

#NLP4CALL2019

# Predicting learner's knowledge of single words using machine learning

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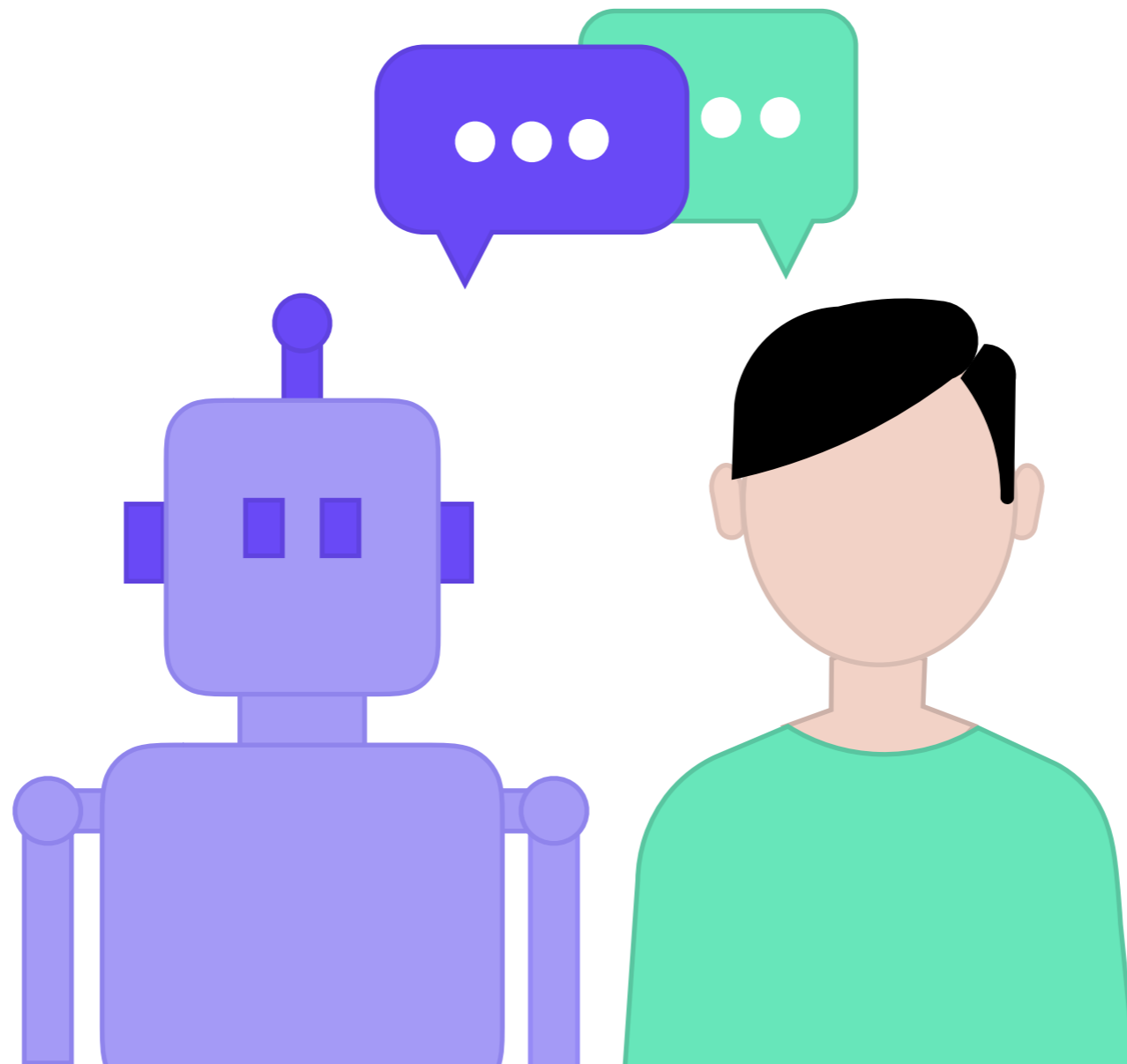
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  - Learner-specific
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# MOTIVATION

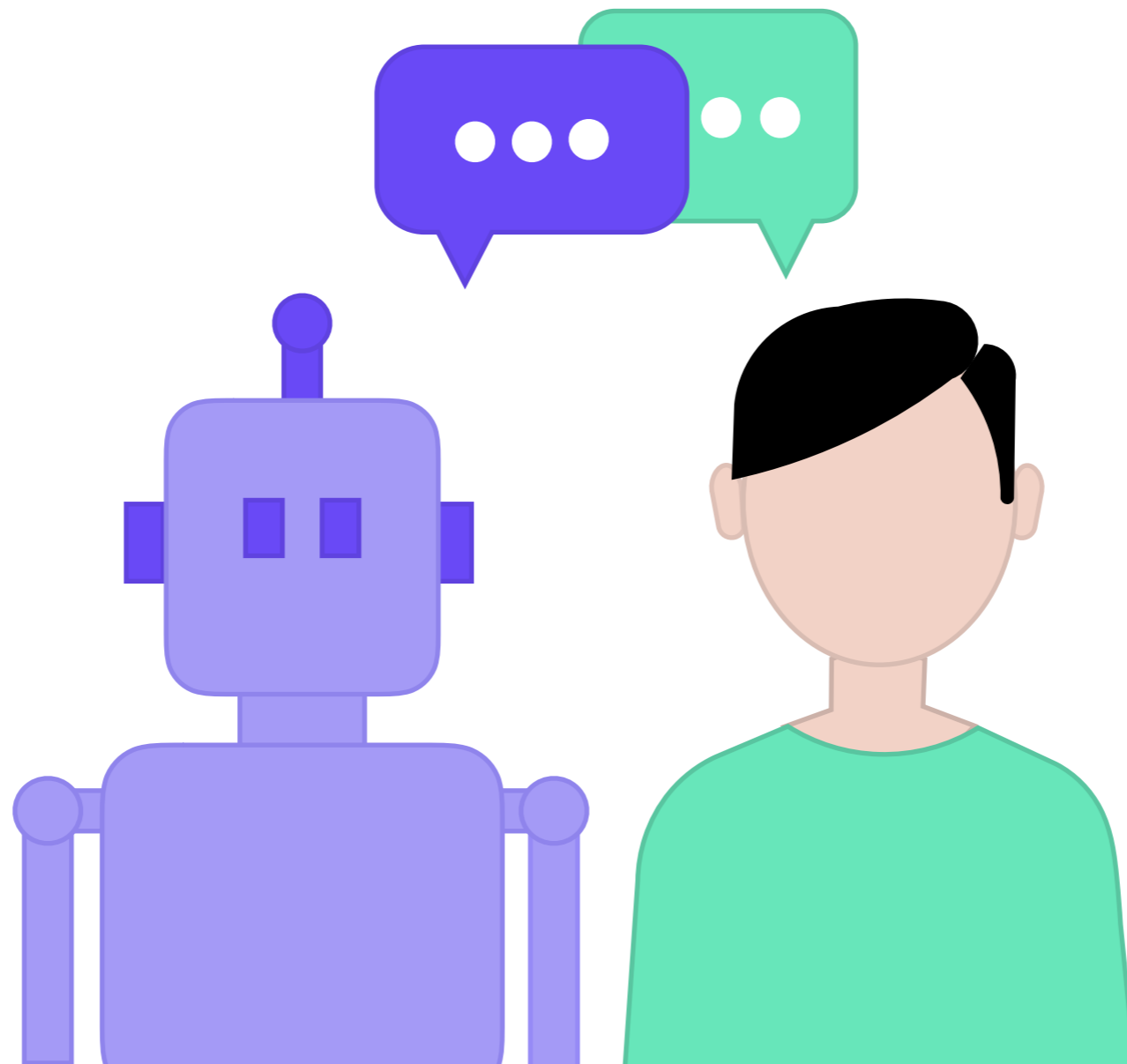
# MOTIVATION

for personalized interaction  
of intelligent tutoring systems  
with language learners



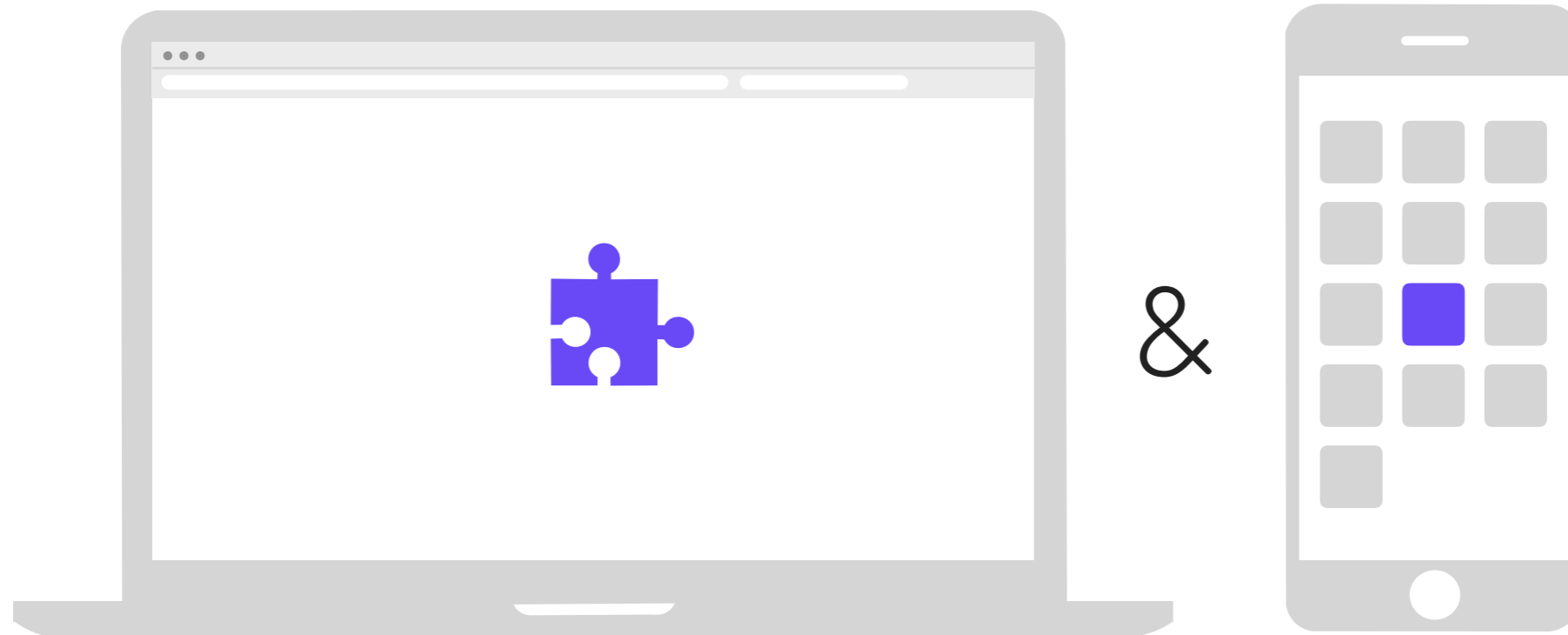
# MOTIVATION

for personalized interaction  
of the intelligent tutoring system **Elia**  
with language learners



# MOTIVATION

for personalized interaction  
of the intelligent tutoring system **Elia**  
with language learners



# MOTIVATION

FORM

**MEANING**

USE

Guess what it means

making a lot of noise

having two decisions

moving back and forth

# MOTIVATION

FORM

MEANING

USE

You might know this word!

It comes from the same word family as:

attend VERB

NEXT



# MOTIVATION

FORM

MEANING

USE

You have seen this word before in these sentence(s)!

This was a nice a **nice** evening.

They gave me a very **nice** present.

**NEXT**

# MOTIVATION

Fill the gap with a correct word!

The plan gained rapid media

shortly afterwards.

**NEXT**

# MOTIVATION

Choose the closest synonym of  
the word **-attention!**

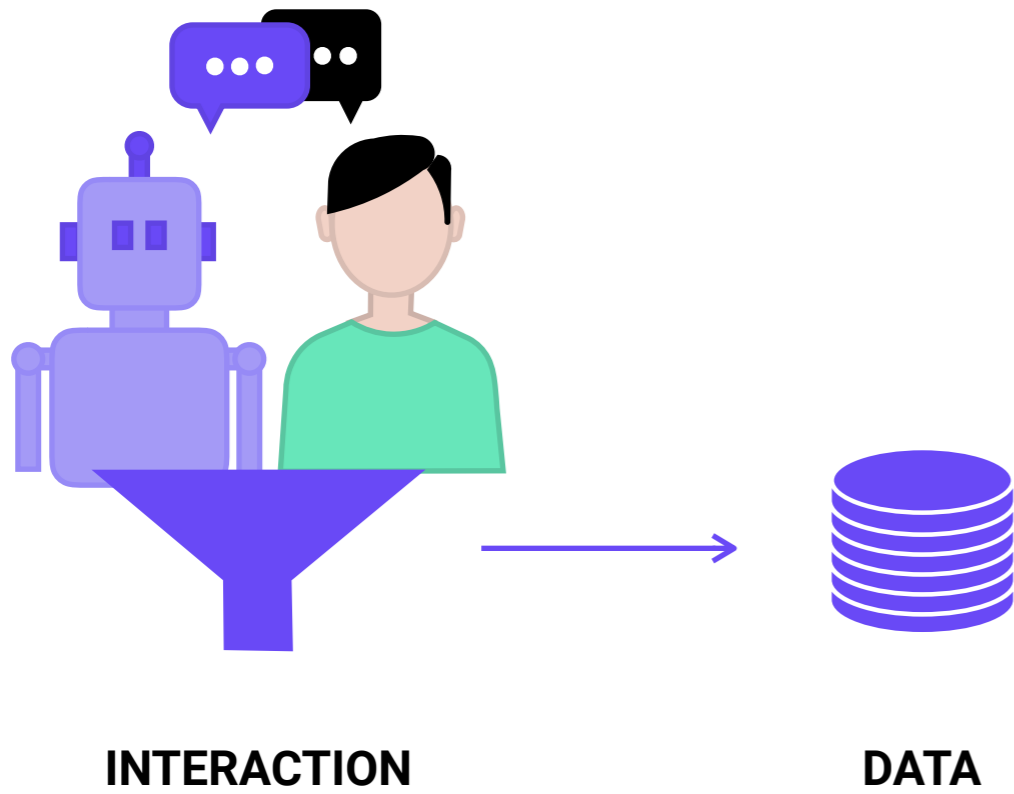
focus

characteristic

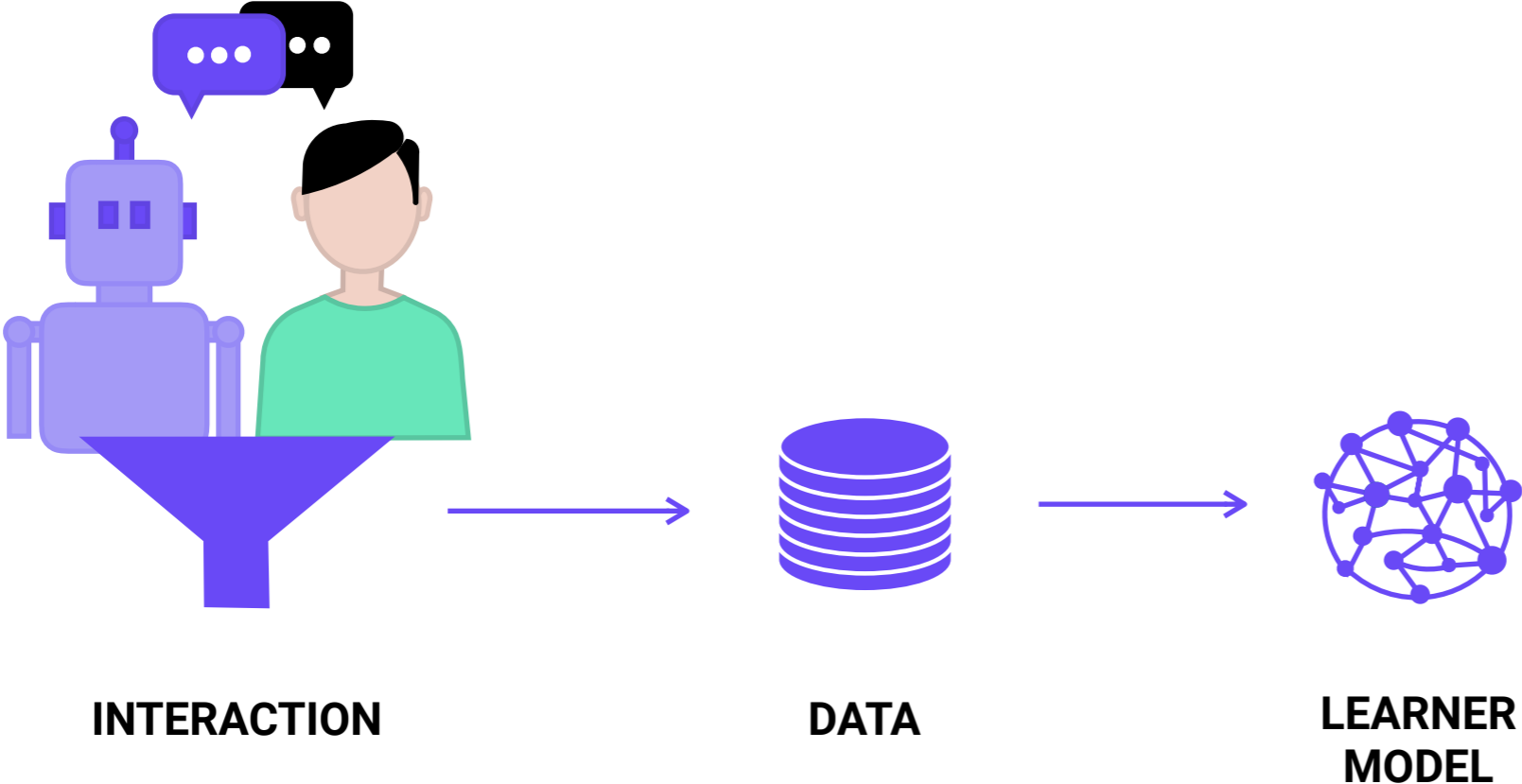
uniqueness

**NEXT**

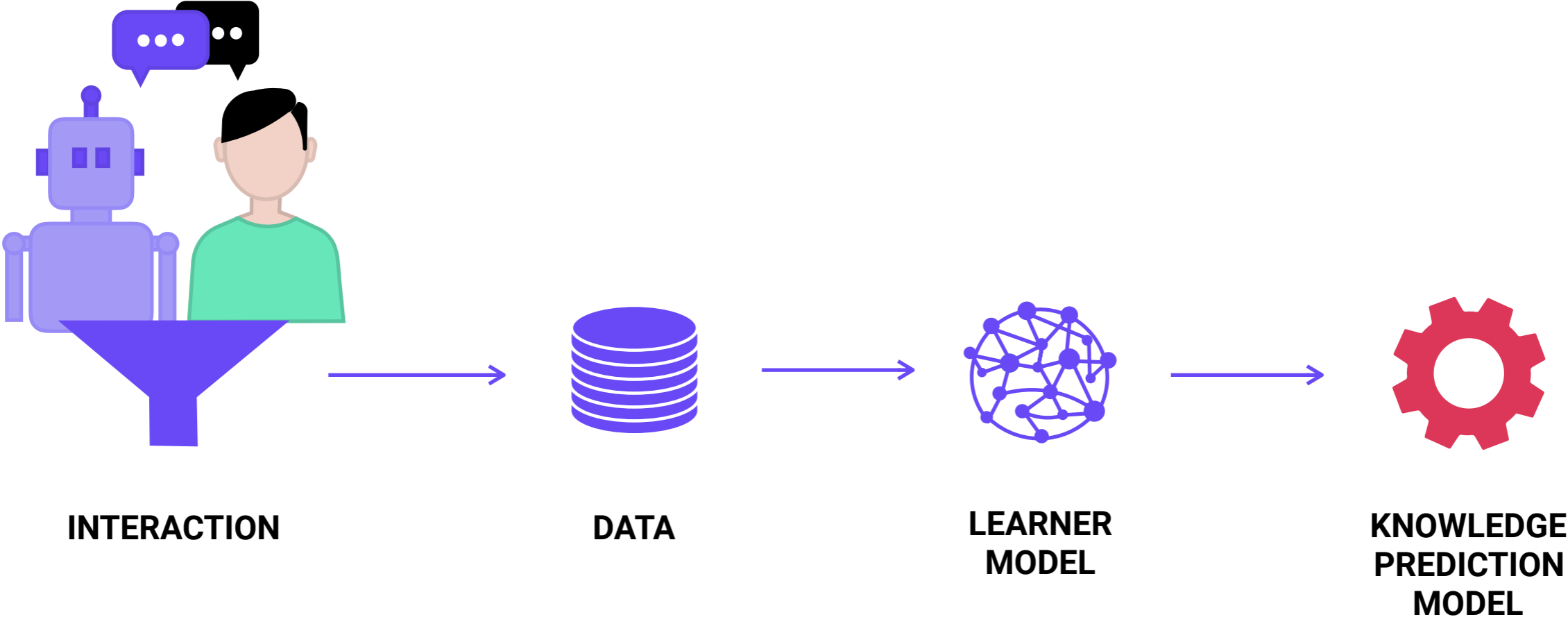
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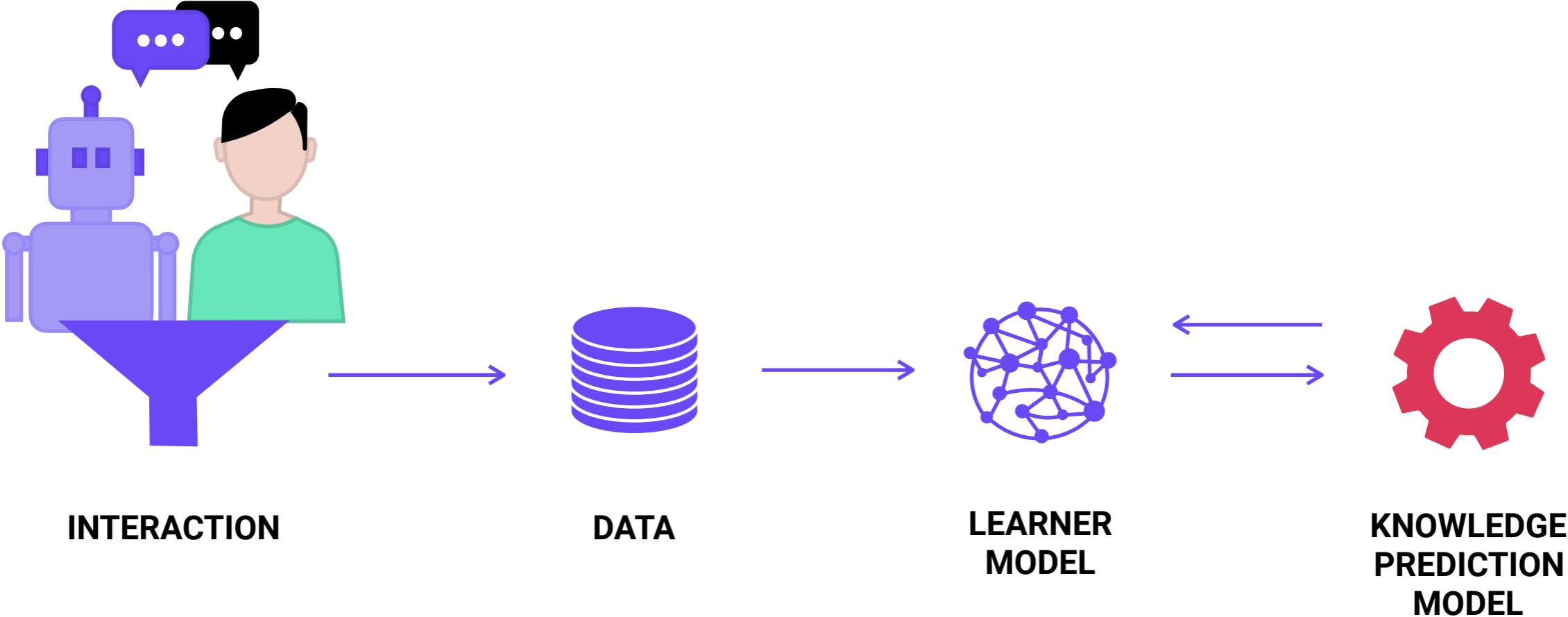
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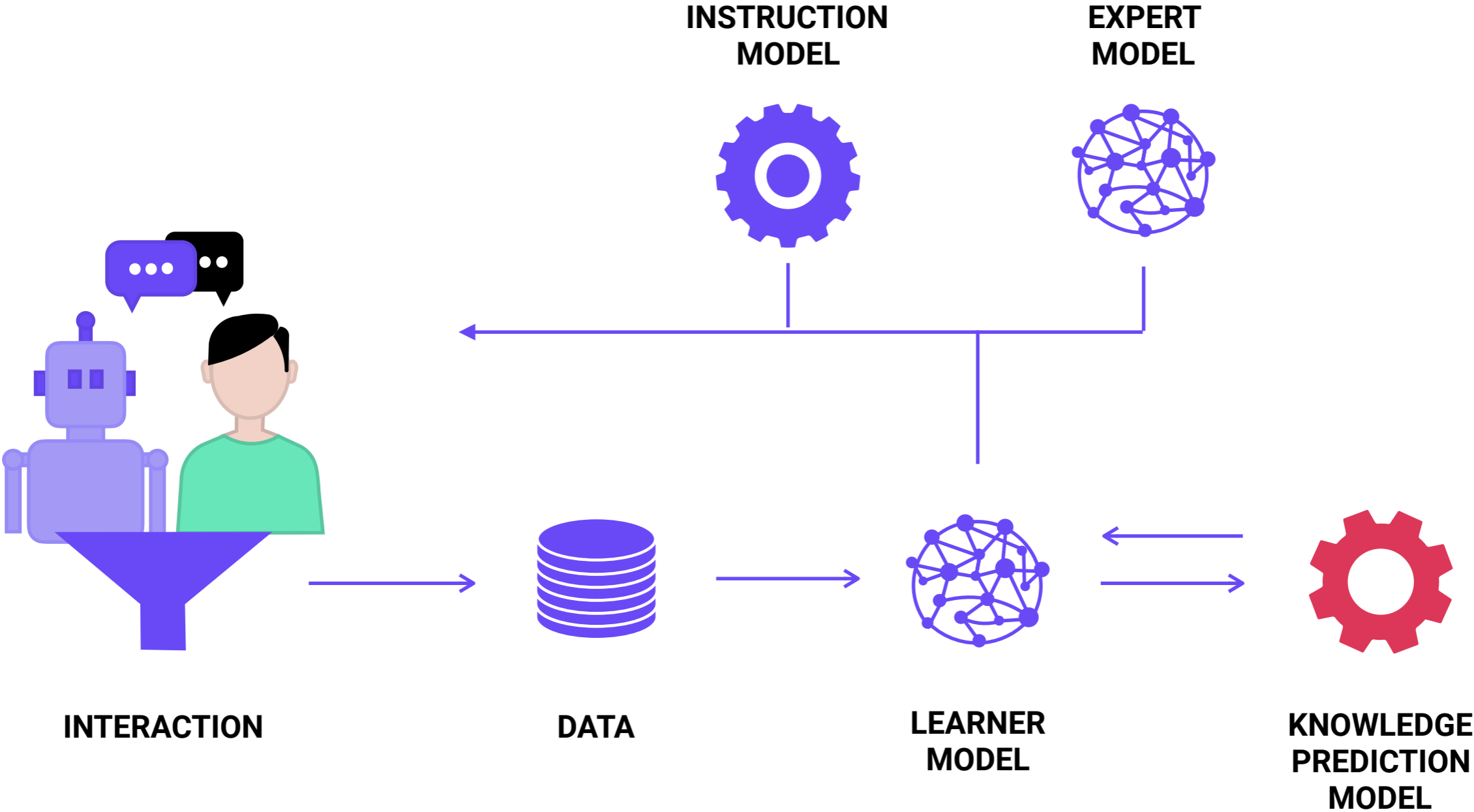
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# MOTIVATION



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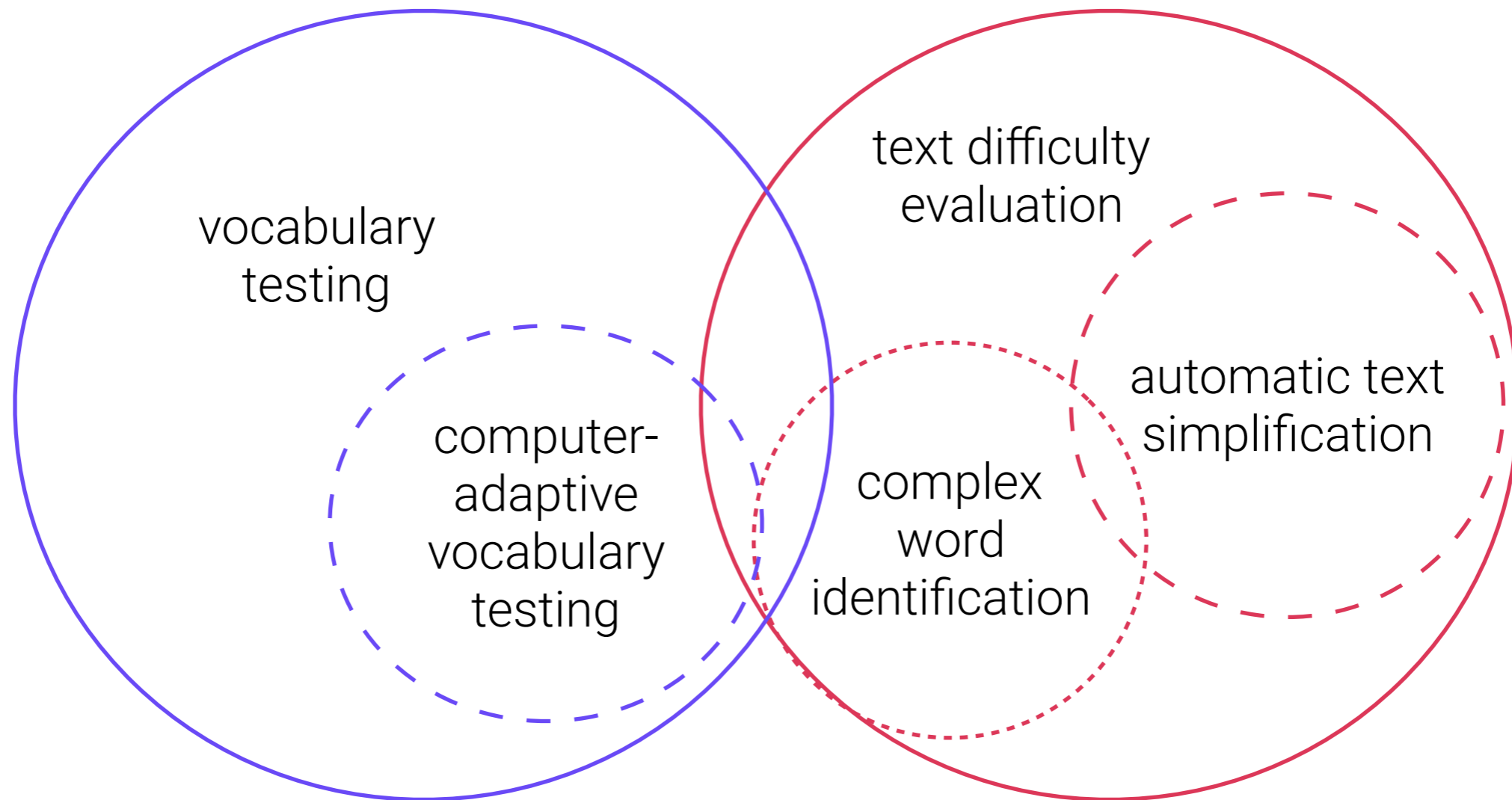




# SINGLE-WORD KNOWLEDGE PREDICTION

	<b>Word 1</b>	<b>Word 2</b>	<b>Word 3</b>	<b>Word 4</b>	<b>Word 5</b>	<b>Word ...</b>
<b>Learner 1</b>	???	KNOWN	UNKNOWN	???	UNKNOWN	KNOWN
<b>Learner 2</b>	UNKNOWN	???	???	KNOWN	KNOWN	KNOWN
<b>Learner 3</b>	KNOWN	KNOWN	UNKNOWN	UNKNOWN	???	KNOWN
<b>Learner 4</b>	???	KNOWN	???	KNOWN	???	UNKNOWN
<b>Learner 5</b>	UNKNOWN	???	???	???	UNKNOWN	???
<b>Learner 6</b>	KNOWN	???	KNOWN	UNKNOWN	KNOWN	KNOWN
<b>Learner ...</b>	UNKNOWN	???	KNOWN	KNOWN	???	???

# SINGLE-WORD KNOWLEDGE PREDICTION





# RELATED STUDIES

Studies which tried to predict the knowledge of **single words** for **language learners**.

Tack et al. (2016)      Alfter & Volodina (2018)      Lee & Yeung (2018)      Yancey & Lepage (2018)      Ehara et al. (2018)      *This study (2019)*

<b>application</b>	<b>text simplification /adaptation</b>	<b>exercise/ material generation</b>	<b>search for reading materials</b>	<b>text simplification</b>	<b>reading support</b>	<b>intelligent tutoring system</b>

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<b>personalized</b>	NO	NO	YES	NO	YES	YES

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<b>personalized</b>	NO	NO	YES	NO	YES	YES
<b>machine learning</b>	NO	YES	YES	YES	NO	YES

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<b>L2 learners</b>	French	Swedish	Chinese	Korean	English	English
<b>personalized</b>	NO	NO	YES	NO	YES	YES
<b>machine learning</b>	NO	YES	YES	YES	NO	YES
<b>unseen words</b>	NO	NO	YES	YES	YES	YES



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<b>L2 learners</b>	French	Swedish	Chinese	Korean	English	English
<b>personalized</b>	NO	NO	YES	NO	YES	YES
<b>machine learning</b>	NO	YES	YES	YES	NO	YES
<b>unseen words</b>	NO	NO	YES	YES	YES	YES
<b>all-learner-together training</b>	-	-	NO	-	-	YES

# OUR STUDY

# OUR STUDY



dataset



features



models

# DATASET

Ehara (2012)

	Word 1	Word 2	Word 3	Word 4	Word 5	Word ...
Learner 1						
Learner 2						
Learner 3						
Learner 4						
Learner 5						
Learner ...						

# DATASET

Ehara (2012)

**16 Japanese learners** x **12,000 words** from the SVL list

	Word 1	Word 2	Word 3	Word 4	Word 5	Word ...
Learner 1						
Learner 2						
Learner 3						
Learner 4						
Learner 5						
Learner ...						

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**Knowledge scale** from “Never seen the word” to “Know the word’s meaning”



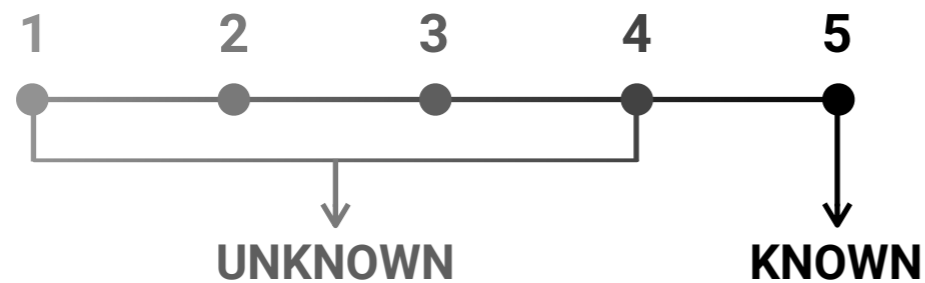
	Word 1	Word 2	Word 3	Word 4	Word 5	Word ...
Learner 1	5	3	5	1	5	3
Learner 2	2	5	2	2	4	5
Learner 3	3	5	1	4	1	1
Learner 4	4	5	5	1	2	2
Learner 5	3	4	5	3	5	5
Learner ...	2	4	2	4	2	5

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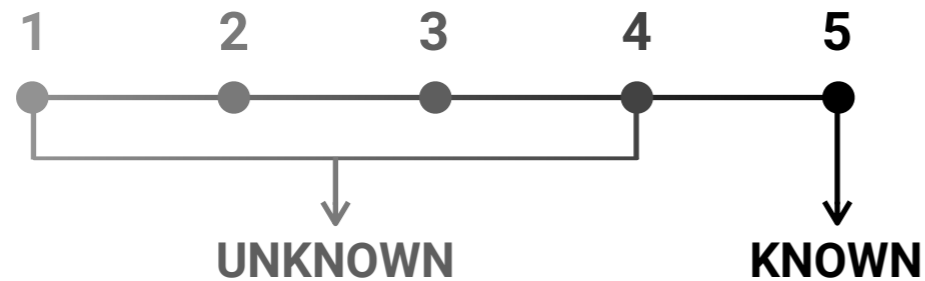
	Word 1	Word 2	Word 3	Word 4	Word 5	Word ...
Learner 1	5	3	5	1	5	3
Learner 2	2	5	2	2	4	5
Learner 3	3	5	1	4	1	1
Learner 4	4	5	5	1	2	2
Learner 5	3	4	5	3	5	5
Learner ...	2	4	2	4	2	5

# DATASET

Ehara (2012)

**16 Japanese learners** x **12,000 words** from the SVL list

**Knowledge scale** from “Never seen the word” to “Know the word’s meaning”



	Word 1	Word 2	Word 3	Word 4	Word 5	Word ...
Learner 1	KNOWN	UNKNOWN	KNOWN	UNKNOWN	KNOWN	UNKNOWN
Learner 2	UNKNOWN	KNOWN	UNKNOWN	UNKNOWN	UNKNOWN	KNOWN
Learner 3	UNKNOWN	KNOWN	UNKNOWN	UNKNOWN	UNKNOWN	UNKNOWN
Learner 4	UNKNOWN	KNOWN	KNOWN	UNKNOWN	UNKNOWN	UNKNOWN
Learner 5	UNKNOWN	UNKNOWN	KNOWN	UNKNOWN	KNOWN	KNOWN
Learner ...	UNKNOWN	UNKNOWN	UNKNOWN	UNKNOWN	UNKNOWN	KNOWN















# FEATURES

normalized frequency of the target word in different genres → COCA corpus (Davies, 2008)  
psycholinguistic characteristics of the word

## Word-specific

	History	Movies	...	Meaningful	Feature 24	Feature 25	...	Feature 78	Label
<b>Learner 1</b> x <b>Word 1</b>	345261	568432	...	3					<b>KNOWN</b>
<b>Learner 1</b> x <b>Word 2</b>	7432	6930	...	5					<b>KNOWN</b>
<b>Learner 1</b> x <b>Word 3</b>	90234112	89996010	...	2					<b>UNKNOWN</b>
<b>Learner 2</b> x <b>Word 1</b>	345261	568432	...	3					<b>KNOWN</b>
<b>Learner 2</b> x <b>Word 2</b>	7432	6930	...	5					<b>UNKNOWN</b>
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## Word-specific

## Learner-specific

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# FEATURES

learner's knowledge of vocabulary in different genres

## Word-specific

## Learner-specific

	History	Movies	...	Meaningful	History	Movies	...	Sports	Label
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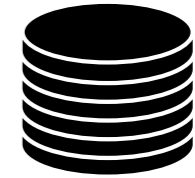
learner's knowledge of vocabulary in different genres and frequencies

## Word-specific

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learner's knowledge of vocabulary in different genres and frequencies

→ COCA corpus (Davies, 2008)

## Word-specific

## Learner-specific

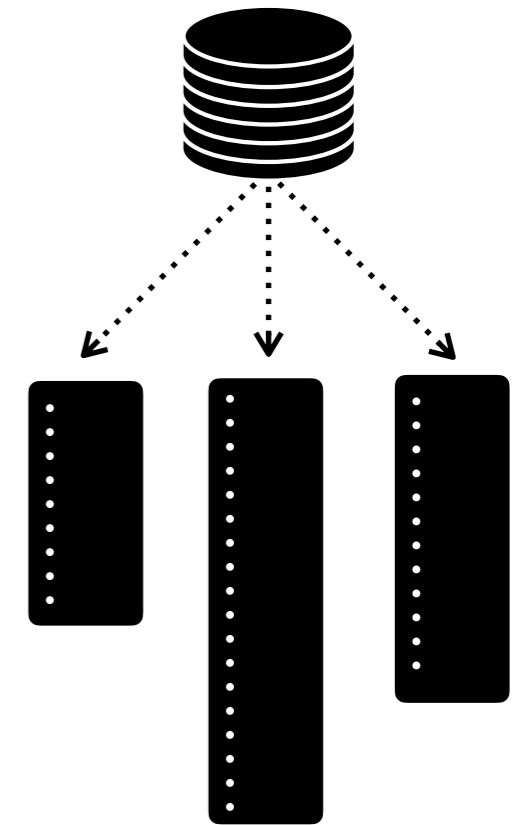
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learner's knowledge of vocabulary in different genres and frequencies

→ COCA corpus (Davies, 2008)

1. For each COCA subcorpus, keywordlist created



## Word-specific

## Learner-specific

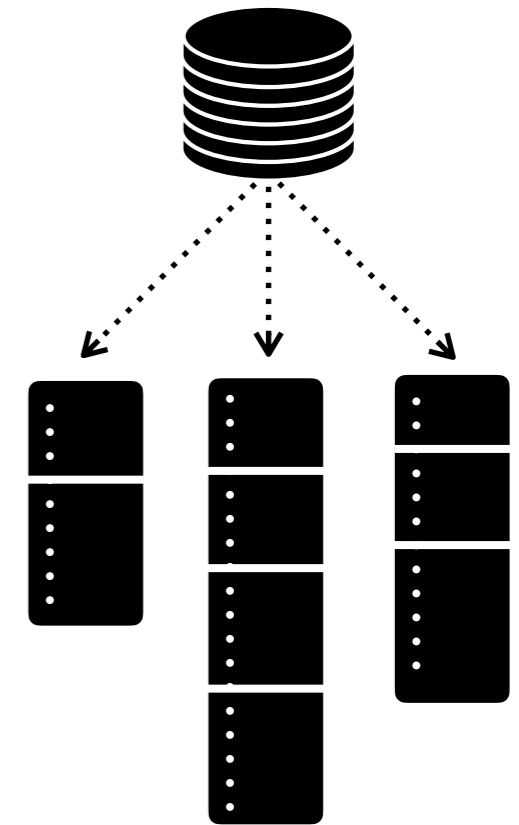
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# FEATURES

learner's knowledge of vocabulary in different genres and frequencies

→ COCA corpus (Davies, 2008)

1. For each COCA subcorpus, keywordlist created
2. For each keywordlist, frequency bands created



## Word-specific

## Learner-specific

	History	Movies	...	Meaningful	History	Movies	...	Sports	Label
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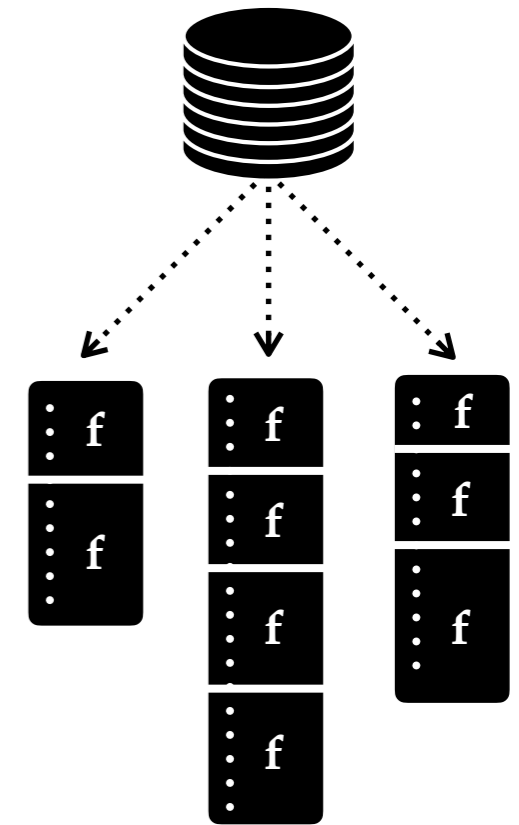


# FEATURES

learner's knowledge of vocabulary in different genres and frequencies

→ COCA corpus (Davies, 2008)

1. For each COCA subcorpus, keywordlist created
2. For each keywordlist, frequency bands created
3. For each frequency band, "frequency power" calculated



## Word-specific

## Learner-specific

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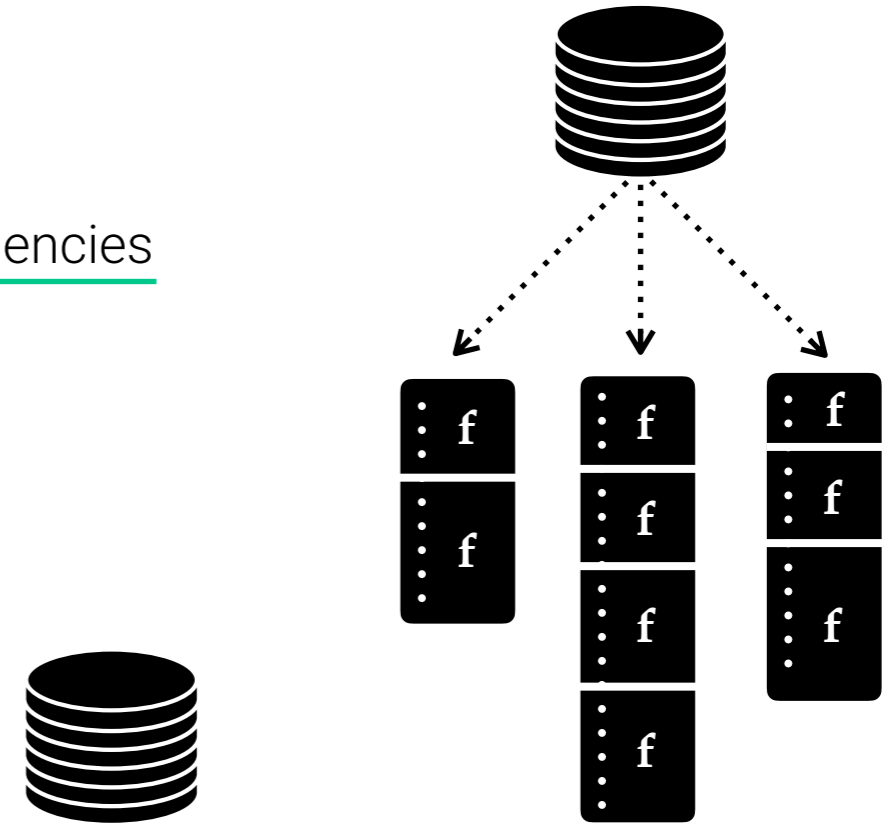
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learner's knowledge of vocabulary in different genres and frequencies

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→ Vocabulary Knowledge Dataset (Ehara, 2012)



## Word-specific

## Learner-specific

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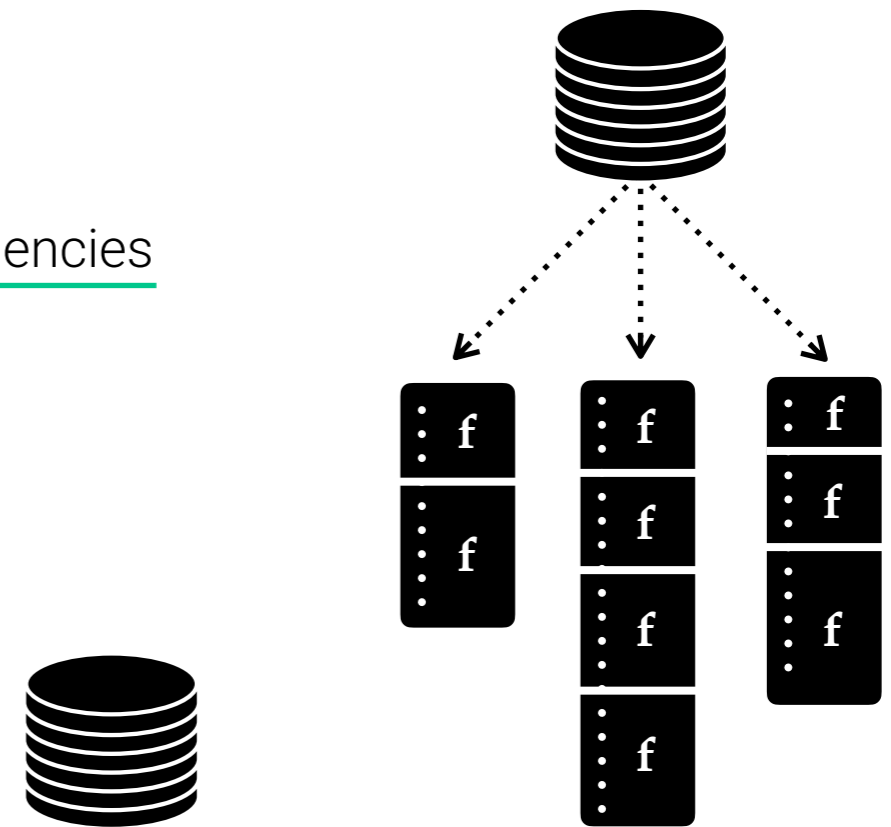
learner's knowledge of vocabulary in different genres and frequencies

→ COCA corpus (Davies, 2008)

1. For each COCA subcorpus, keywordlist created
2. For each keywordlist, frequency bands created
3. For each frequency band, "frequency power" calculated

→ Vocabulary Knowledge Dataset (Ehara, 2012)

4. For each keywordlist, learner's knowledge calculated



## Word-specific

## Learner-specific

	History	Movies	...	Meaningful	History	Movies	...	Sports	Label
<b>Learner 1</b> x <b>Word 1</b>	345261	568432	...	3					KNOWN
<b>Learner 1</b> x <b>Word 2</b>	7432	6930	...	5					KNOWN
<b>Learner 1</b> x <b>Word 3</b>	90234112	89996010	...	2					UNKNOWN
<b>Learner 2</b> x <b>Word 1</b>	345261	568432	...	3					KNOWN
<b>Learner 2</b> x <b>Word 2</b>	7432	6930	...	5					UNKNOWN
<b>Learner 2</b> x <b>Word ...</b>	90234112	89996010	...	2					UNKNOWN

# FEATURES

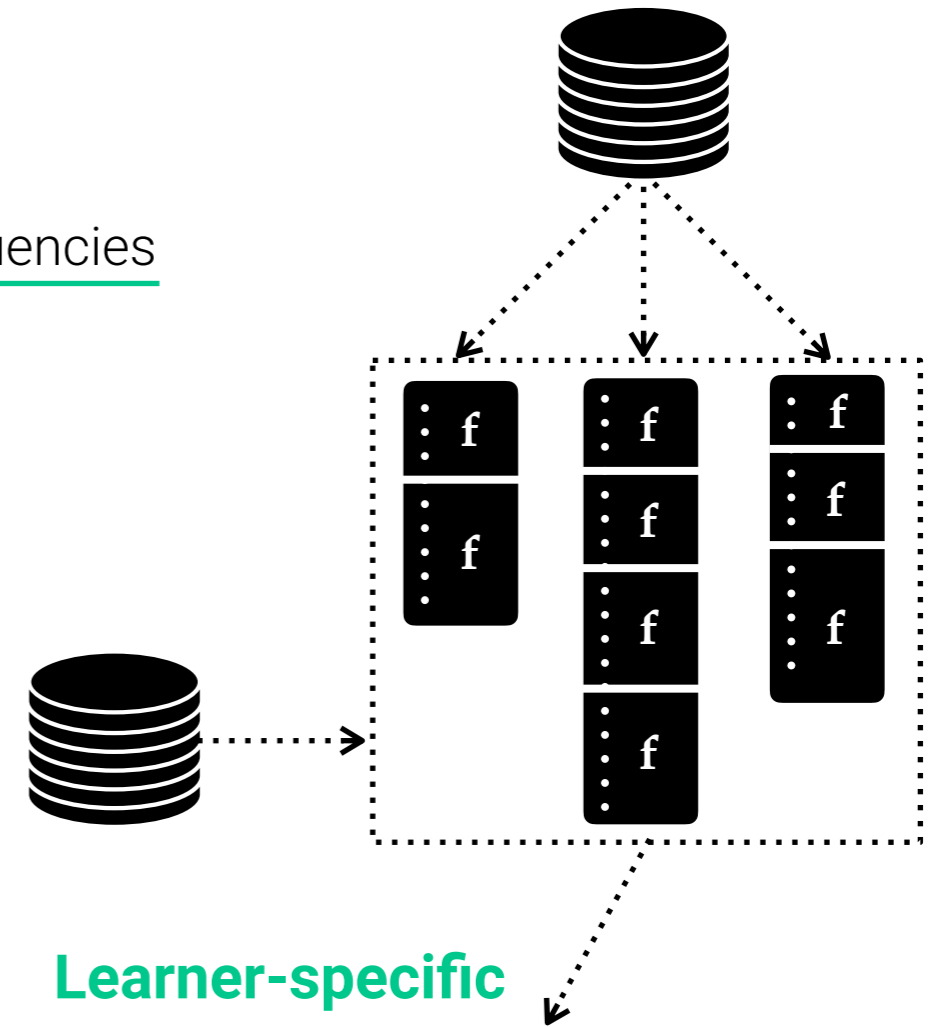
learner's knowledge of vocabulary in different genres and frequencies

→ COCA corpus (Davies, 2008)

1. For each COCA subcorpus, keywordlist created
2. For each keywordlist, frequency bands created
3. For each frequency band, "frequency power" calculated

→ Vocabulary Knowledge Dataset (Ehara, 2012)

4. For each keywordlist, learner's knowledge calculated



## Word-specific

## Learner-specific

	History	Movies	...	Meaningful	History	Movies	...	Sports	Label
<b>Learner 1</b> x <b>Word 1</b>	345261	568432	...	3	40852	2526	...	1679	KNOWN
<b>Learner 1</b> x <b>Word 2</b>	7432	6930	...	5	40852	2526	...	1679	KNOWN
<b>Learner 1</b> x <b>Word 3</b>	90234112	89996010	...	2	40852	2526	...	1679	UNKNOWN
<b>Learner 2</b> x <b>Word 1</b>	345261	568432	...	3	75679	99781	...	87540	KNOWN
<b>Learner 2</b> x <b>Word 2</b>	7432	6930	...	5	75679	99781	...	87540	UNKNOWN
<b>Learner 2</b> x <b>Word ...</b>	90234112	89996010	...	2	75679	99781	...	87540	UNKNOWN

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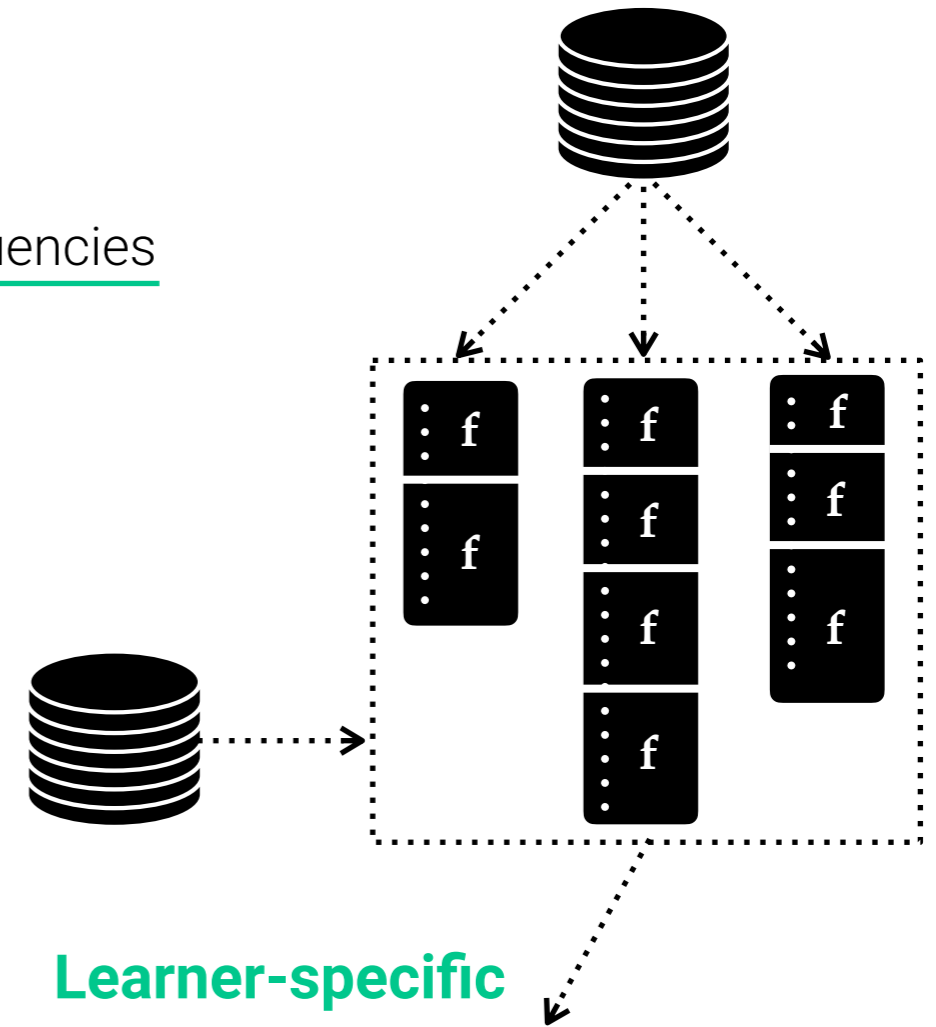
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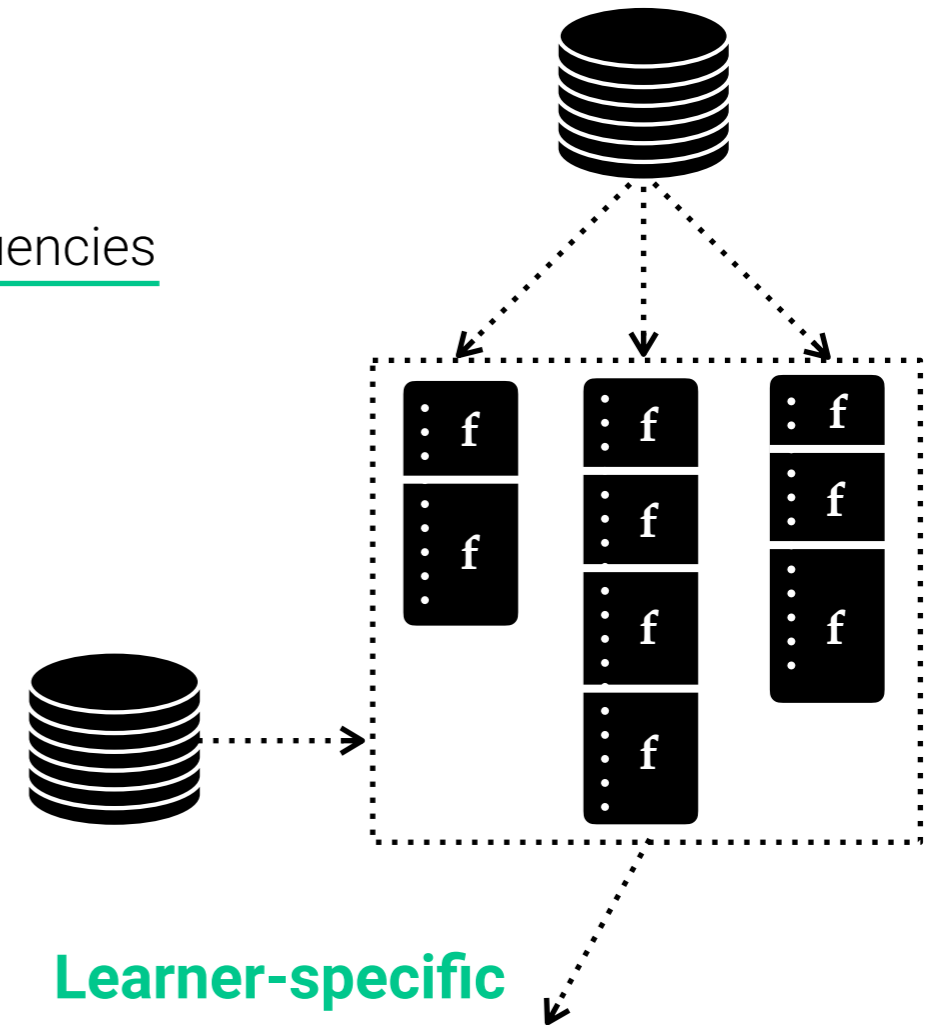
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<b>Learner 2</b> x <b>Word ...</b>	90234112	89996010	...	2	75679	99781	...	87540	UNKNOWN

# FEATURES

105 features in total

## Word-specific

## Learner-specific

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<b>Learner 1</b> x <b>Word 2</b>	7432	6930	...	5	40852	2526	...	1679	KNOWN
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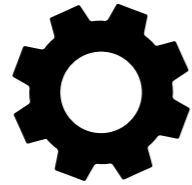
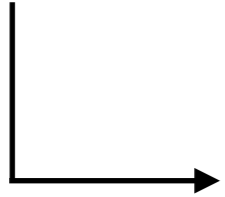
# FEATURE SELECTION

**105** features in total



# FEATURE SELECTION

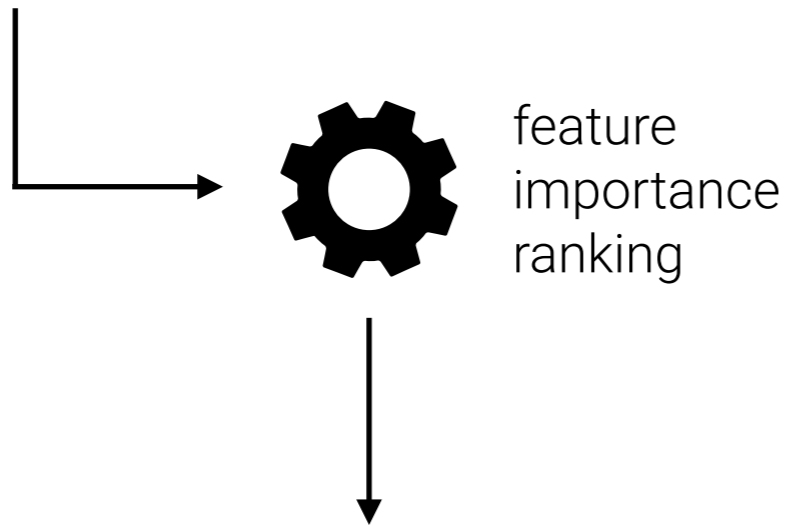
**105** features in total



feature  
importance  
ranking

# FEATURE SELECTION

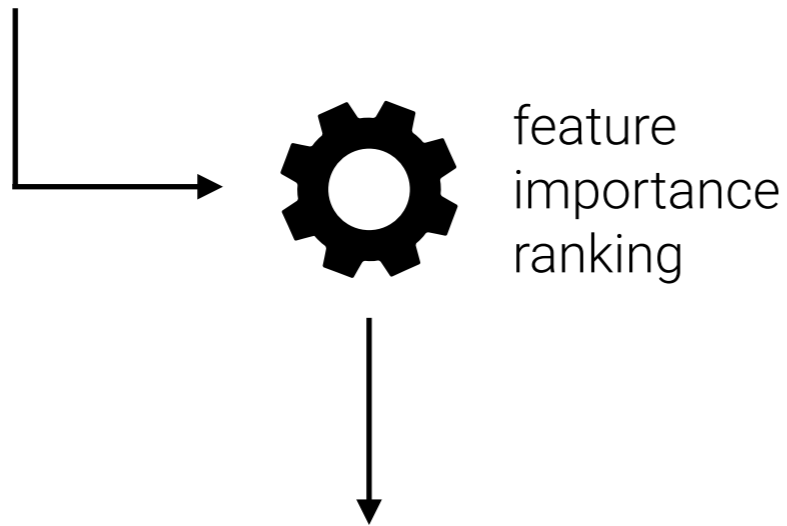
105 features in total



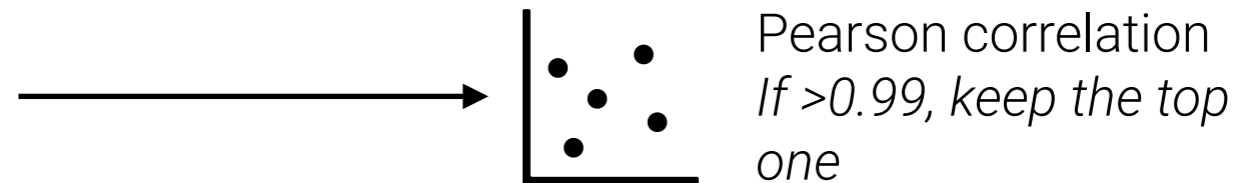
RANK	TYPE	NAME
1	word	total
2	word	spoken (national public radio)
3	word	academic
4	word	magazine (women/men)
5	word	fiction (journals)
...	...	...
101	learner	magazine (African American)
102	learner	magazine (home/health)
103	learner	magazine (science/tech)
104	learner	news (editorial)
105	learner	academic (history)

# FEATURE SELECTION

105 features in total

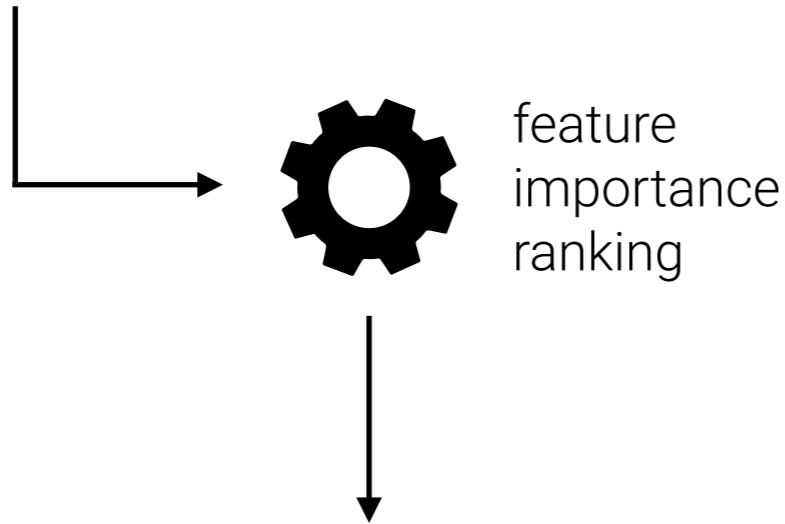


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# FEATURE SELECTION

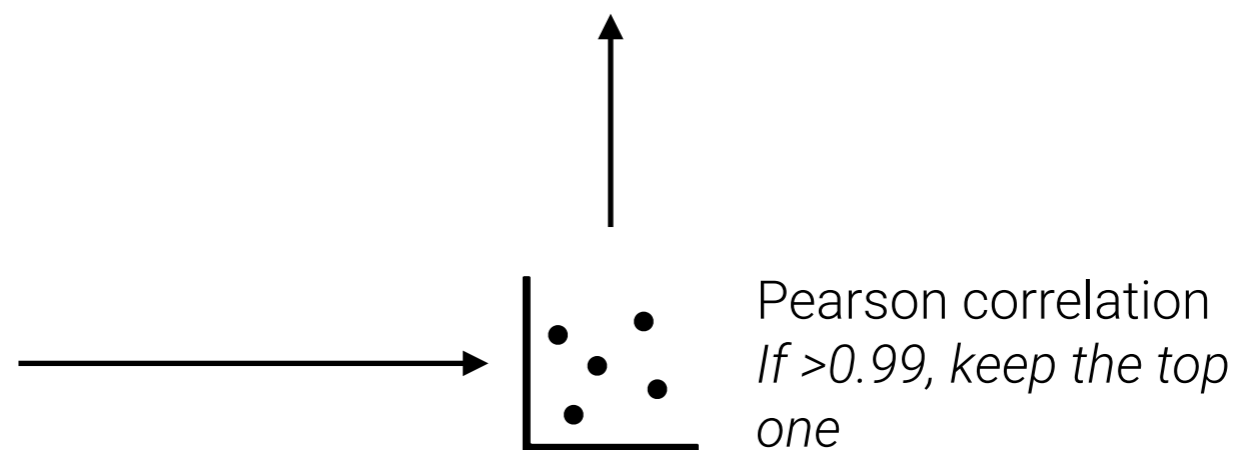
105 features in total



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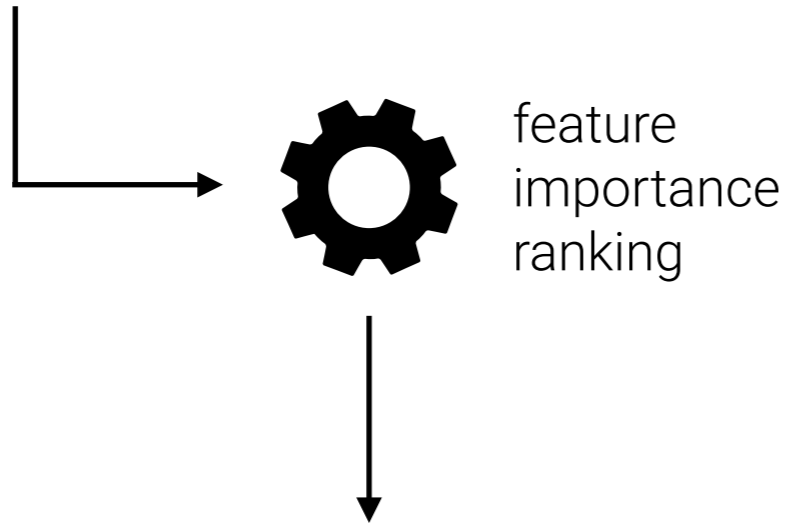
39 features in total

RANK	TYPE	NAME
1	word	total
2	word	spoken (national public radio)
3	word	academic
4	word	magazine (women/men)
5	word	fiction (journals)
...	...	...
35	learner	news (local)
36	learner	spoken (independent)
37	learner	fiction
38	learner	spoken (public broadcasting)
39	learner	spoken (national public radio)



# FEATURE SELECTION

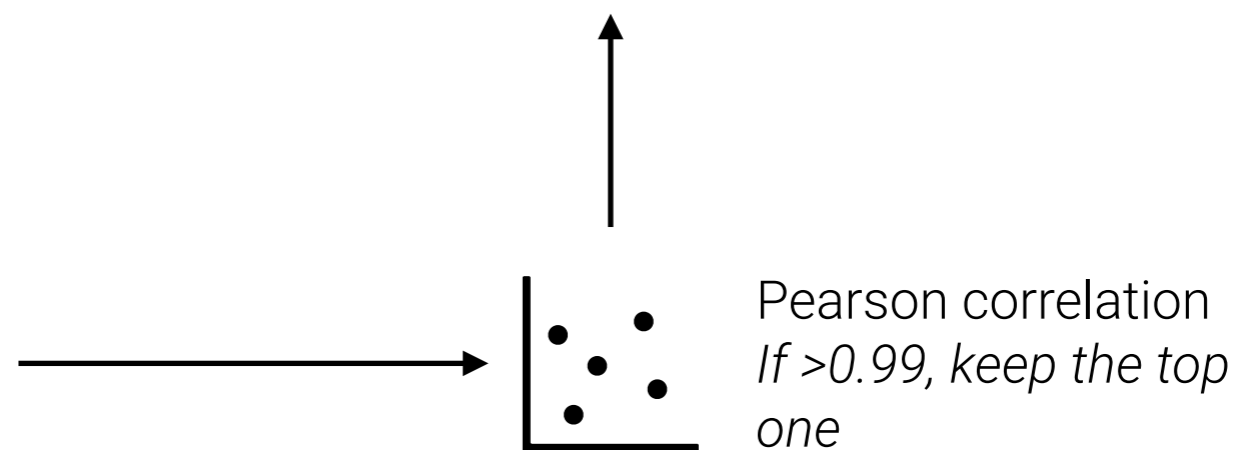
105 features in total



RANK	TYPE	NAME
1	word	total
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...	...	...
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36	learner	spoken (independent)
37	learner	fiction
38	learner	spoken (public broadcasting)
39	learner	spoken (national public radio)



# MODEL

**Complete Feature Space Dependent**

# MODEL

## Complete Feature Space Dependent

word-specific  
learner-specific

# MODEL

## Complete Feature Space Dependent

word-specific  
learner-specific

Random Forest  
Support Vector Machine  
k Nearest Neighbors  
Logistic Regression.



# MODEL

## Complete Feature Space Dependent

word-specific  
learner-specific

Random Forest

~~Support Vector Machine~~

~~k Nearest Neighbors~~

~~Logistic Regression.~~

# MODEL

## Complete Feature Space Dependent

word-specific  
learner-specific

Random Forest  
~~Support Vector Machine~~  
~~k Nearest Neighbors~~  
~~Logistic Regression.~~

80% - 3% - 17%  
training - validation - testing

# MODEL

## Complete Feature Space Dependent

word-specific  
learner-specific

Random Forest  
~~Support Vector Machine~~  
~~k Nearest Neighbors~~  
~~Logistic Regression.~~

80% - 3% - 17%  
training - validation - testing

## Neural Network Based

# MODEL

## Complete Feature Space Dependent

word-specific  
learner-specific

Random Forest  
~~Support Vector Machine~~  
~~k Nearest Neighbors~~  
~~Logistic Regression.~~

80% - 3% - 17%  
training - validation - testing

## Neural Network Based

word-specific

# MODEL

## Complete Feature Space Dependent

word-specific  
learner-specific

Random Forest  
~~Support Vector Machine~~  
~~k Nearest Neighbors~~  
~~Logistic Regression.~~

80% - 3% - 17%  
training - validation - testing

## Neural Network Based

word-specific

Fully Connected Neural Network  
*5 hidden layers*

# MODEL

## Complete Feature Space Dependent

word-specific  
learner-specific

Random Forest  
~~Support Vector Machine~~  
~~k Nearest Neighbors~~  
~~Logistic Regression.~~

80% - 3% - 17%  
training - validation - testing

## Neural Network Based

word-specific  
learner representation

Fully Connected Neural Network  
*5 hidden layers*



# MODEL

## Complete Feature Space Dependent

word-specific  
learner-specific

Random Forest  
~~Support Vector Machine~~  
~~k Nearest Neighbors~~  
~~Logistic Regression.~~

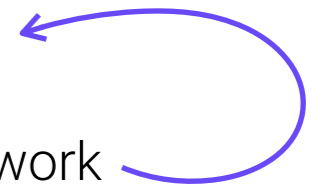
80% - 3% - 17%  
training - validation - testing

## Neural Network Based

word-specific  
learner representation

Fully Connected Neural Network  
*5 hidden layers*

80% - 3% - 17%  
training - validation - testing



# MODEL

## Complete Feature Space Dependent

word-specific  
learner-specific

Random Forest  
~~Support Vector Machine~~  
~~k Nearest Neighbors~~  
~~Logistic Regression.~~

80% - 3% - 17%  
training - validation - testing

**79.89%**

ACC

## Neural Network Based

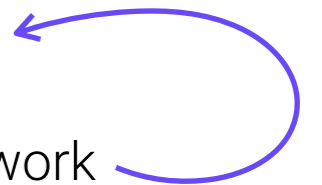
word-specific  
learner representation

Fully Connected Neural Network  
*5 hidden layers*

80% - 3% - 17%  
training - validation - testing

**79.18%**

ACC





# COMPARISON TO OTHER STUDIES

Tack et al. (2016)      Alfter & Volodina (2018)      Lee & Yeung (2018)      Yancey & Lepage (2018)      Ehara et al. (2018)      *This study*

<b>L2 learners</b>	<b>French</b>	<b>Swedish</b>	<b>Chinese</b>	<b>Korean</b>	<b>English</b>	<b>English</b>
<b>personalized</b>	<b>NO</b>	<b>NO</b>	<b>YES</b>	<b>NO</b>	<b>YES</b>	<b>YES</b>
<b>machine learning</b>	<b>NO</b>	<b>YES</b>	<b>YES</b>	<b>YES</b>	<b>NO</b>	<b>YES</b>
<b>for unseen words</b>	<b>NO</b>	<b>NO</b>	<b>YES</b>	<b>YES</b>	<b>YES</b>	<b>YES</b>
<b>all-learner-together training</b>	-	-	<b>NO</b>	-	<b>YES</b>	<b>YES</b>
<b>features</b>						
<b>models &amp; results (accuracy)</b>						

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<b>personalized</b>	<b>NO</b>	<b>NO</b>	<b>YES</b>	<b>NO</b>	<b>YES</b>	<b>YES</b>
<b>machine learning</b>	<b>NO</b>	<b>YES</b>	<b>YES</b>	<b>YES</b>	<b>NO</b>	<b>YES</b>
<b>for unseen words</b>	<b>NO</b>	<b>NO</b>	<b>YES</b>	<b>YES</b>	<b>YES</b>	<b>YES</b>
<b>all-learner-together training</b>	-	-	<b>NO</b>	-	<b>YES</b>	<b>YES</b>
<b>features</b>	-	count-based, morphological, semantic, context-based	difficulty level, frequency, count-based	frequency, count-based, proficiency level	frequency in 6 corpora, learner ability learnt	freq, genre, learner's knowledge in genres
<b>models &amp; results (accuracy)</b>	<b>FLELex 92.3%</b> (only for known words)	<b>SVM 48%</b> <b>MLPerc 53%</b> <b>ExtraTree 59%</b>	<b>SVM 78.0%</b> <b>Oracle 79.1%</b>	<b>SVM Linear 83.0%</b> <b>SVM RBF 84.3%</b>	<b>Rasch 66.3%</b> <b>Shared 77.7%</b> <b>Proposd 77.8%</b>	<b>RForest 79.89%</b> <b>NNet 79.18%</b>

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<b>machine learning</b>	<b>NO</b>	<b>YES</b>	<b>YES</b>	<b>YES</b>	<b>NO</b>	<b>YES</b>
<b>for unseen words</b>	<b>NO</b>	<b>NO</b>	<b>YES</b>	<b>YES</b>	<b>YES</b>	<b>YES</b>
<b>all-learner-together training</b>	-	-	<b>NO</b>	-	<b>YES</b>	<b>YES</b>
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<b>machine learning</b>	<b>NO</b>	<b>YES</b>	<b>YES</b>	<b>YES</b>	<b>NO</b>	<b>YES</b>
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<b>all-learner-together training</b>	-	-	<b>NO</b>	-	<b>YES</b>	<b>YES</b>
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# **FUTURE WORK**


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- adaptation & integration of this model to an intelligent system Elia

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- adaptation & integration of this model to an intelligent system Elia
- data size effect on the accuracy of the model

# FUTURE WORK

- adaptation & integration of this model to an intelligent system Elia
  - data size effect on the accuracy of the model
- 
- collection & creation of a new dataset for single-word knowledge prediction
    - more learnes
    - less words
    - different levels of knowledge
    - different aspects of knowledge



# FUTURE WORK

- adaptation & integration of this model to an intelligent system Elia
  - data size effect on the accuracy of the model
- 
- collection & creation of a new dataset for single-word knowledge prediction
    - more learners
    - less words
    - different levels of knowledge
    - different aspects of knowledge
  - more studies
    - different features
    - different algorithms

# THANK YOU

Klara Ptacinova Klimcikova ([k.klimcikova@lmu.de](mailto:k.klimcikova@lmu.de))  
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