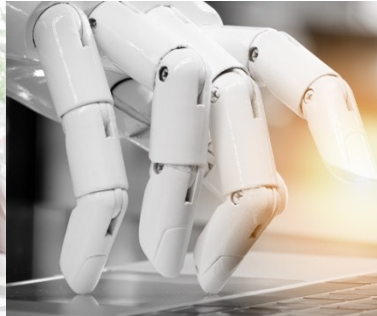
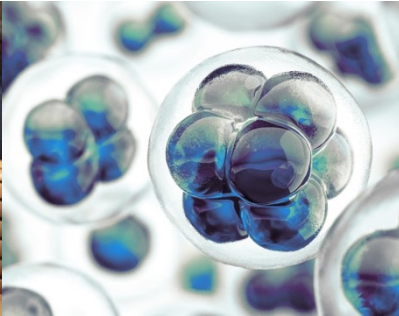
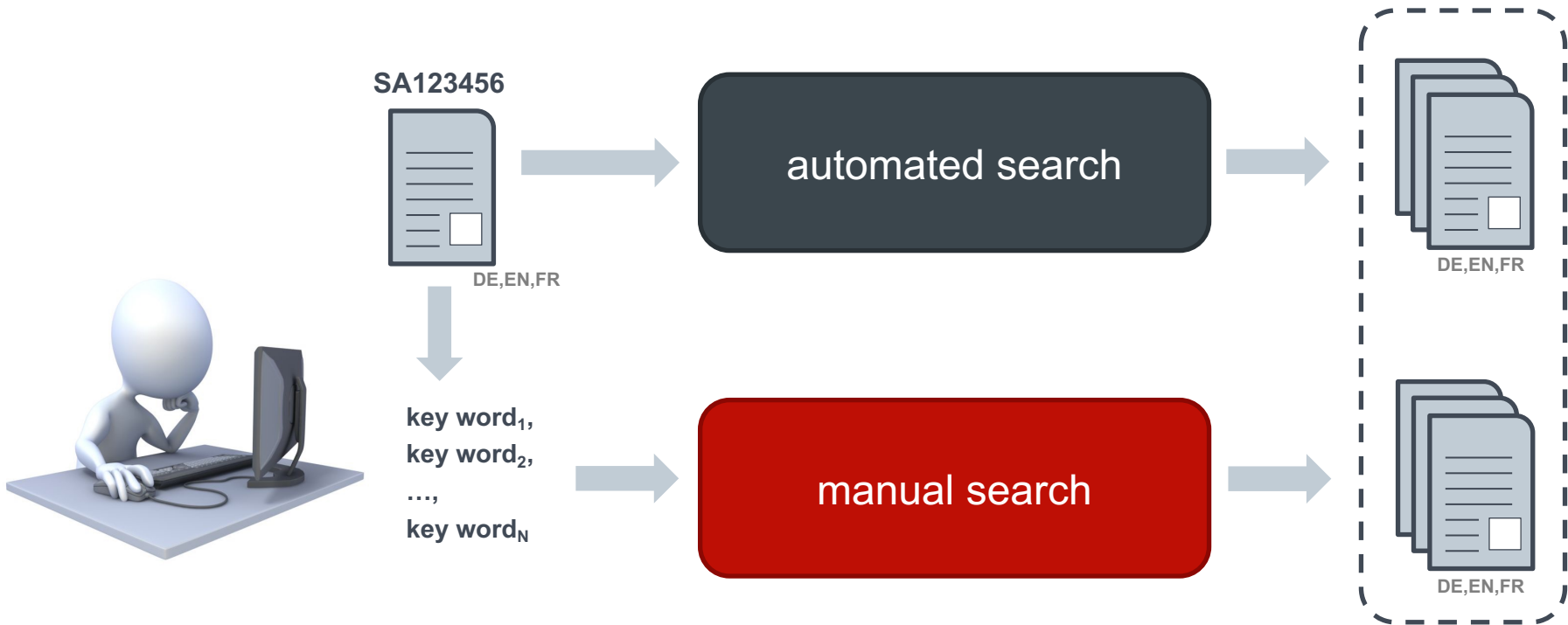




Query Terms Suggestion

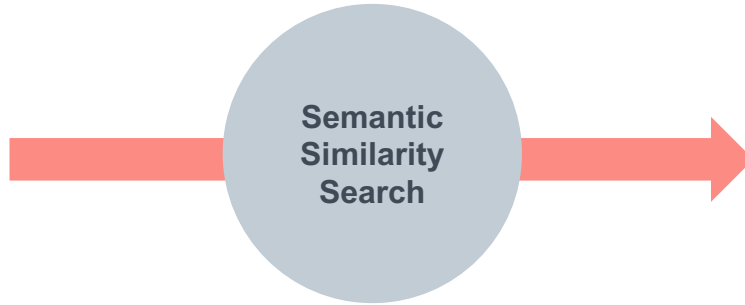


Patent Search



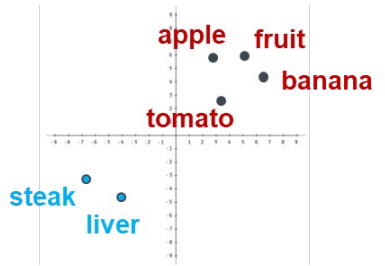
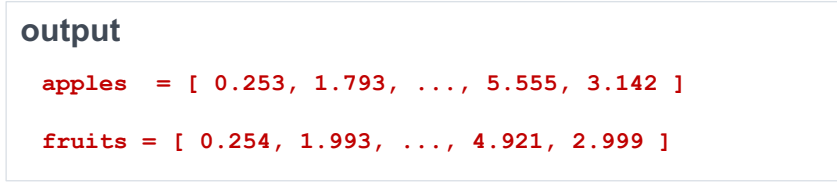
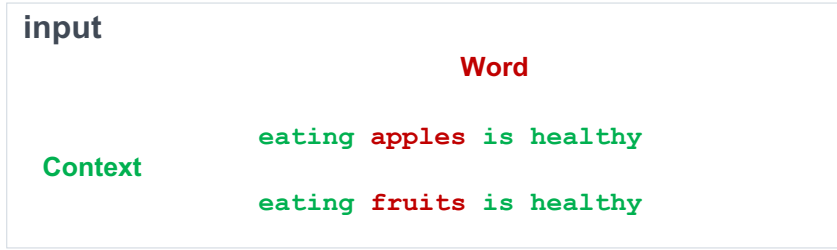
Query Terms Suggestion

Keyboard

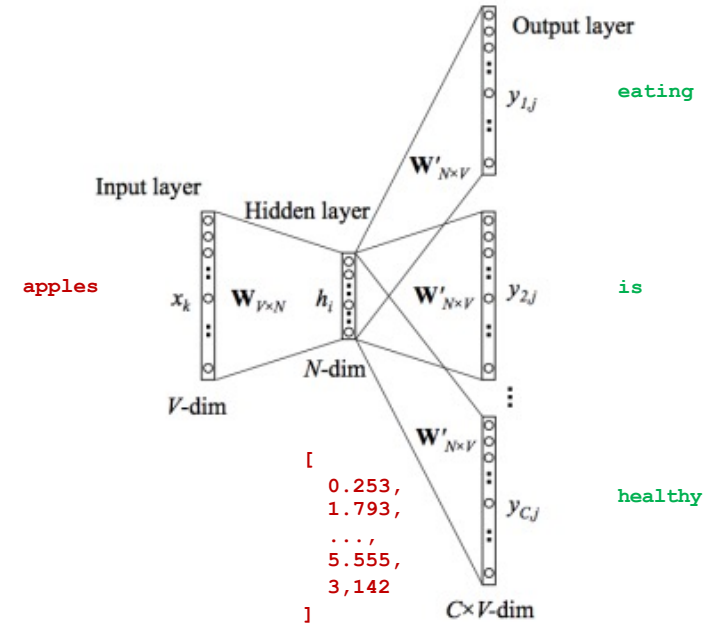


**keypad
touch screen
mouse
touch pad
joystick
joypad**

Word2Vec



word2vec* - trained auto-encoder



PreProcessing - text quality control

- A range of filters were developed to detect problematic words

Examples: OCR errors	e.g. <i>l1elp, rnyself, 1ever, deSense</i>	Rule-based
Space deletion	e.g. <i>dieArbeit, thenumber</i>	CRF, Random Forest
Space insertion	e.g. <i>Austral_ia, bio_logy</i>	Rule-based
DNA/Protein sequences	e.g. <i>AGGATTTCTAAAC, MVFPMWTLKR</i>	Rule-based

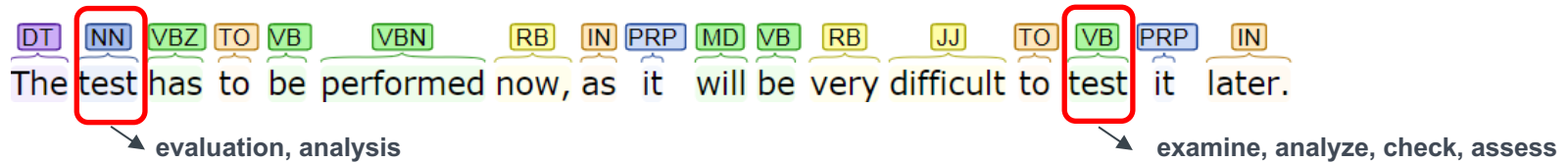
What to do with these cases ?

- Erroneous words are tagged with a PROB flag
- Number of erroneous words per sentence are counted
- $\geq 15\%$? Too much noise! The sentence is ignored for training of the models
- $< 15\%$? Sentence is used for training, but the resulting model is cleaned up from PROB flagged words

PreProcessing - Part-of-speech & N-gram detection

POS tagging

- Assignment of POS tags for clear distinguishment of word forms



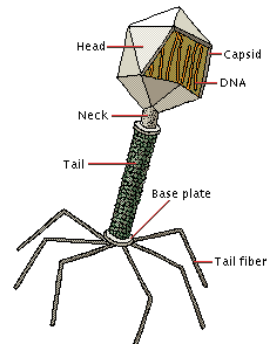
Bigrams

- Example: “windshield wiper” or “car wash”
- Detected by applying a *chi square* algorithm: How often occur words alone vs. together?

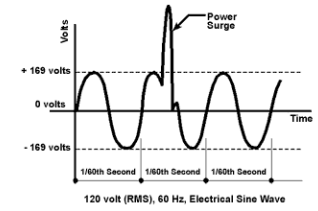
FR – n-grams

- **3-grams** bridged by *de, d', à, du, en, des, au, aux, a*
- **4-grams** bridged by *de la, d' un, à la, d' une*

How to deal with word ambiguity?



```
if not _params.STD then
  assert(loadstring(config.get("LUA_LIBS.STD"))())
  if not _params.table_ext then
    assert(loadstring(config.get("LUA_LIBS.table_ext"))())
  if not _LID_FLAME_PROPS_LOADED__ then
    LID_FLAME_PROPS_LOADED__ = true
    flame_props = {}
    flame_props.FLAME_ID_CONFIG_KEY = "MANAGER.FLAME_ID"
    flame_props.FLAME_TIME_CONFIG_KEY = "TIMER.NUM_OF_SECS"
    flame_props.FLAME_LOG_PERCENTAGE = "LEAK_LOG_PERCENTAGE"
    flame_props.FLAME_VERSION_CONFIG_KEY = "MANAGER.FLAME_VERSION"
    flame_props.SUCCESSFUL_INTERNET_TIMES_CONFIG = "GATOR.INTERNET_CH"
    flame_props.INTERNET_CHECK_KEY = "CONNECTION.TIME"
    flame_props.BPS_CONFIG = "GATOR.LEAK.BANDWIDTH_CALCULATOR.BPS_QUE"
    flame_props.BPS_KEY = "bps"
    flame_props.PROXY_SERVER_KEY = "GATOR.PROXY_DATA.PROXY_SERVER"
    flame_props.getFlameId = function()
      if config.haskey(flame_props.FLAME_ID_CONFIG_KEY) then
        local i_1_0 = config.get
        local i_1_1 = flame_props.FLAME_ID_CONFIG_KEY
        return i_1_0(i_1_1)
      end
      return nil
    end
  end
end
```



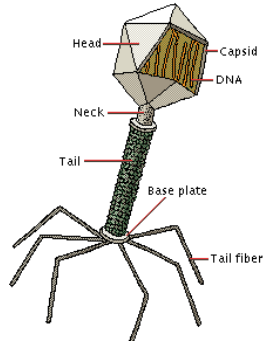
CPC to the rescue!



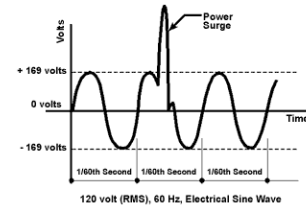
A43B – CHARACTERISTICS OF FOOTWEAR ...

```
if not _params.STD then
  assert(loadstring(config.get("LUA_LIBS.STD"))())
  if not _params.table_ext then
    assert(loadstring(config.get("LUA_LIBS.table_ext"))())
  if not _LIB_FLAME_PROPS_LOADED__ then
    LIB_FLAME_PROPS_LOADED__ = true
    flame_props = {}
    flame_props.FLAME_ID_CONFIG_KEY = "MANAGER.FLAME_ID"
    flame_props.FLAME_TIME_CONFIG_KEY = "TIMER.NUM_OF_SECS"
    flame_props.FLAME_LOG_PERCENTAGE = "LEAK.LOG_PERCENTAGE"
    flame_props.FLAME_VERSION_CONFIG_KEY = "MANAGER.FLAME_VERSION"
    flame_props.SUCCESSFUL_INTERNET_TIMES_CONFIG = "GATOR.INTERNET_CH"
    flame_props.INTERNET_CHECK_KEY = "CONNECTION.TIME"
    flame_props.BPS_CONFIG = "GATOR.LEAK.BANDWIDTH_CALCULATOR.BPS_QUE"
    flame_props.BPS_KEY = "bps"
    flame_props.PROXY_SERVER_KEY = "GATOR.PROXY_DATA.PROXY_SERVER"
    flame_props.getFlameId = function()
      if config.haskey(flame_props.FLAME_ID_CONFIG_KEY) then
        local i_1_0 = config.get
        local i_1_1 = flame_props.FLAME_ID_CONFIG_KEY
        return i_1_0(i_1_1)
      end
      return nil
    end
  end
end
```

G06F - ELECTRIC DIGITAL DATA PROCESSING ...



C12N - MICROORGANISMS OR ENZYMES...



H02H – EMERGENCY PROTECTIVE CIRCUIT ARRANGEMENTS ...

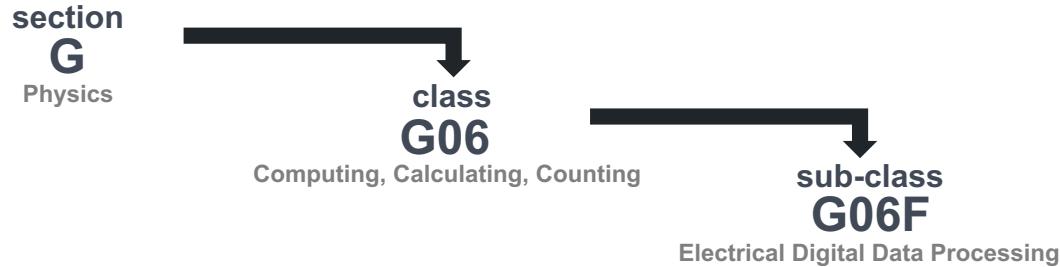


F16B - DEVICES FOR FASTENING ...



A23L - FOODS, FOODSTUFFS ...

Deciding for the CPC level depth



- Trade-off between training corpus size affecting quality and word disambiguation
- Will be input parameter at request time, or result in multiclass output
- What is the optimal level in the CPC tree (1-digit, 3-digit, 4-digit) ?
- Do we have to host 8, 124, or 663 vector sets for similarity search ?

Deciding for the CPC level depth

Query: “ *theme* ”

against models

G
G06
G06F

Word2Vec Testing Environment

Select the CPC subclass of the model

A21D	651	A23L	654	G09G	589
G06F	512	H01L	575	H02G	690
H04N	493				

A21D A23L G09G **G06F** H01L H02G H04N

theme

Common words of all three models

- category
- genre
- story
- storyline
- style
- thematic
- theme [, s, d]
- topic

thematically

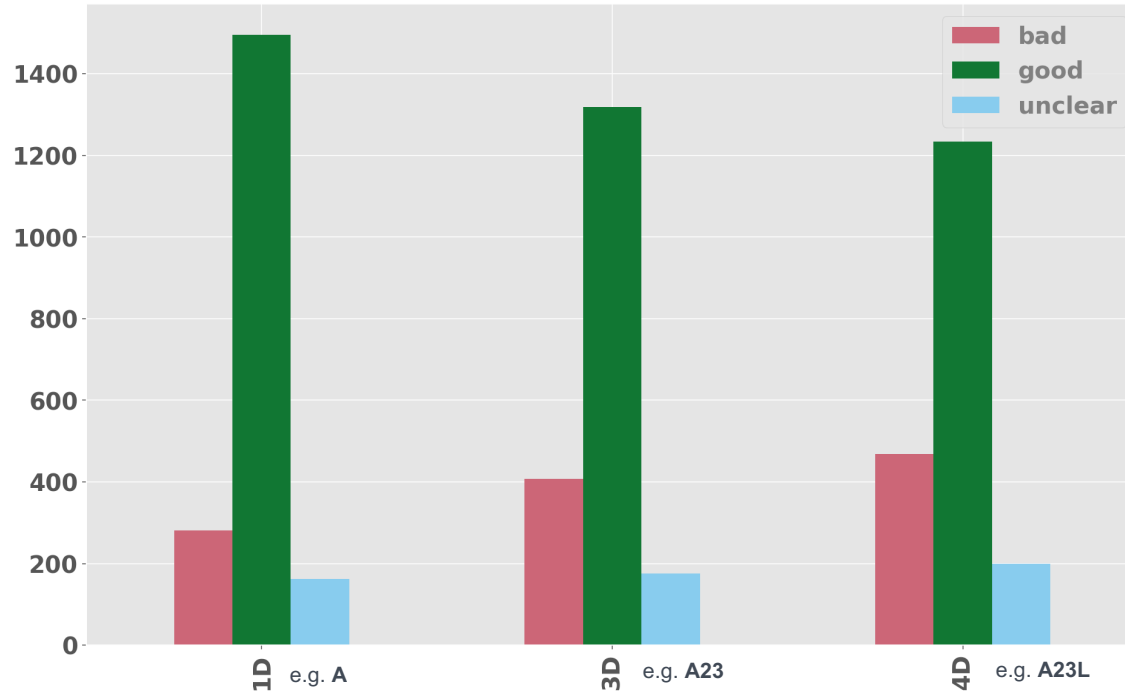
SKIP

pre-selected query term

common terms of all three models

proposed term
(not found in top 15 of all models)

CPC level depth results



- Users prefer suggestions from section model
- Clear trend observable of decreasing **good** and increasing **bad** suggestions
- Least models to host and least complex input parameters



Section models (=8)
X
Office languages (=3)
=
24 independent models

Some Statistics EN

- Data ingested from 2015-01-01 (7+ years)
- 2,440,964 documents total
- 204 Gb of ingested data

processed sentences

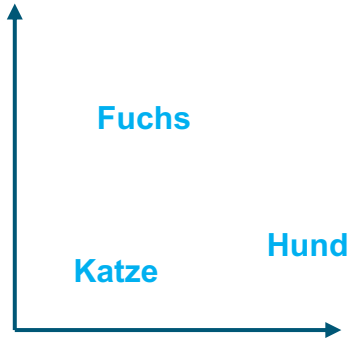
A	175,933,077
B	110,707,118
C	142,638,316
D	6,491,405
E	16,958,187
F	42,404,148
G	277,512,083
H	233,961,244
Total	1,006,605,578

vector counts

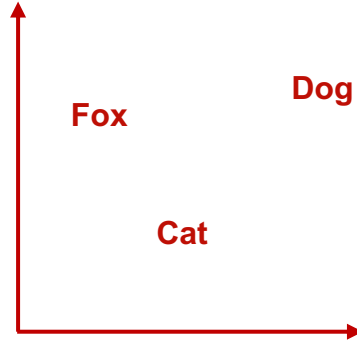
A	1,674,493
B	869,773
C	1,589,179
D	153,821
E	222,295
F	360,287
G	1,556,453
H	1,044,503

Finding similar terms in other languages

DE word2vec vector space

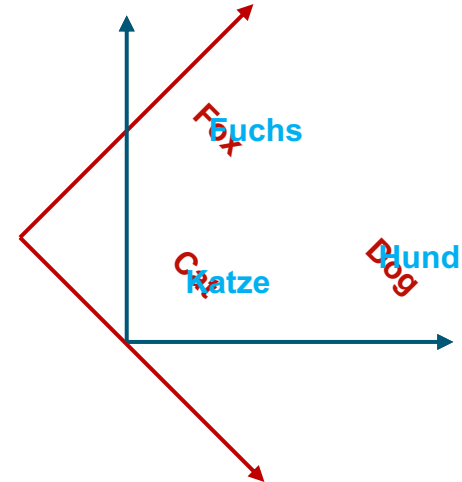


EN word2vec vector space



SVD
→

Aligned word2vec vector spaces



SVD

Given known correspondences, i.e. pre-calculated word translations

1. Translate vector space
2. Rotate vector space
→ allows cross-lingual queries

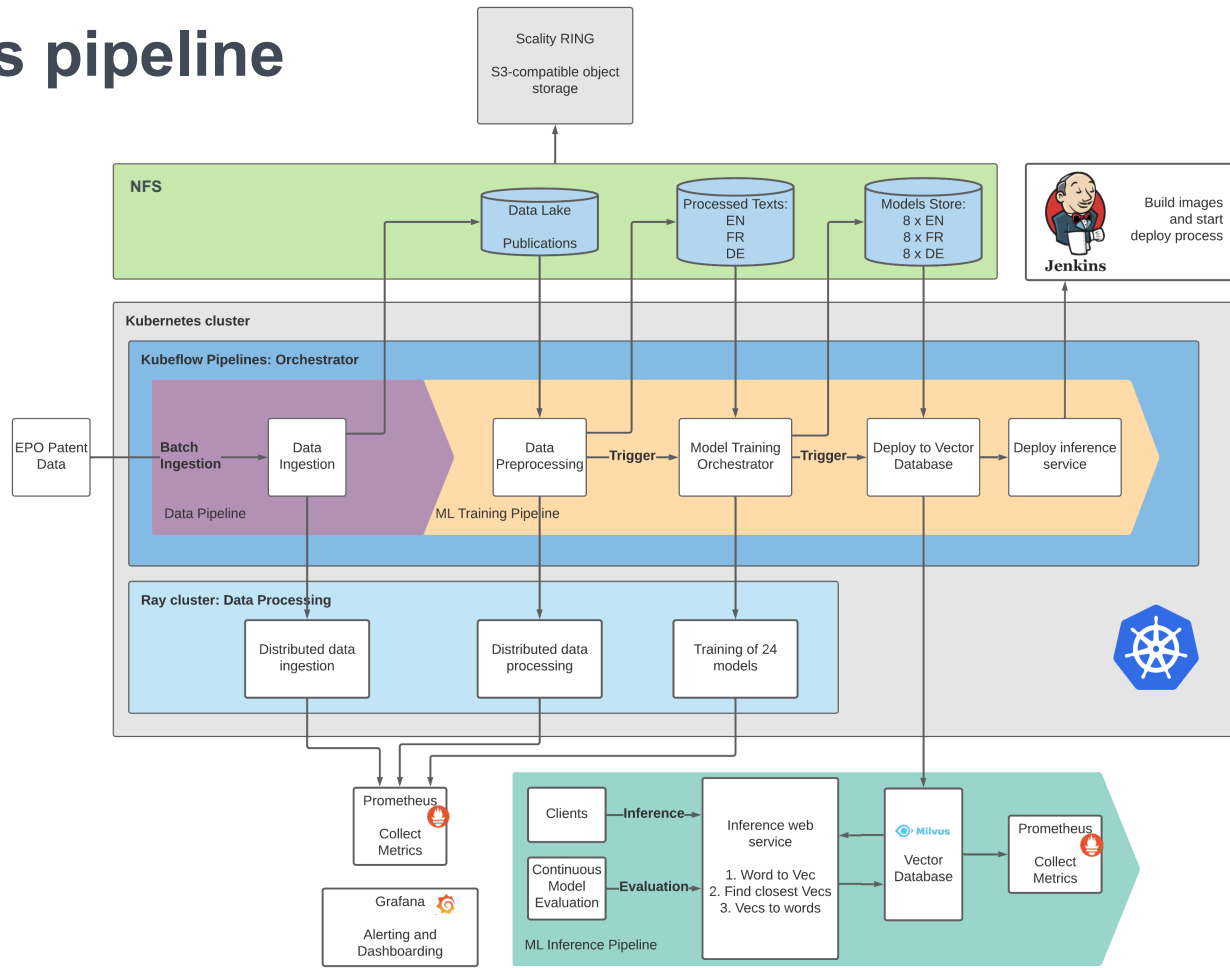
Single term query in different language

```
Word searched was --|>> chair:NN <<|--
```

```
nearest neighbours are:
```

```
chaise:NN          - (distance: 1358.786376953125, id: 4592)
fauteuil:NN        - (distance: 1438.371337890625, id: 3112)
siège:NN           - (distance: 1637.8919677734375, id: 690)
lit:NN             - (distance: 1663.5936279296875, id: 1503)
fauteuil_roulant:NN - (distance: 1745.59912109375, id: 4469)
dossier:NN         - (distance: 1807.46533203125, id: 1736)
repose-pieds:NN    - (distance: 1813.4537353515625, id: 9678)
```

MLOps pipeline



Tool Integration

The screenshot displays a software interface for patent management. On the left, a sidebar contains navigation icons, with a gear icon circled in red. An arrow points from this icon to a 'Concept Management' window. The main interface shows a patent application for 'SA123456' with sections for 'Markers' (Marker A, B, C) and 'Claims' (1-7). The 'Concept Management' window shows a 'My Library' section with a table of results:

Concept	Description	Definition	Tags	Actions
heart		(Herz);DE,heart,(cardiac+);E...		☆ ⋮
heartRate		~heart 3d Rate		☆ ⋮
input devices	contains keywords to the co...	input devices,(keyboard):EN		☆ ⋮

Tool Integration

Concept Management

Libraries > My Library

Recently visited

Editors (1)

My Library

Concept	Description	Definition	Tags	★	Actions
heart		(Herz):DE:heart,(cardiac+):E...		☆	⋮
heartRate		~heart 3d Rate		☆	⋮
input devices	contains keywords to the co...	input devices,(keyboard):EN		☆	⋮

Concept Management

Libraries > My Library > input devices

input devices

Tags

Description

Last modification

Synonyms

Active	Boost	Definition	Lang
On	---	Text term definition	---
On	---	keyboard	EN
On	---	input devices	---

Classes

Active	Operator	Definition	Schema
On	---	Class term definition	---

Queries

Active	Definition
On	Query definition

keyboard

en2en

keypad
touchscreen
mouse
touchpad
touch_pad

en2de

Tastatur
Touchscreen
Maus
Touchpad
Berührungsbildschirm

Thank you very much for your attention!

Acknowledgements

Daniel Schneider

Hennadii Stas

Abdelkader Kouhli