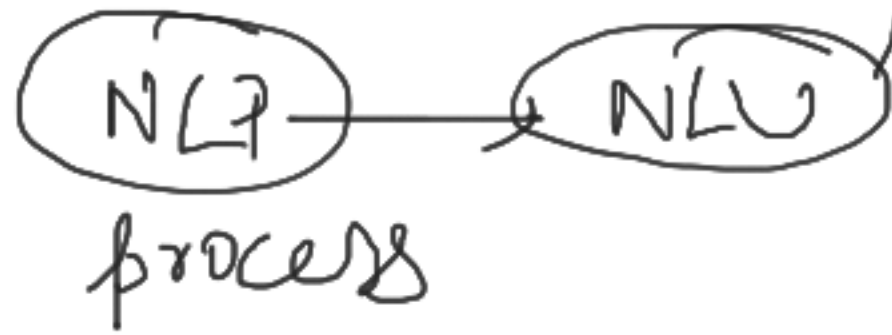


Review (NLP)



words / token

text → numeric
↓
⇒ count
⇒ tfidf
⇒ encoding
DTM

→ Topic modelling (unsupervised)

{ clustering diff articles }
topic and then
one topic

google news, NIS

↳ global sports

Elon ✓
Rupesh ✓
Google ✓
Nifty ✓

Business

~~name of topic~~

- Intuitive way

↓
- what is happen
↓
- what is outcome
↓
code

NLU

TOC

- Algebra
- Trig
- calculus
 - dif
 - integ
- Stats

A/M/S/D
Matrix

mimic today

machine

Deepali are

Unigram, bigram, trigram, n-gram

feature
+ multi

DTM

I am

I	am	tan	is
✓	✓	✓	✓

token independent

I am

→ Tantrique is am

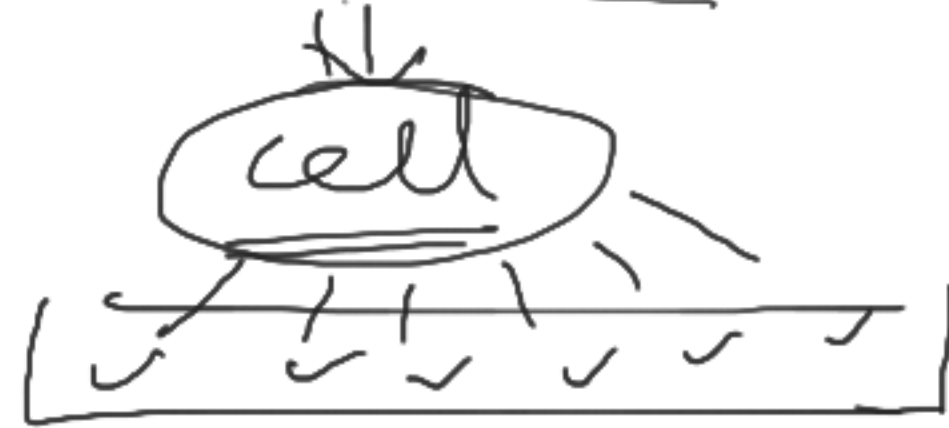
I am	Tantrique is	Amant	Deepali
------	--------------	-------	---------

Deepali

Amant & Deepali are studying W & F

Usage of Topic Modelling

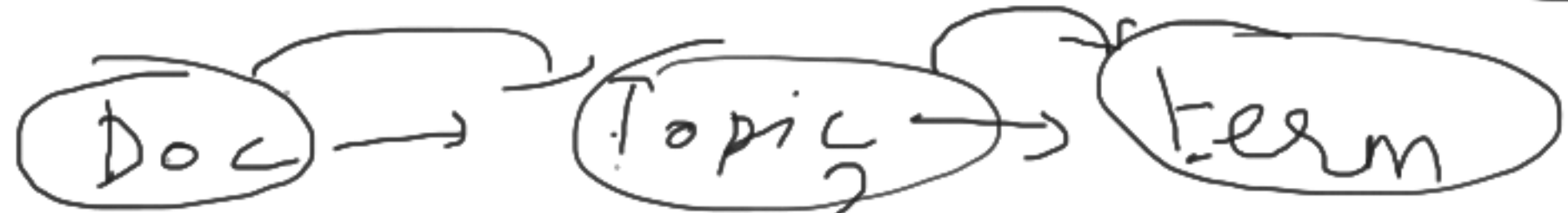
1 cells \rightarrow 1 tissues \rightarrow 1 organ \rightarrow human body



1 words \rightarrow 1 sentence \rightarrow 1 para \rightarrow 1 doc

topic modelling

Assumption \rightarrow start



1 topic

DTM

1 doc

n topic \rightarrow 1 term

1 doc \rightarrow 1 topic \rightarrow 1 term

find out

~~topic model~~

1 doc \rightarrow 1 topic

\downarrow
n terms

doc \rightarrow mixture topic

Doc \rightarrow topic \rightarrow words

topic \rightarrow mixture terms } topic modelling

assump
1 do

Souren Ganguly & Greg Chappell

\rightarrow Sports, Controversy

SG
GC

\rightarrow SG + GC

~~Latent~~
topic

Dirichlet

Allocation

DTM

Latent

Doc \rightarrow topic \rightarrow words

100°C \rightarrow 100°C

Maths Diric \rightarrow intuition

doc \rightarrow topic \rightarrow words



base

Foot

control

LHS \rightarrow RHS

1 doc \rightarrow mul topi

1 topic \rightarrow n words



topics \rightarrow edge of shape

words



LDA

T₁

doc

DD

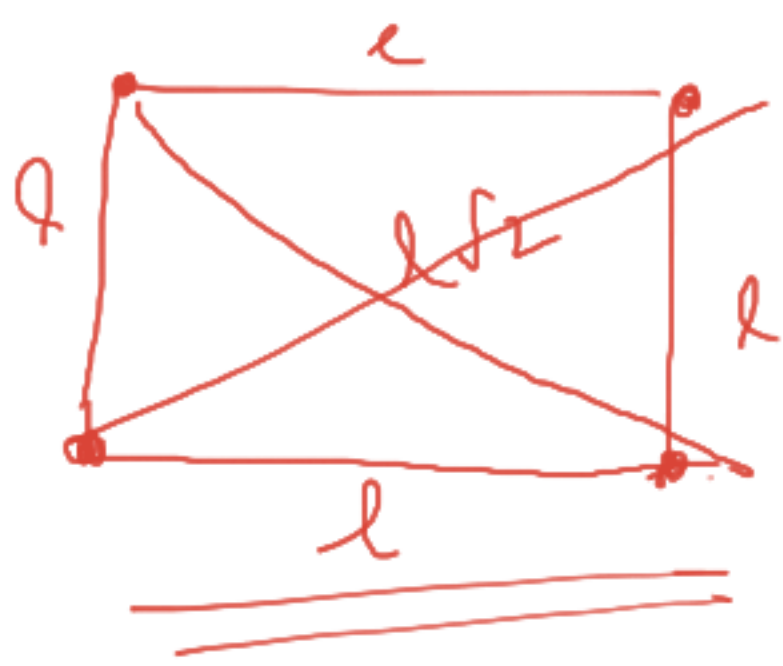
T₂

song

T₃

drink

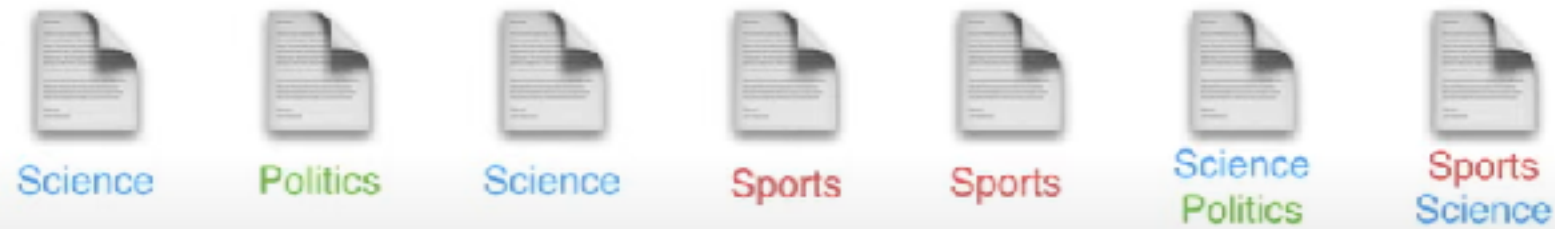
T_i equally distant



$$3 \text{ length} = l$$

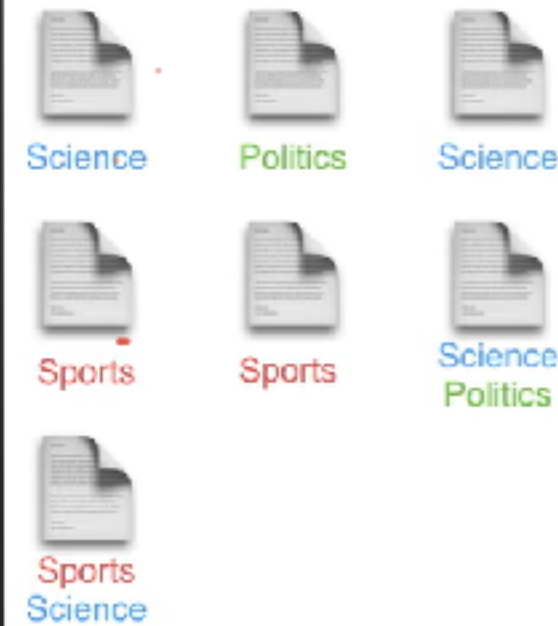
The problem

Goal

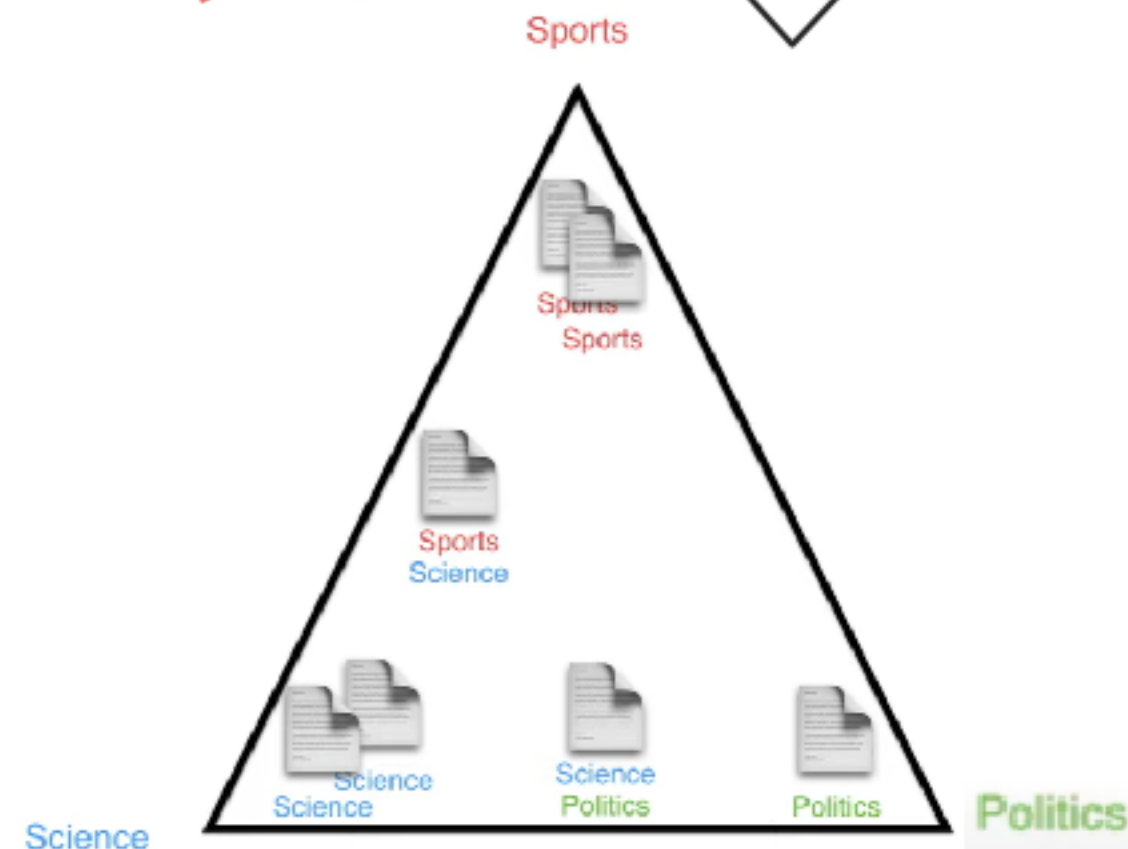
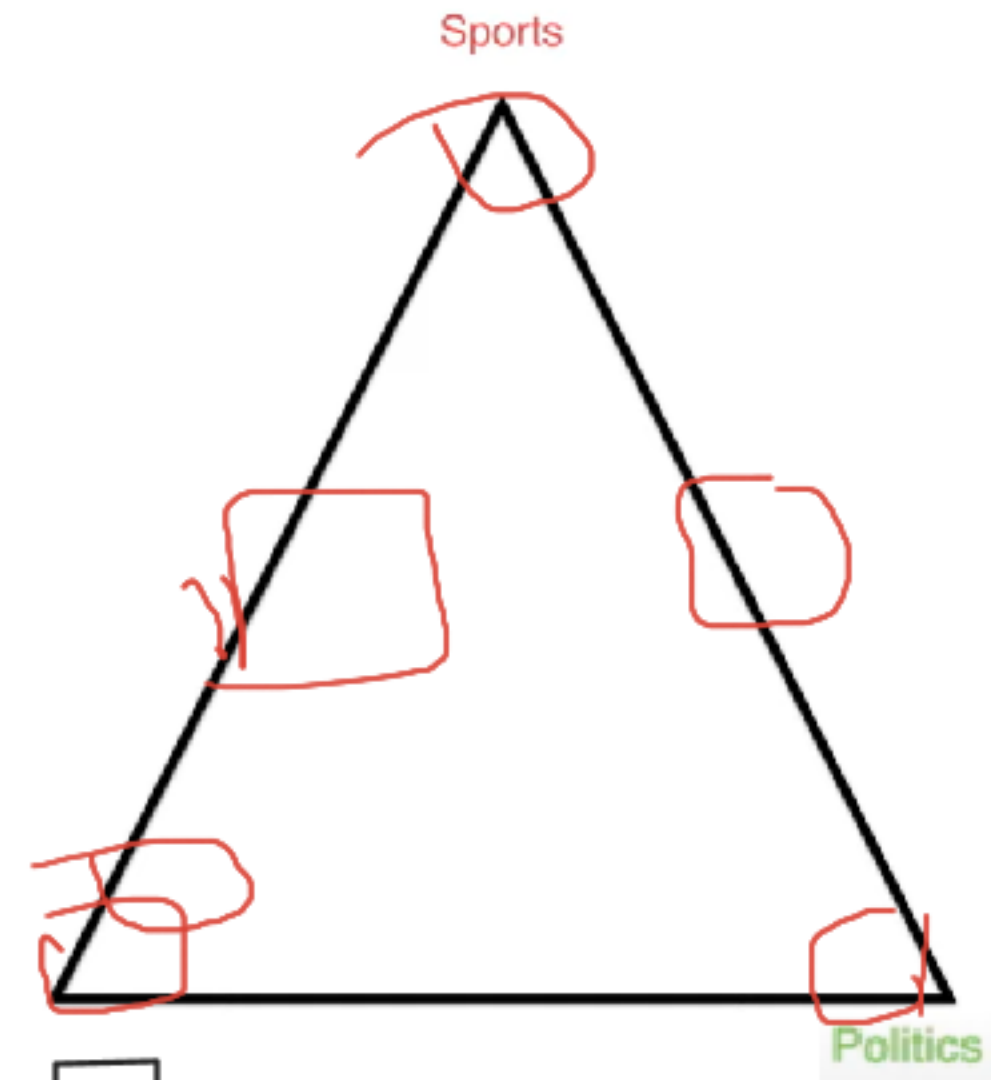


Sports
Science

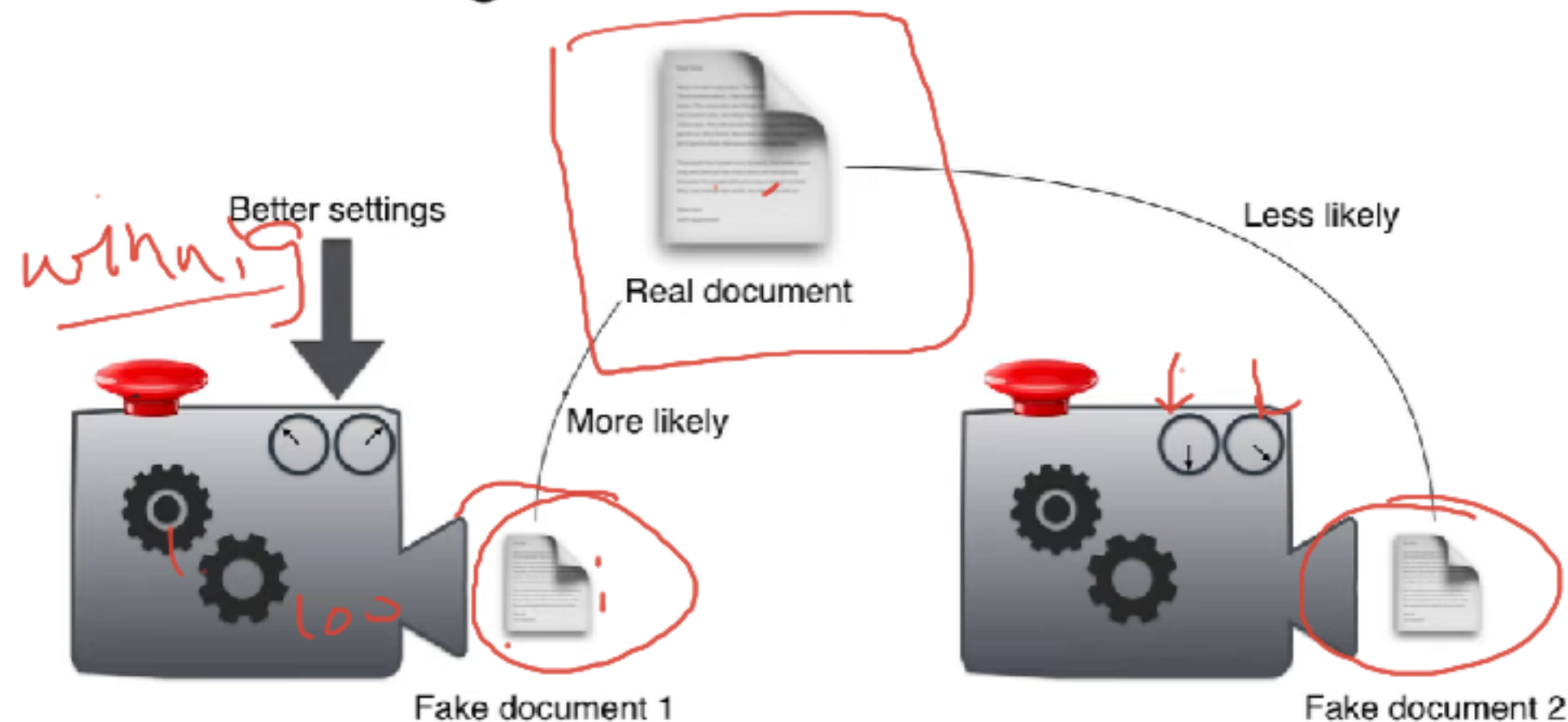
LDA



know



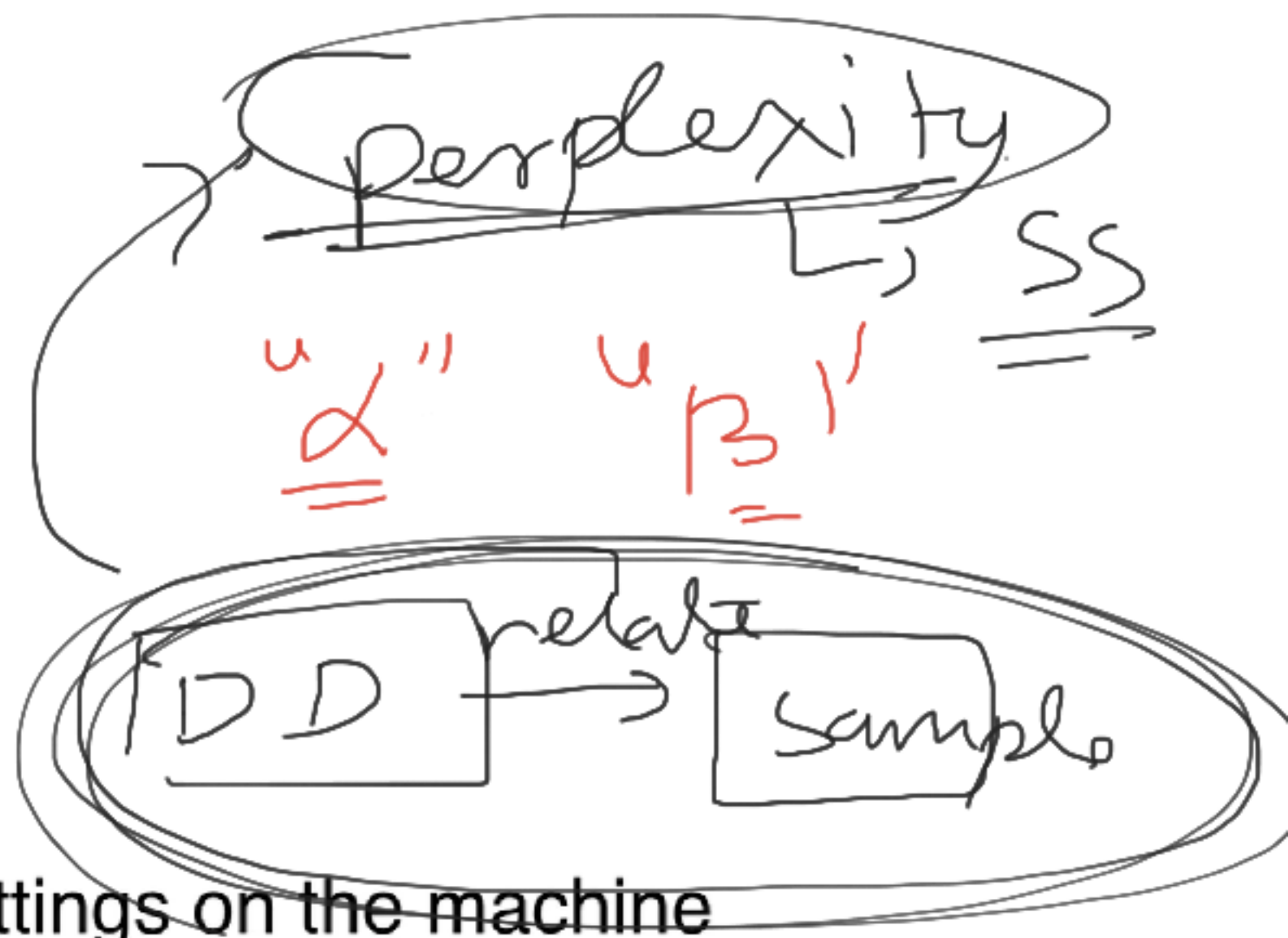
Best settings on the machine



gibbers

$D_1 = 100 \text{ words}$

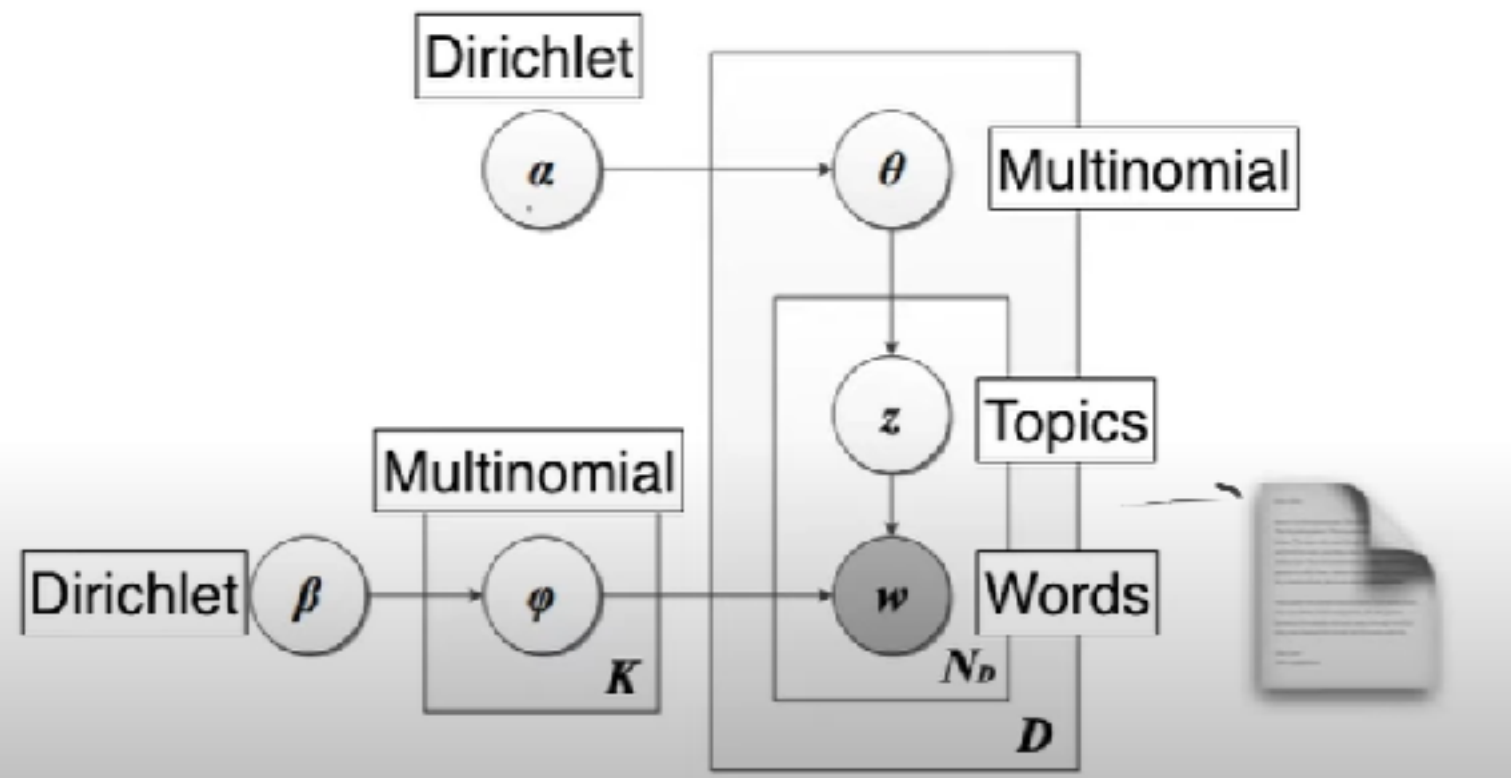
(LDA) - module do?



Best settings on the machine

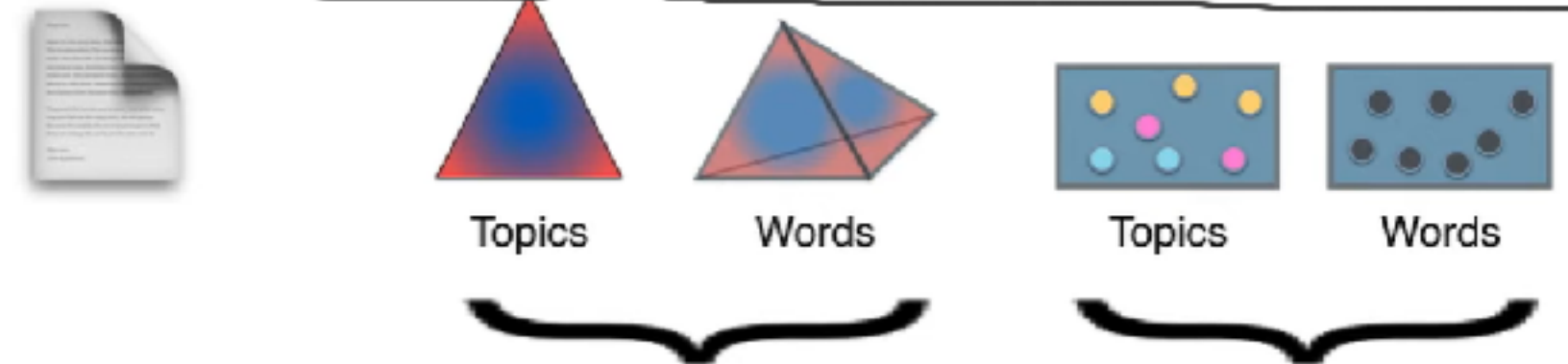


Blueprint for the LDA machine



Probability of a document

$$P(\mathbf{W}, \mathbf{Z}, \boldsymbol{\theta}, \boldsymbol{\varphi}; \alpha, \beta) = \prod_{j=1}^M P(\theta_j; \alpha) \prod_{i=1}^K P(\varphi_i; \beta) \prod_{t=1}^N P(Z_{j,t} | \theta_j) P(W_{j,t} | \varphi_{Z_{j,t}})$$

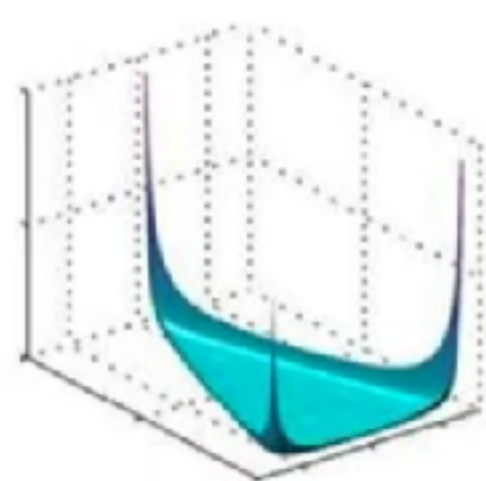


Steps Intuition-2

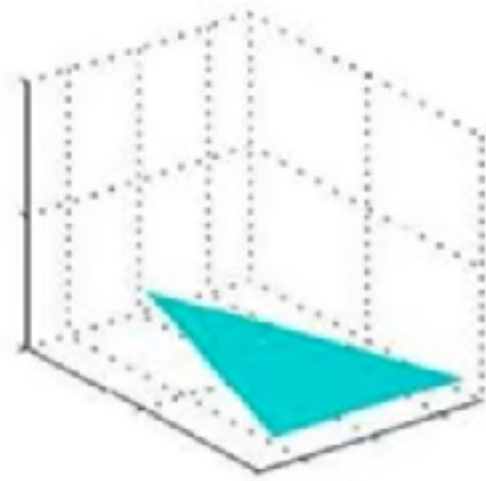
(0, '0.024*"ban" + 0.017*"order" + 0.015*"refugee" + 0.015*"law" + 0.013*"trump"
' + 0.011*"kill" + 0.011*"country" + 0.010*"attack" + 0.009*"state" + '
'0.009*"immigration"') topic \rightarrow RUW topic 1
(1, '0.020*"student" + 0.020*"work" + 0.019*"great" + 0.017*"learn" + '
'0.017*"school" + 0.015*"talk" + 0.014*"support" + 0.012*"community" + '
'0.010*"share" + 0.009*"event"') topic \rightarrow RUW

Dirichlet Distributions

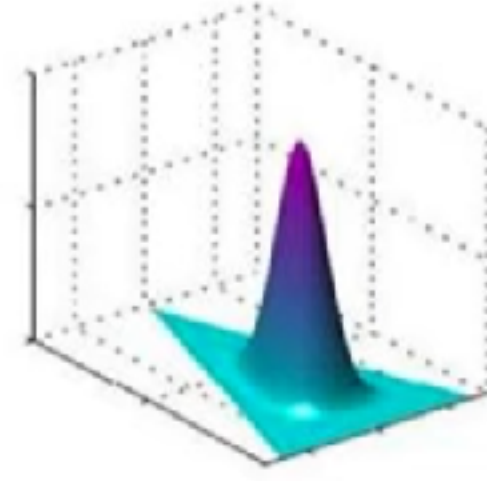
$$f(x_1, \dots, x_K; \alpha_1, \dots, \alpha_K) = \frac{1}{B(\alpha)} \prod_{i=1}^K x_i^{\alpha_i-1}$$



0.7, 0.7, 0.7



1, 1, 1



5, 5, 5

~~in.vat~~ {index, () }²

For ~~col~~ is col-names

sum & np.sum(, axis

sum (is), =

Euclidean $SS \rightarrow PD$ $P \rightarrow PD$ unsupervised

SS

Perplexity

