

face of  
fonts | 字之魂

face of  
fonts

字  
之  
魂

中国梦



If you want to make a superhero movie,

What font will you choose?

THE AVENGERS

THE AVENGERS

**TRAINING CENTERS**

***THE AVENGERS***



A font can make or  
break your **brand**

字之魂

A typographic designer  
must convey **meaning**  
through visuals

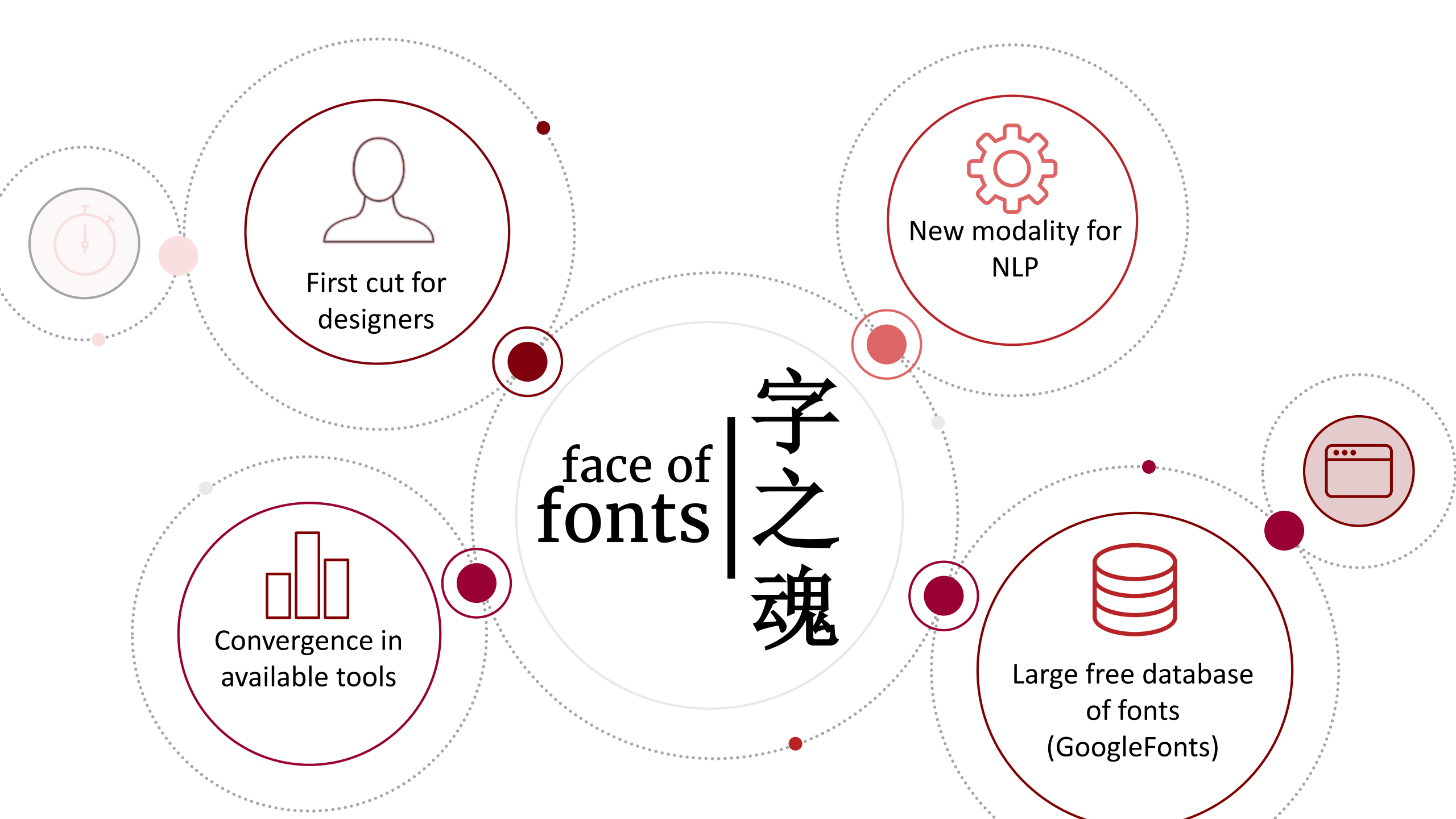
字之魂

image **classification** to fonts  
+  
**CNNs** for NLP

字之魂

**data-driven** approach to  
augment the process

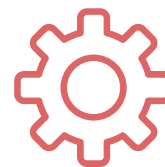
字之魂



face of  
fonts | 字之魂



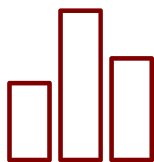
First cut for  
designers



New modality for  
NLP



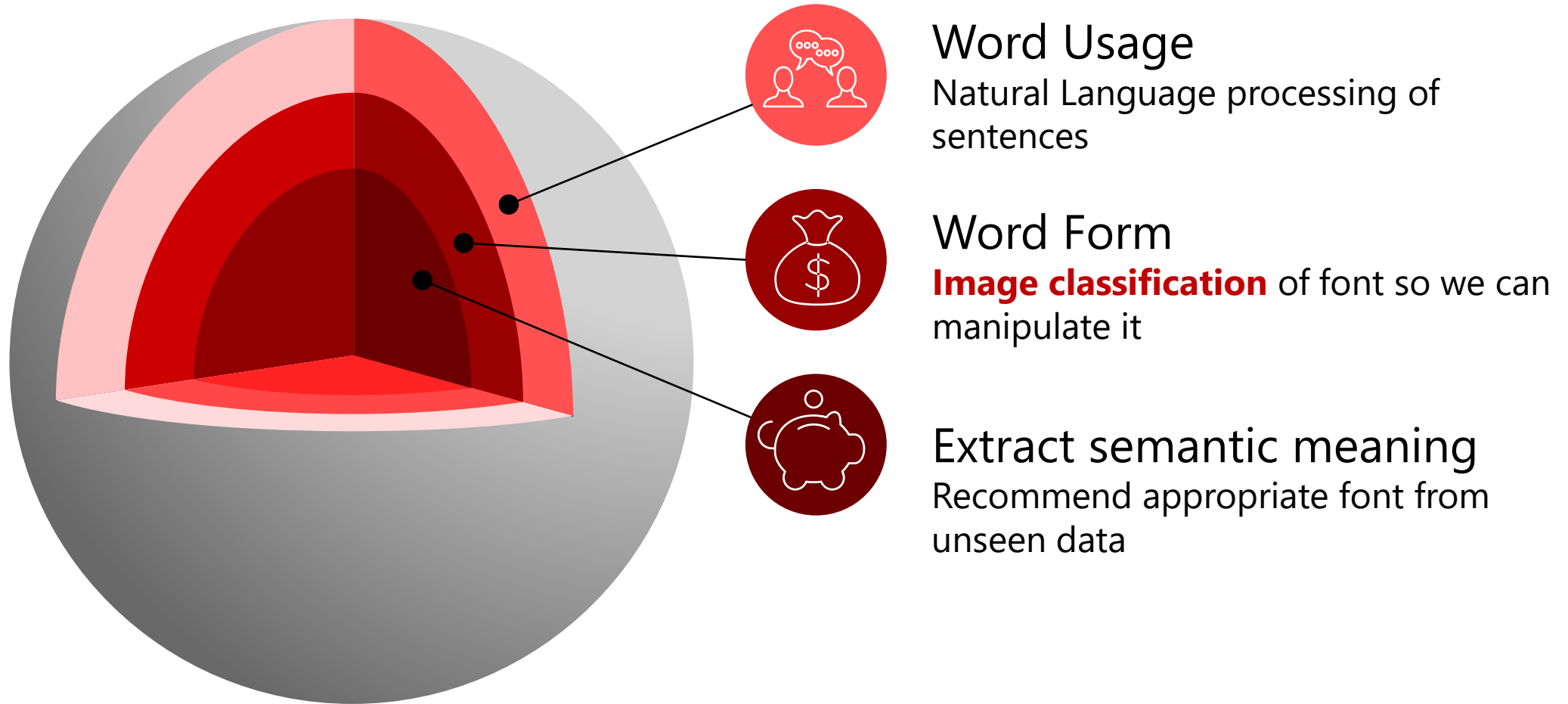
Large free database  
of fonts  
(GoogleFonts)



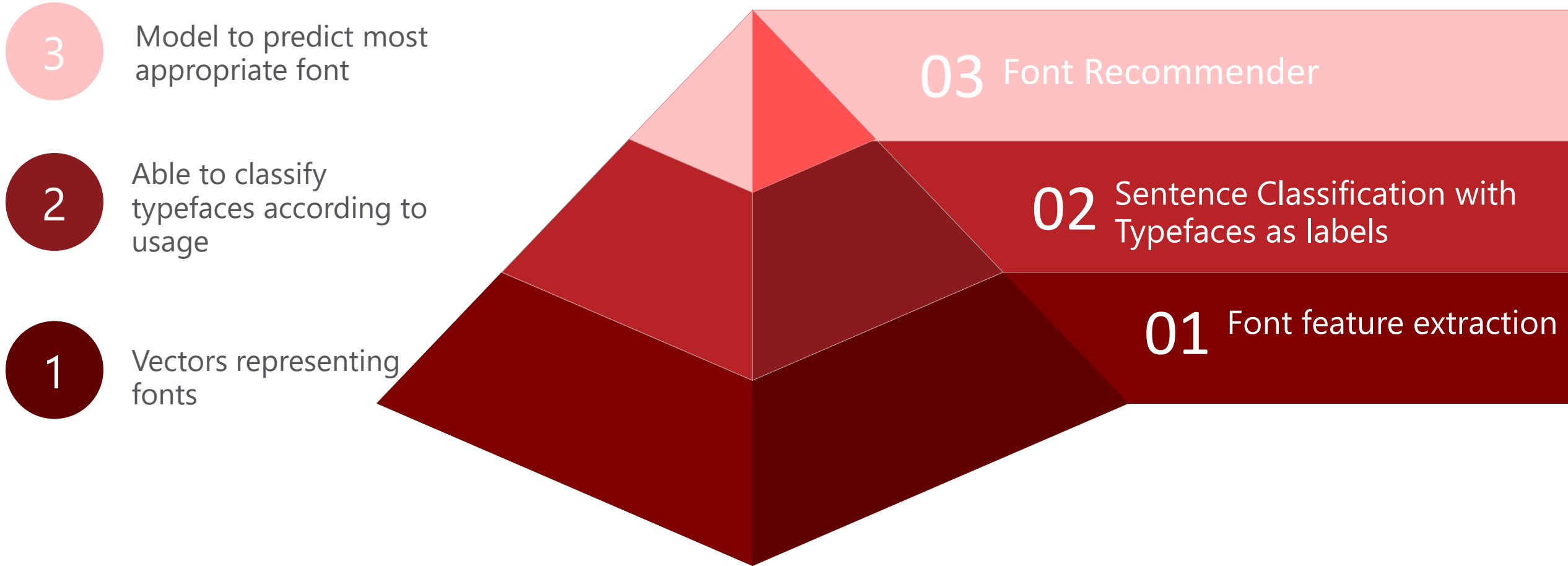
Convergence in  
available tools



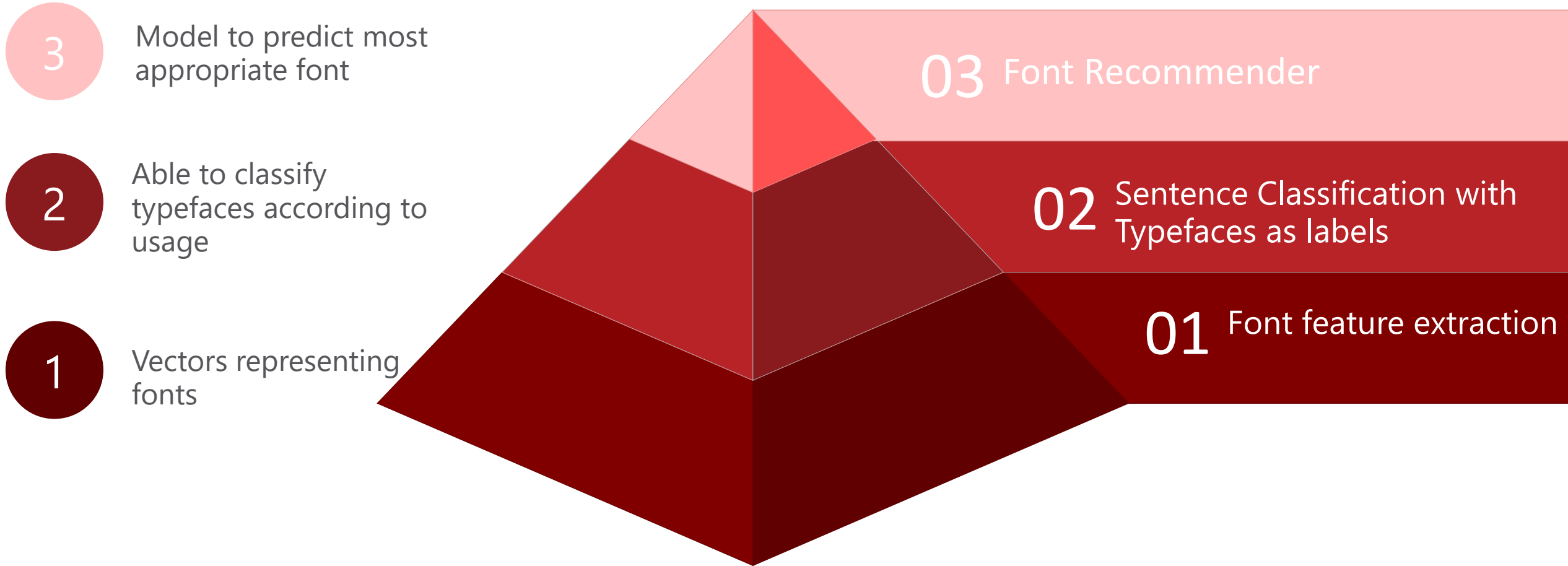
# Objective



# Methodology



# Methodology



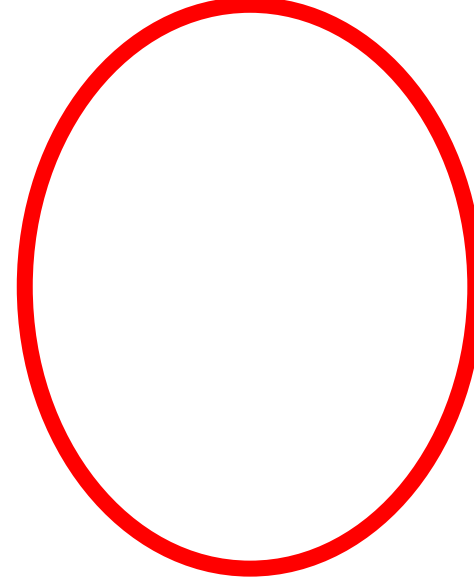


# Fontjoy

**1883 different fonts from GoogleFonts**

**Font image to font vector**

**Allow you to manipulate fonts**



**Can identify clusters of similar fonts**

Qiao,J (2017)

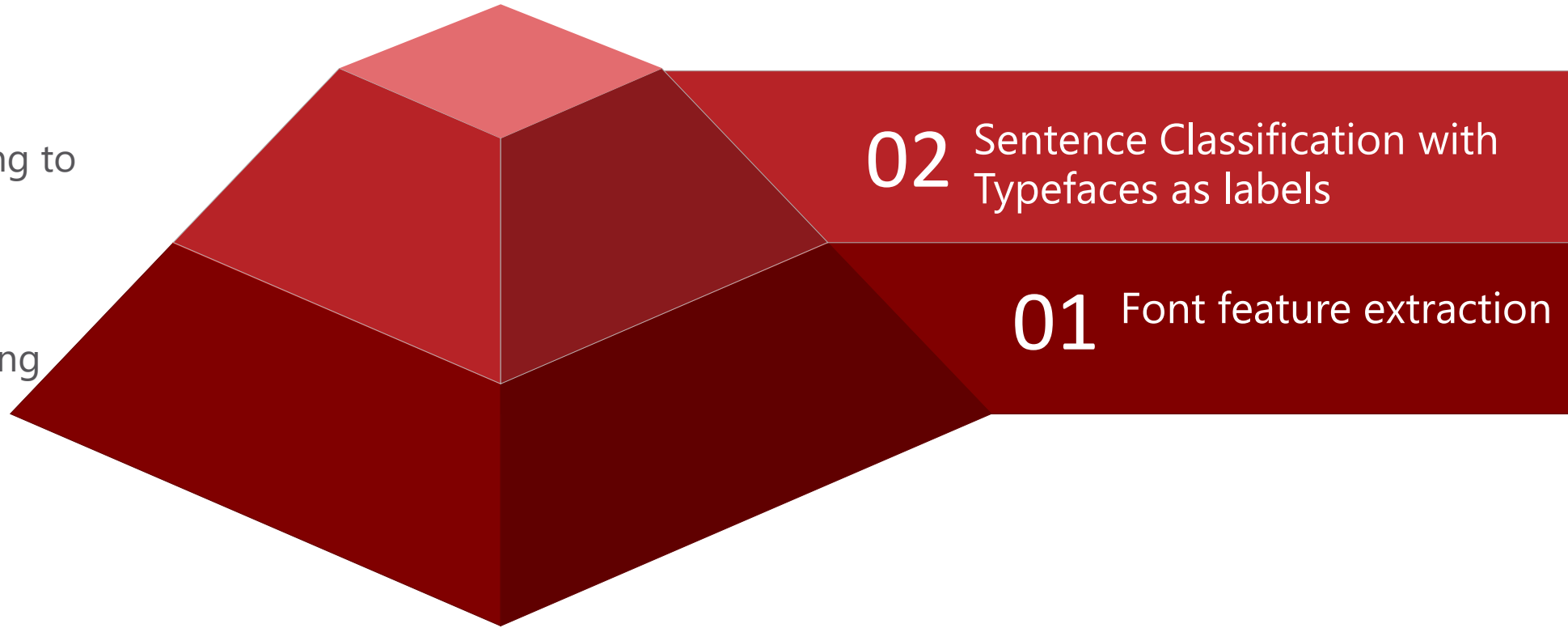
# Methodology

2

Able to classify  
typefaces according to  
usage

1

Vectors representing  
fonts



# workflow

Must be from the  
same source  
Example: **Linked in**

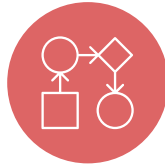
**Dataset**  
Taken from LinkedIn

**Pre-processing**  
Convert dataset into usable format



Natural Language Analysis  
with Python NLTK  
Clean up dataset

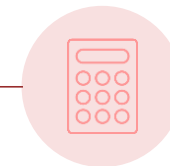
Hypothesis:  
Typography bears  
semantic meaning



**LIBLINEAR**  
Baseline result to test hypothesis

**CNN**

To train a usable model



Proven to be effective  
in sentence  
classification

# workflow

Must be from the  
same source  
Example: **Linked in**

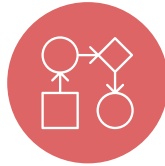
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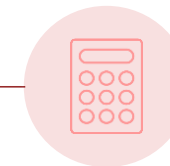
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classification

# workflow

Must be from the  
same source

Example: **Linked** 

**Dataset**  
Taken from LinkedIn

**No available dataset that suited our purpose**

# workflow

Must be from the  
same source

Example: **Linked** 



## Criteria:

### 1. Font from the same domain is used

*Meaning is constrained by the different domains that the typeface is used on*

### 2. Large enough number of instances

*Sufficient amount of data*

### 3. Publicly available corporate branding guides

*Official information provided by companies*

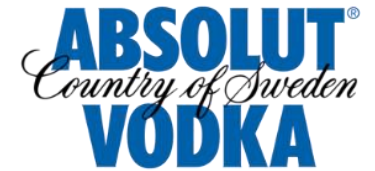
# workflow



Must be from the  
same source

Dataset

Taken from LinkedIn



# Garamond

# Futura



**Universities and government  
agencies**

**Lifestyle brands**

# Pre-processing

Convert dataset into usable format



Natural Language Analysis  
with Python NLTK

P "Provid", "challeng", "reward", "environment", "facilitat", "creat", "think", "teamwork", "open", "communicat"

ons.

Stop-word removal

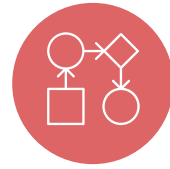
Stemming

Name removal

**Example Sentence**



Hypothesis:  
Typography bears  
semantic meaning



## LIBLINEAR

Baseline result to test hypothesis

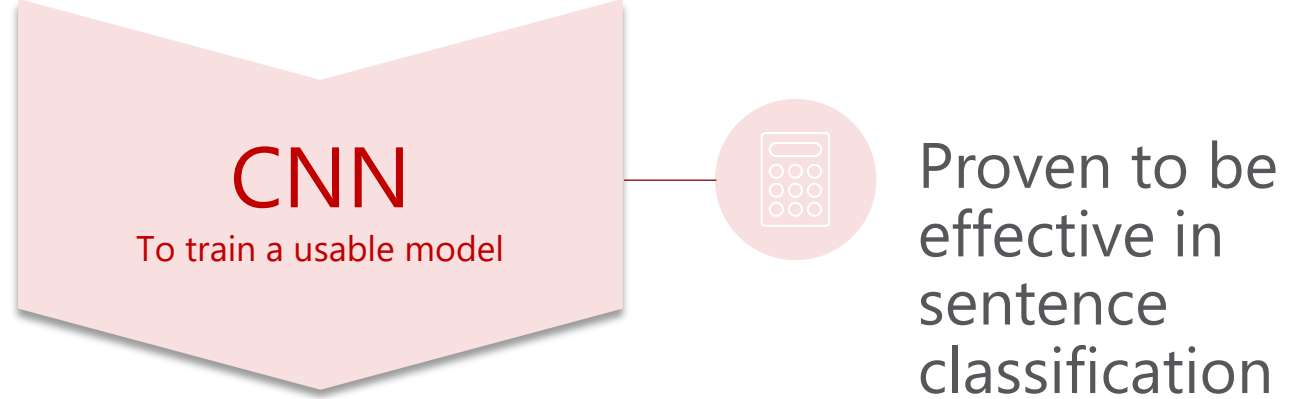
**Chosen due to:**

**1. Fast classification time**

*Designed to be highly efficient*

**2. Able to try multi-class classification**

*Easily tune parameters to provide preliminary result*



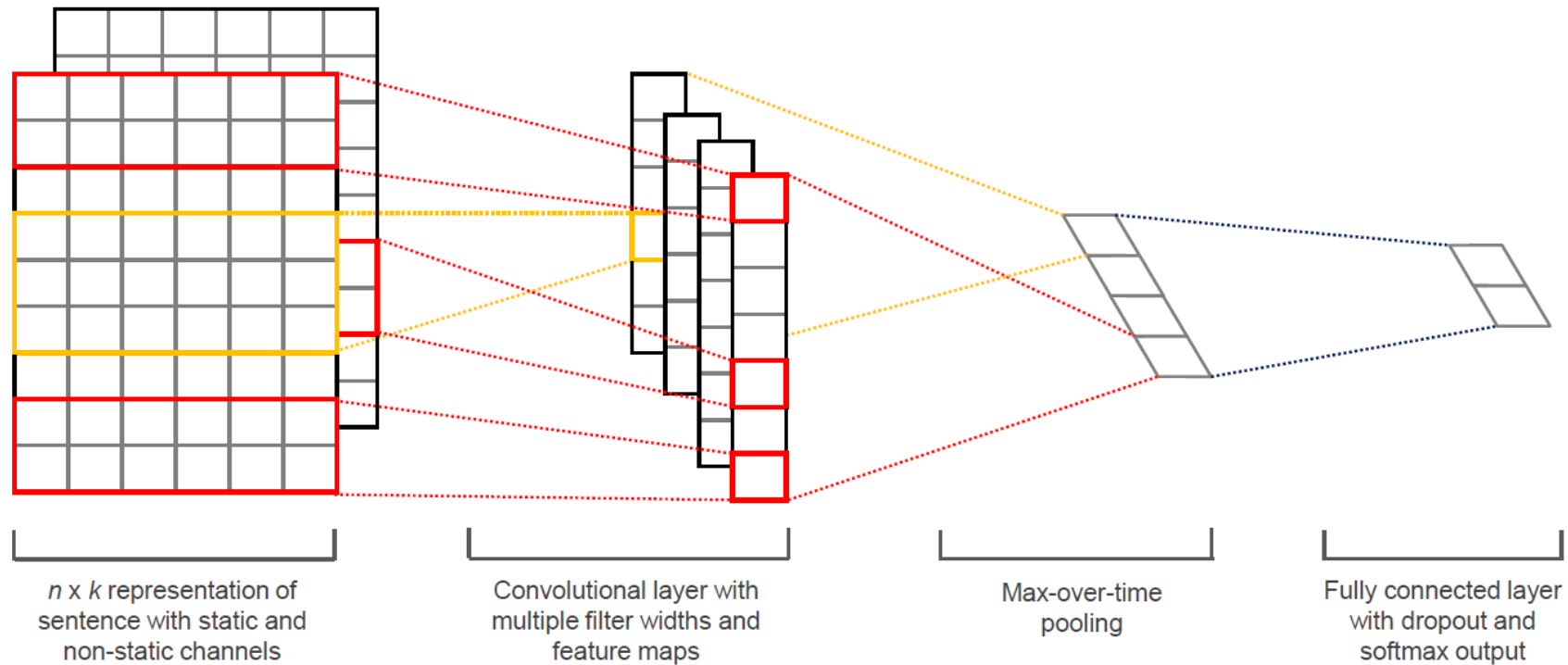
**Utilize the state-of-the-art in sentence classification  
using CNN referencing a practitioner's guide**

# CNN

To train a usable model



Proven to be effective in sentence classification



**CNN model inspired by Yoon Kim's paper**

Y, Kim (2014)

# CNN

To train a usable model



Proven to be effective in sentence classification

Maxlength = 200

nb_filter/ n_gram	1	2	3	4
1000	38.75%	40.53%	39.42%	39.64%
3000	37.86%	39.42 %	37.63%	40.98%
5000	44.58%	45.75%	37.86%	34.96%

**The highest accuracy obtained was 45.75%**

**Therefore we utilized this model to generate font recommendations**

# workflow



Must be from the  
same source

**Dataset**

Taken from LinkedIn

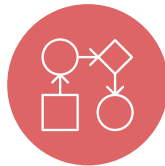
**Pre-processing**

Convert dataset into usable format



Natural Language Analysis  
with Python NLTK  
Clean up dataset

Hypothesis:  
Typography bears  
semantic meaning



**LIBLINEAR**

Baseline result to test hypothesis

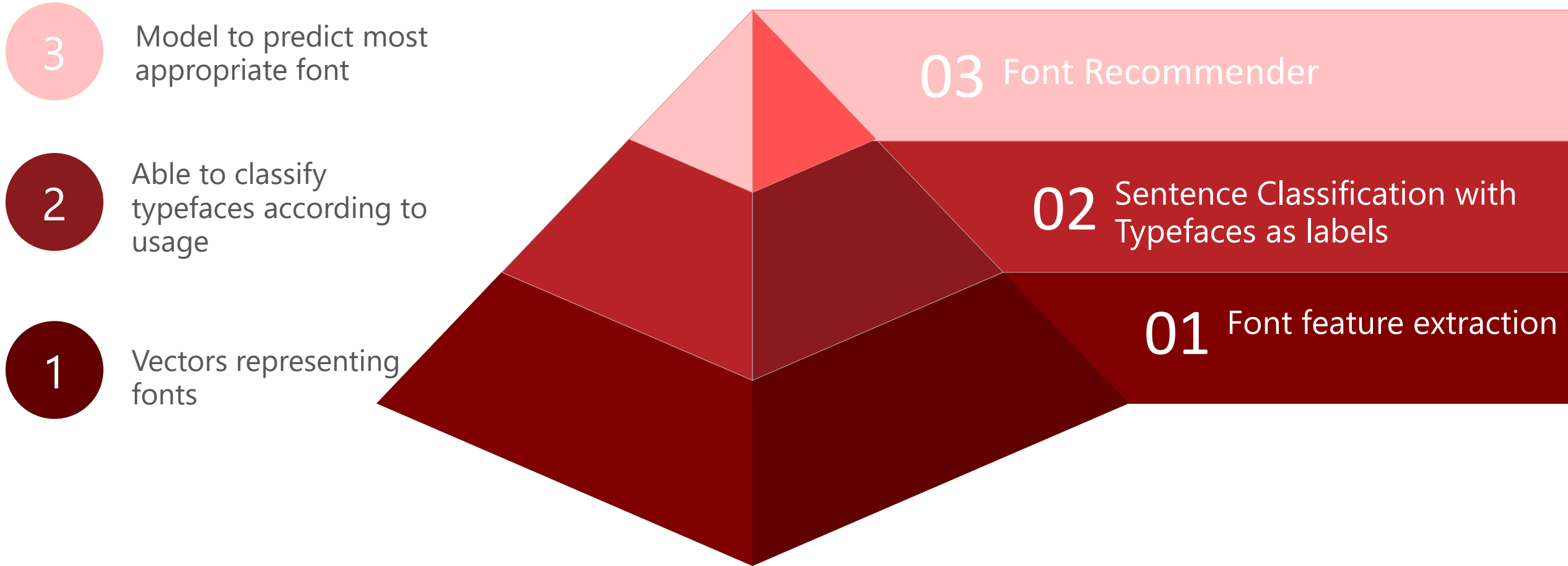
**CNN**

To train a usable model

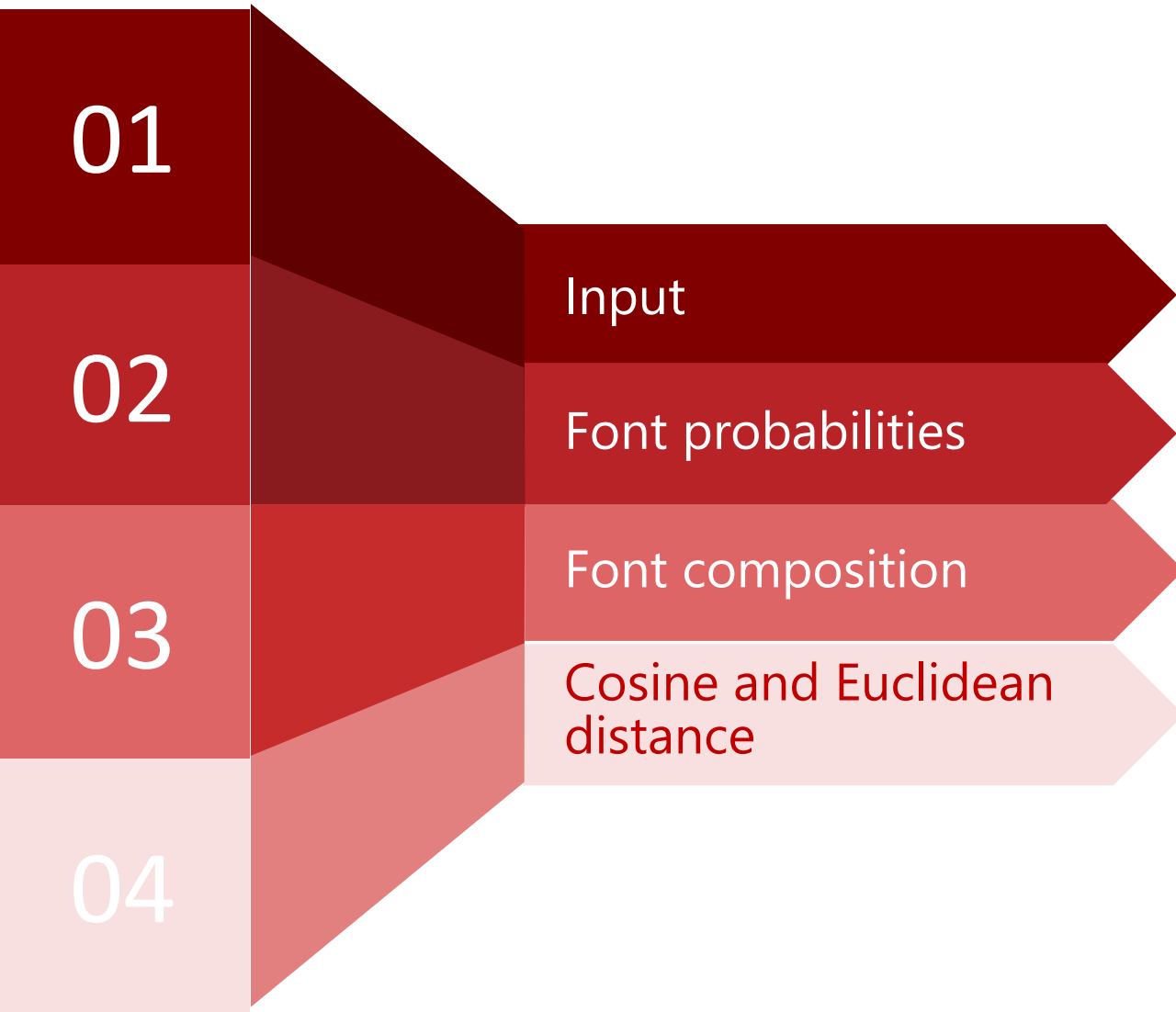


Proven to be effective  
in sentence  
classification

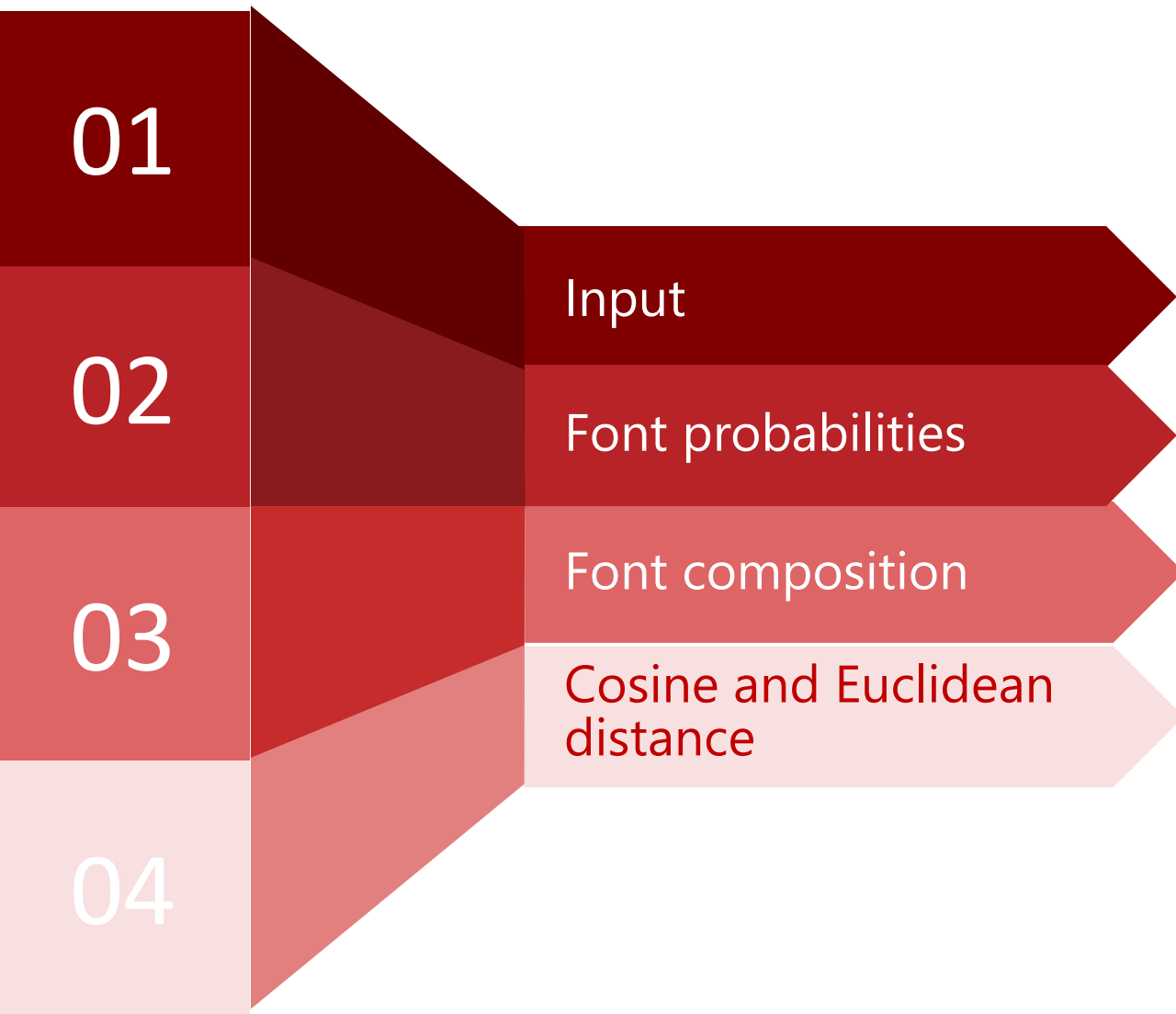
# Methodology



# Font Recommender



# Font Recommender

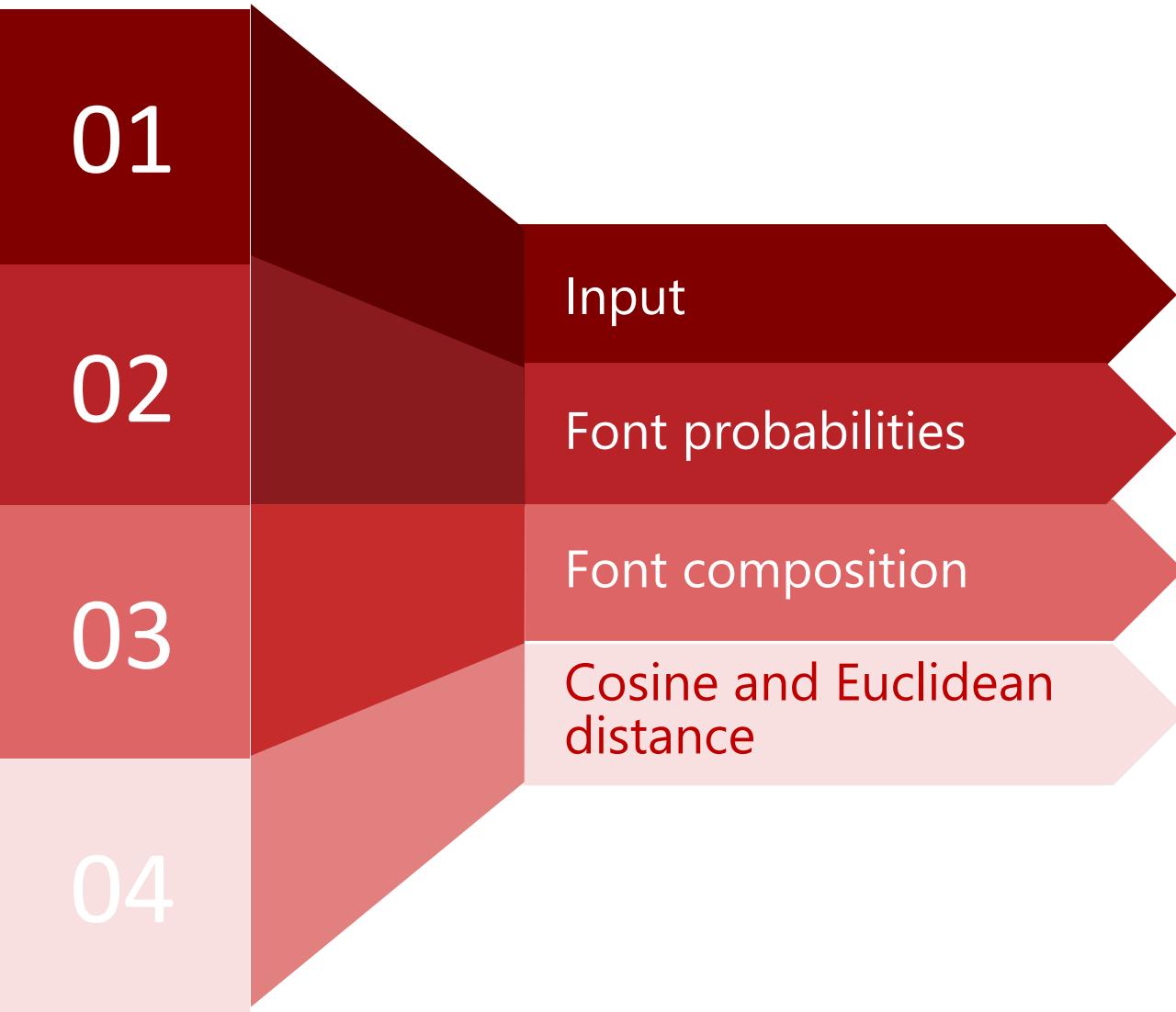


Type out a sentence that best represents what you want to convey

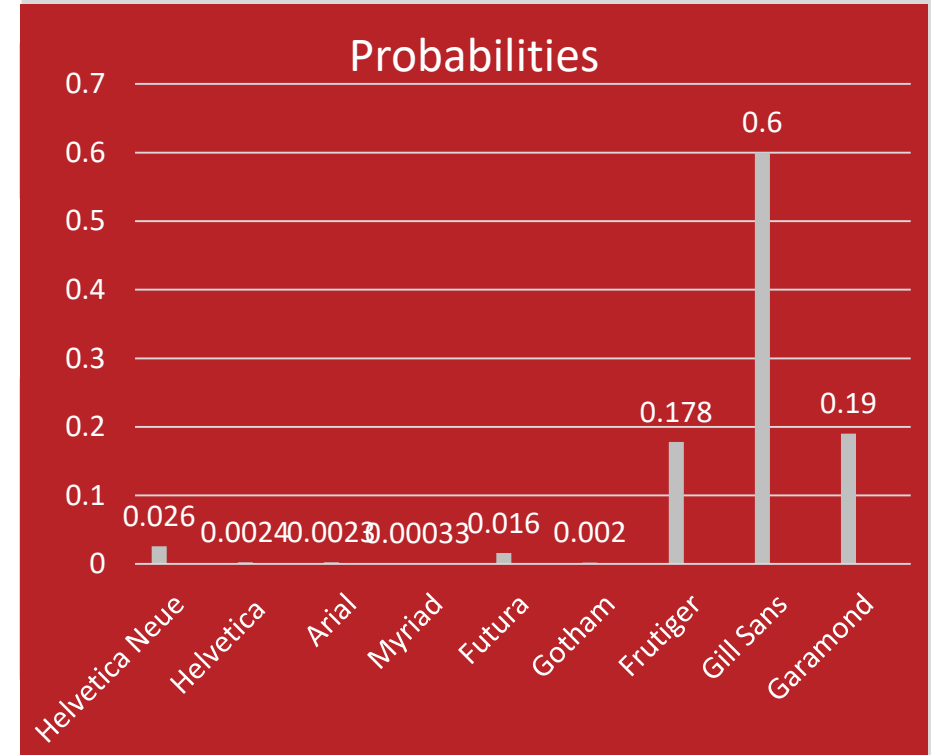
*"The Palestinian Museum is an independent institution dedicated to supporting an open and dynamic Palestinian culture..."*



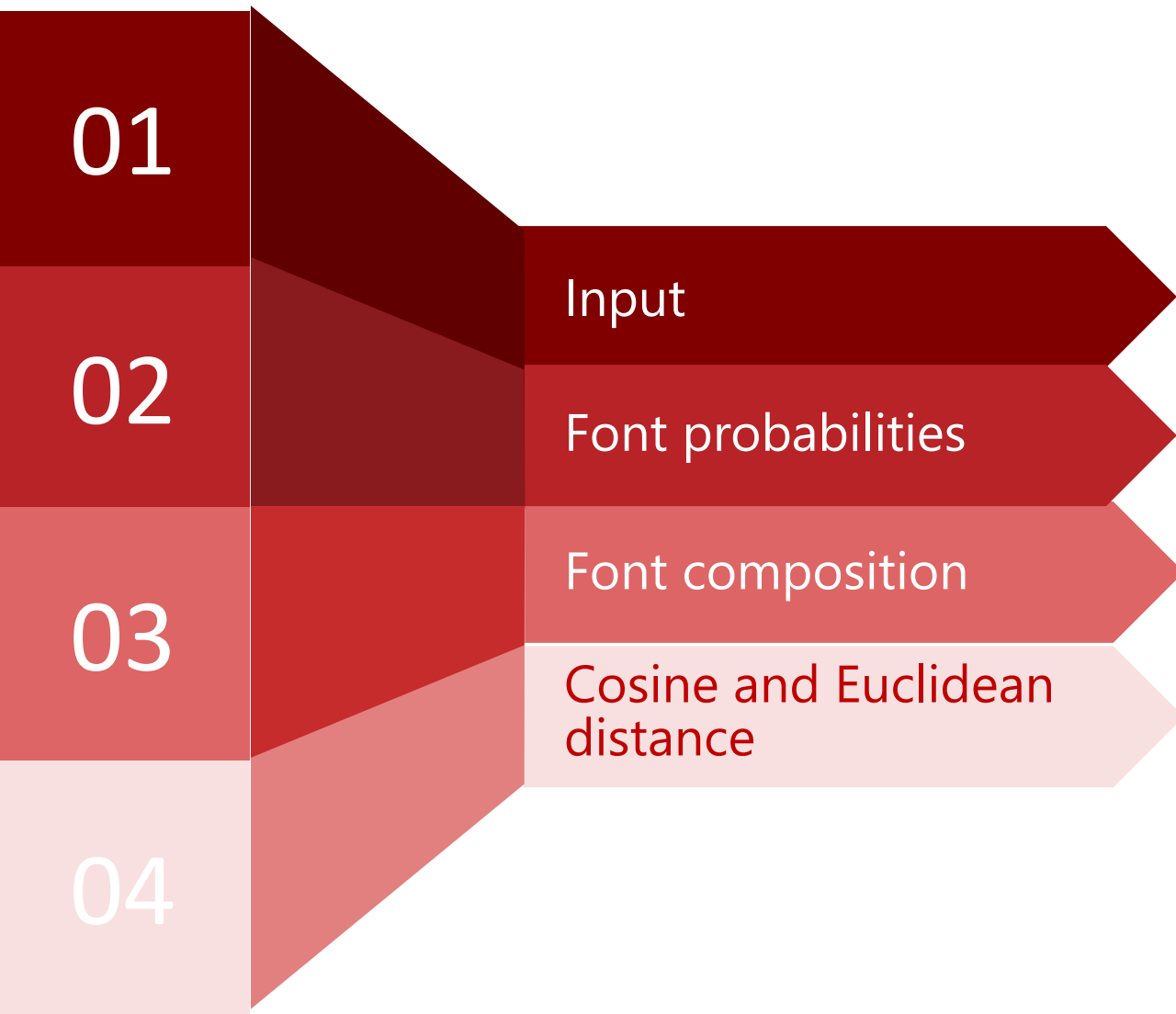
# Font Recommender



## Behind the scenes:



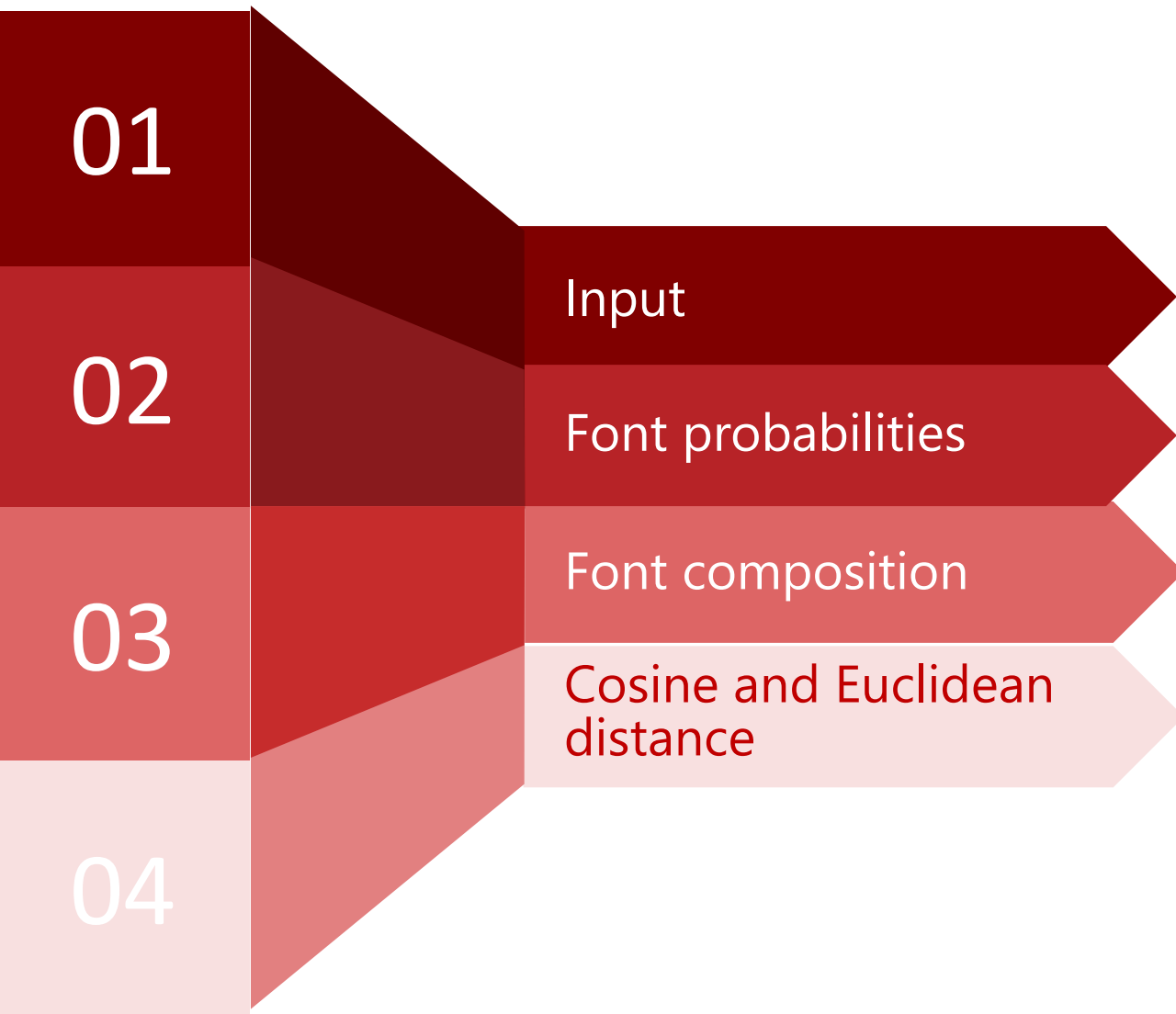
# Font Recommender



Behind the scenes:

$$\sum_{i=0}^n font_i prob_i$$

# Font Recommender



Here is your recommended font!

Font: **Oxygen**

the palestinian museum  
the palestinian museum  
**the palestinian museum**

# Example

Our test case was taken from the **International Society of Typographic Designers Certificate of Excellence Awardees**

example 1



example 2



# Demo



the recommendation:

enter mission here:

enter ideal font here:

Get font

analysis

main font

# Demo



enter mission here:

e with reliable technology

enter ideal font here:

Get font

the recommendation:

PT Sans regular

cosine similarity:

analysis

Helvetica Neue	0.095361	Helvetica	0.0775914
Arial	0.007729	Myriad	0.467078
Futura	0.003081	Gotham	0.292116
Frutiger	0.010451	Gill Sans	0.0395129
Garamond	0.00707802		

main font

Myriad

Laseg  
dhum  
Hloiv

# Demo



the recommendation:

enter mission here:

| | I

enter ideal font here:

|

Get font

analysis

main font

# Demo



the recommendation:

cosine similarity:

Istok Web regular

enter mission here:

abcdefghi

enter ideal font here:

Get font

analysis

Helvetica Neue	0.113091	Helvetica	0.111404
Arial	0.111231	Myriad	0.113985
Futura	0.121818	Gotham	0.106576
Frutiger	0.108691	Gill Sans	0.107164
Garamond	0.106041		

main font

Futura

All the probabilities are  
**11%**

Which is random chance



# Summary

## **input**

Paragraph describing  
corporation or project

**Data-aided  
recommendation by  
CNN**

## **intermediary**

Probability makeup of  
each font

## **composition**

Vector composition of  
vectors using probability  
estimates as weights

## **output**

Obtain top 3 fonts that are  
closest by Euclidean or  
Cosine distance

# Conclusion

1. First time working on semantic of typography, **hypothesis holds true**
2. System for intelligence augmentation using **data-driven contextual analysis**
3. Build larger and diverse dataset across **languages and domains**
4. Refine font recommender with designers according to their **needs**