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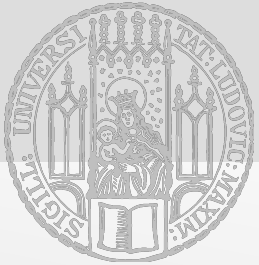

Vision-Based Approach to Noisy Text Recognition

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Computer Vision and Deep Learning: Automatic Image Understanding and Recognition SoSe23

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Content

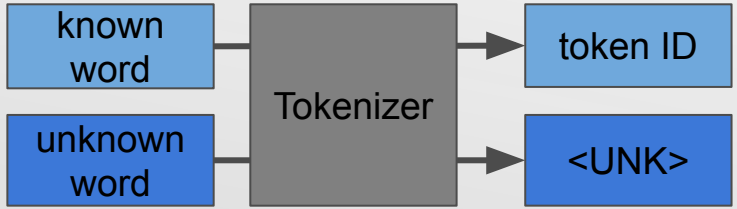
1. Motivation & Background

1. Out-of-Vocabulary (OOV) Problem in Natural Language Processing (NLP)

   unknown words appear in test set but not in training set.

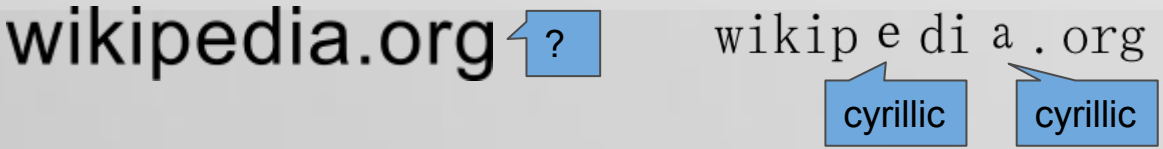
   caused by *small* training set or *noise*

   e.g. “word” → 5, “w0rd” → <UNK>



The OOV problem

2. Human Vision Robustness



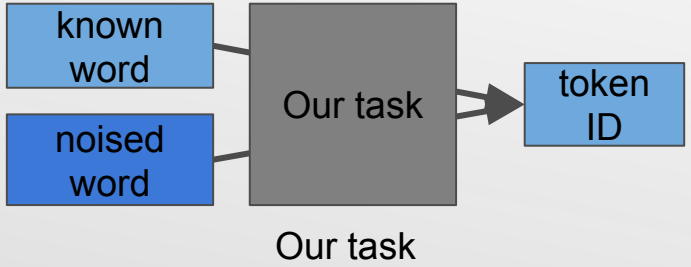
Can we improve **Robustness** again **Noise** by vision-based method?

e.g. “word” → 5, “w0rd” → 5

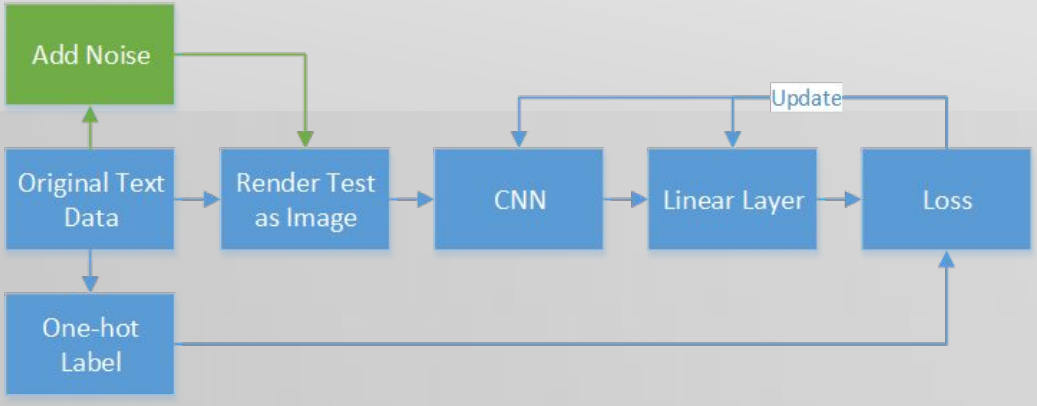
Image source: https://en.wikipedia.org/wiki/File:IDN_homograph_attack_1.svg

2. Task Definition & Conceptual Design

1. Robustness: OOV problem caused by Noise
2. Our system should recognize noised word and predict a correct Token ID



Methodology: a pipeline



3. Implementation: Model and Training

1. Dictionary:
   a. Word → ID
   b. ID → Word
2. Render Image with Pygame

definitely → definitely

3. CNN
   a. 1 Convolutional Layer
   b. Relu
   c. Max Pooling Layer
4. Linear Layer

   Output dimension: Size of Vocabulary
5. Loss: Cross Entropy
6. Optimizer: Adam

Image source: https://www.pygame.org/news

4. Experiments

4.1 Dataset

- **Multitarget TED Talks Task (MTTT) Dataset**
  - focus on the **English** portion of the en-de (English-German) translation set
  - count the frequency of each word
  - The most frequent **4571 words** were selected as token

4.2 Variables and Experimental Condition

4 fonts:

- Noto Sans
- Mandatory
- Turok
- Typographer

Figure 3. Four Fonts

4.2 Variables and Experimental Condition

➔ 3 types of noise: **Greek letters**, **Cyrillic letters**, and **leetspeak**

a → α


```
{
"a" : "a", "b" : "ь", "c" : "c", "d" : "", "e" : "e", "f" : "", "g" : "", "h" : "",
"i" : "", "j" : "", "k" : "к", "l" : "", "m" : "м", "n" : "п", "o" : "o", "p" : "p",
"q" : "", "r" : "г", "s" : "", "t" : "", "u" : "ц", "v" : "", "w" : "ш", "x" : "x",
"y" : "y", "z" : "", "A" : "A", "B" : "B", "C" : "C", "D" : "", "E" : "E", "F" : "Г",
"G" : "", "H" : "H", "I" : "", "J" : "", "K" : "K", "L" : "", "M" : "M", "N" : "И",
"O" : "O", "P" : "P", "Q" : "", "R" : "я", "S" : "", "T" : "T", "U" : "Ц", "V" : "",
"W" : "Ш", "X" : "X", "Y" : "Y", "Z" : ""
}
```


(a) Alphabet to Cyrillic Dictionary Cross Reference

a → a


```
{
"a" : "α", "b" : "β", "c" : "ς", "d" : "δ", "e" : "ε", "f" : "", "g" : "", "h" : "",
"i" : "ι", "j" : "", "k" : "κ", "l" : "", "m" : "", "n" : "η", "o" : "o", "p" : "ρ",
"q" : "", "r" : "", "s" : "", "t" : "τ", "u" : "μ", "v" : "υ", "w" : "ω", "x" : "χ",
"y" : "γ", "z" : "ζ", "A" : "Λ", "B" : "B", "C" : "", "D" : "", "E" : "E", "F" : "Γ",
"G" : "", "H" : "H", "I" : "I", "J" : "", "K" : "K", "L" : "", "M" : "M", "N" : "N",
"O" : "O", "P" : "P", "Q" : "", "R" : "", "S" : "", "T" : "T", "U" : "", "V" : "",
"W" : "", "X" : "X", "Y" : "Y", "Z" : "Z"
}
```


(b) Alphabet to Greek Dictionary Cross Reference

o → 0


```
{
"a" : "@", "b" : "6", "c" : "<", "d" : "ol", "e" : "3", "f" : "ſ", "g" : "9", "h" : "",
"i" : "!", "j" : "]", "k" : "|<", "l" : "1", "m" : "", "n" : "", "o" : "0", "p" : "þ",
"q" : "9", "r" : "", "s" : "5", "t" : "", "u" : "µ", "v" : "√", "w" : "vv", "x" : "×",
"y" : "¥", "z" : "2", "A" : "4", "B" : "8", "C" : "<", "D" : "", "E" : "3", "F" : "|=",
"G" : "6", "H" : "|-|", "I" : "1", "J" : "]", "K" : "|<", "L" : "└", "M" : "|V|", "N" : "|\\|",
"O" : "0", "P" : "|D", "Q" : "0_", "R" : "", "S" : "5", "T" : "+", "U" : "U", "V" : "\/",
"W" : "VV", "X" : "×", "Y" : "¥", "Z" : "2"
}
```

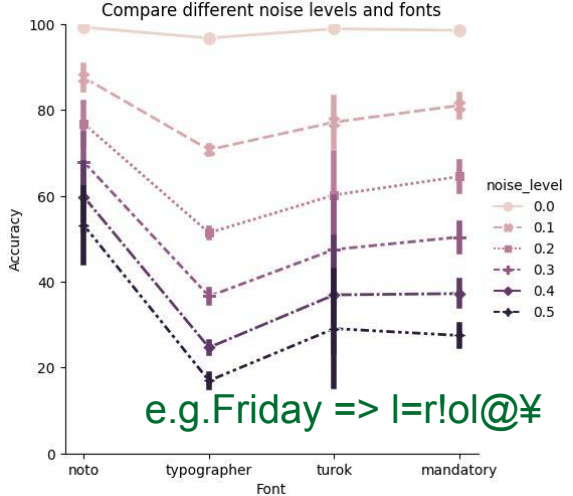

(c) Alphabet to Leetspeak Dictionary Cross Reference

➔ 5 probabilities of **10%**, **20%**, **30%**, **40%** and **50%** for each character replaced

Example words	10%	20%	30%	40%	50%
United	Un!ted	Un!t3ol	Un!t3ol	Un!t3ol	Un!t3ol
illusion	illu5!on	i1lu5!on	i1lu5!on	i1lu5!on	i1lu5!0n
Friday	Frida¥	Frida¥	Fr!d@¥	l=r!d@¥	l=r!ol@¥

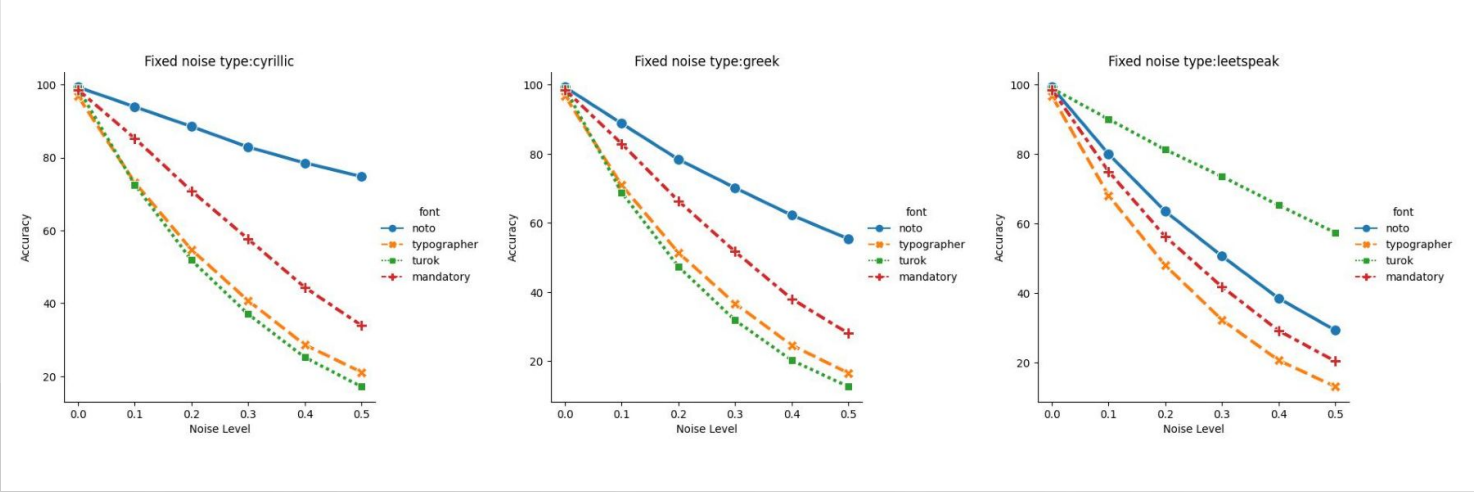
5. Results

5.1 Evaluation



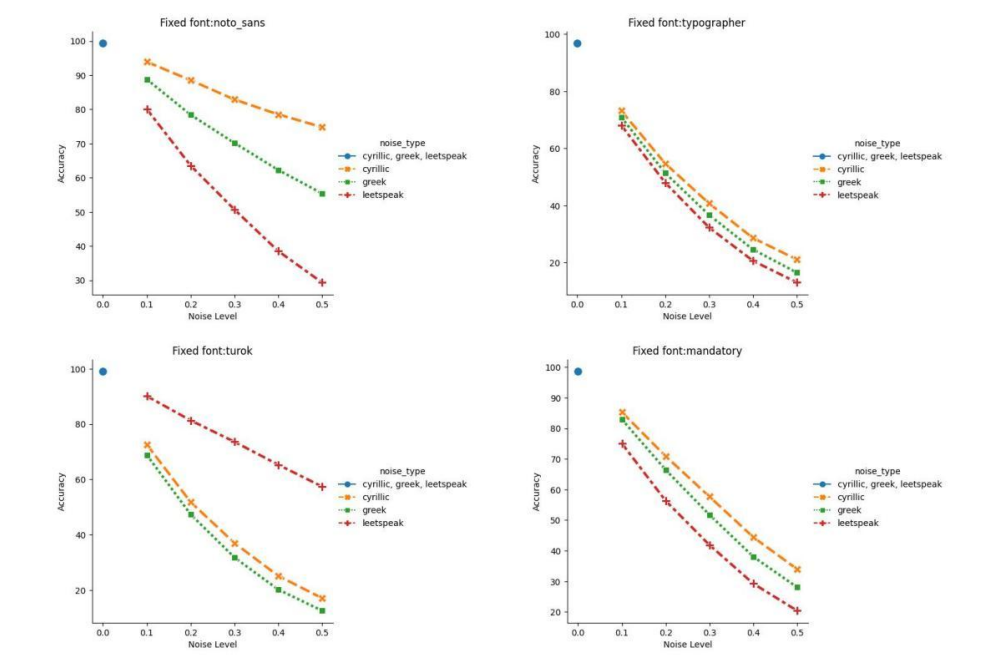
➔ Effect of noise ratio on model robustness in text recognition.

5.1 Evaluation: noise type fixed



➔ Effect of fonts on model robustness in text recognition.

5.1 Evaluation: font fixed



➔ Effect of noise type on model robustness in text recognition.

➔ Challenging: Leetspeak > Greek letters > Cyrillic letters

5.2 Statistical Analysis

	df	sum_sq	mean_sq	F	PR(F)
C(font)	3.0	6159.253119	2053.084373	13.216646	1.472186e − 06
C(noise_type)	3.0	8230.472624	2743.490875	17.661109	4.485511e − 08
C(noise_level)	5.0	17914.068404	3582.813681	23.064215	3.208838e − 12
Residual	53.0	8233.062735	155.340806	NaN	NaN

Table 3. Analysis of Variance (ANOVA) for variables.

- ➔ All of three variables, namely "Font" "Noise Type" and "Noise Level" have p-values lower than 0.05.
- ➔ Statistical significance: have decisive impact on the model's results for text recognition.

5.3 Case Study

Example word	Prediction word	Evaluation	Added noise	Prediction with noise	Evaluation
attract	attract	true	$attra < t$	attract	true
abstract	abstract	true	@b5tract	celebrate	false
previous	previous	true	$prev!ou5$	previous	true
obvious	obvious	true	o6v!0usly	carpeting	false
College	College	true	$< 0lle93$	ended	false
colleagues	colleagues	true	$< olle@gu3s$	imaginative	false

Table 2. Negative examples for case study.

➔ Result of case study for negative examples:

1. No noises added: words with similar characters can be successfully classified.
2. With noises: Classifying words with similar characters becomes challenging.

5. Conclusion and Future Work

- Built a pipeline to improve the Robustness against Noise via Vision Method
- Explored influence of fonts, noise, and noise level
- Case study: robustness against similar words

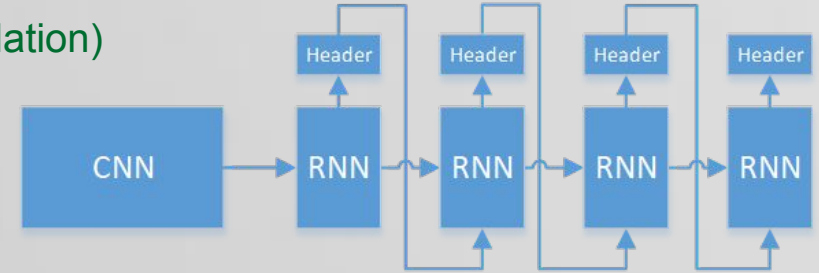
Limitation & Future Work

- Since words have **unfixed length**, we could **split** image into **slices**

- **Downstream Tasks** (e.g. Machine Translation)

  Now: CNN + Linear Layer (Header)

  Future: CNN + RNN (seq.)



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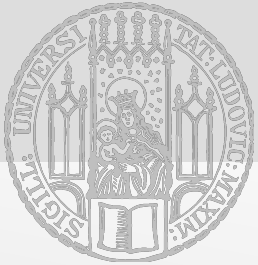

Thank you for listening

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