: Example Topics ($T = 400$)

CY sbaen madrid el la josé sbaeneg
DE de spanischer spanischen spanien madrid la
EL ισπανίας ισπανία de ισπανός ντε μαδρίτη
EN **de spanish spain la madrid y**
FA اسپانيا اسپانيايي کوبا مادريد de ترين
FI espanja de espanjan madrid la real
FR espagnol espagne madrid espagnole juan y
HE ספרד ספרדית דה מדריד הספרדית קובה
IT de spagna spagnolo spagnola madrid el
PL de hiszpański hiszpanii la juan y
RU де мадрид испании испания испанский de
TR ispanya ispanyol madrid la küba real

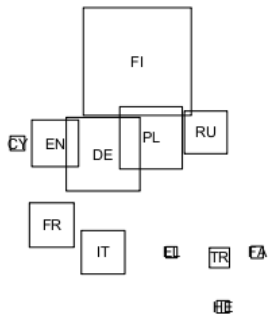
Wikipedia: Example Topics ($T = 400$)

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

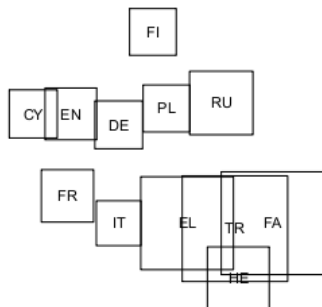
Topic Divergence Between Languages

- ▶ Estimate document-specific distributions over topics
- ▶ Compute Jensen-Shannon divergence between documents in a tuple
- ▶ Average document-document divergences for each language pair:
 - ▶ “Disagreement” score for each language pair
- ▶ Almost all pairs have divergences consistent with EuroParl, even languages that have historically been in conflict
- ▶ Although individual articles may have high between-language divergence, Wikipedia is on average consistent between languages

Differences in Topic Emphasis



world ski km won...



ottoman empire khan byzantine...

Conclusions

- ▶ Can discover topics aligned across multiple languages
- ▶ A small number of aligned documents is sufficient to align topics
- ▶ Can use the model to create bilingual lexica and find translations
- ▶ For comparable corpora, e.g., Wikipedia, someone who speaks *any one* language can perform data-driven analysis of topic trends, similarities and differences in *all* available languages
- ▶ Future work: adapting machine translation and cross-language information retrieval systems to new domains

Questions?

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<http://www.cs.umass.edu/~wallach/>