## **Al and Romanization:** Possibilities and Limitations

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#### Traditional Algorithms vs. Al & Machine Learning

TRADITIONAL ALGORITHMS

AI & MACHINE LEARNING

**Rule-based** 

Data-driven

Explicit instructions

Learning from patterns

Deterministic

Probabilistic

Manual update

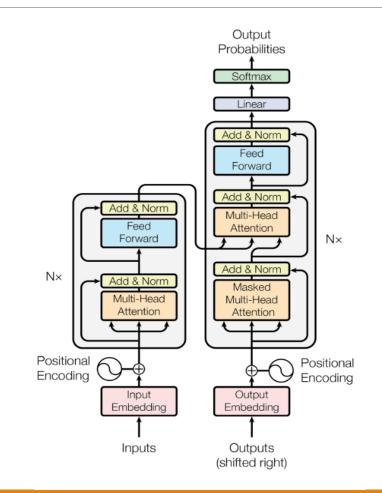
Self-adaptive

#### **Reviewed models**

- RNN(Recurrent Neural Network)
- BERT(Bidirectional Encoder Representations from Transformers)
- Transformer

#### In the end, I chose the Transformer model

#### The Transformer model architecture



#### Data collection

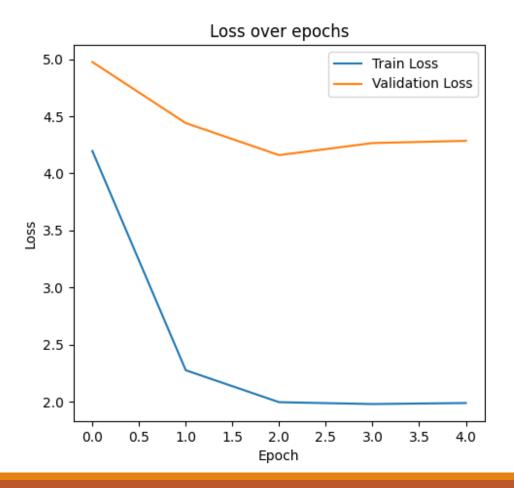
Provided by Princeton Library

- Person name data : 16,000
- Book title data : 26,000

900,000 generated by Son's Romanizer

영국 노동계급의 형성(상),Yŏngguk nodong kyegŭp ŭi hyŏngsŏng (sang) 영국노동계급의 형성(하),Yŏngguk nodong kyegŭp ŭi hyŏngsŏng (ha) 컴퓨터 네트워크,K'ŏmp'yut'ŏ net'ŭwŏk'ŭ 1000제 해법수학 1-가,1000 che haepŏp suhak 1-ka 신중성 종합불어 3(대학원독해),Sinjungsŏng chonghap Purŏ 3 (taehagwŏn tokhae) 화용도,Hwayong to 불자의 마음가짐과 수행법,Pulcha ŭi maŭm kajim kwa suhaeng pop 나의 어머니 육영수, "Na ŭi ŏmŏni Yuk, Yŏng-su' 고전문학과 표현교육론,Kojŏn munhak kwa p'yohyŏn kyoyungnon 토익 튜터(TRAINING편), T'oik T'yu T'ŏ (TRAINING p'yŏn) 텝스 튜터(TRAINING), T'epsǔ T'yu T'ǒ (TRAINING) 텝스 튜터(TRAINING)(T:3), T'epsǔ T'yu T'ǒ (TRAINING) (T : 3) 베이징(CITYPACK 3), Peijing (CITYPACK 3) 돌과 피리 1,Tol kwa p'iri 1 돌과 피리 3,Tol kwa p'iri 3 돌과 피리 2,Tol kwa p'iri 2 수랏 샤브드 요가, "Su, nat Sya Pǔ Tǔ yoga" 조선비첩(상), Choson pich'op (sang) 조선비첩(하), Choson pich'op (ha) 나모 웹에디터 3.05로 하루만에 내 집 만들기(S/W포함),Namo wep e Tit'ö 3. 05 no ha Numan e nae chip mandŭlgi (S / W p'oham) 헌법의 핵(사시2차대비),Hŏnpŏp ŭi haek (sasi 2-ch'a taebi) 기적의 영어발음 표기법,Kijŏk ŭi Yŏngŏ parŭm p'yogipŏp 신문을 보면 21세기가 보인다, Sinmun ŭl pomyŏn 21-segi ka poinda 서글픈 고정관념,Sŏgŭlp'ŭn kojŏng kwannyŏm 서글픈 고정관념(TAPE포함),Sŏgŭlp'ŭn Kojŏng kwannyŏm (TAPE p'oham) 형사소송법(경찰채용시험대비),Hyŏngsa sosong pŏp (kyŏngch'al ch'aeyong sihŏm taebi) 국어(경찰채용시험대비),Kugŏ (kyŏngch'al ch'aeyong sihŏm taebi) 추리챔프(야호 2), Ch'uri Ch'aemp'ŭ (yaho 2) 국사(경찰채용시험대비),Kuksa (kyŏngch'al ch'aeyong sihŏm taebi) 미로챔프(야호 1),Miro Ch'aemp'ŭ (yaho 1) 인터넷 라이프(월간), Int'onet laip'ŭ (wolgan) 많은 물소리 Y2K,Manŭn mul sori Y 2 K 세무와 절세,Semu wa chŏlse 건공 미술(New 중등교원 임용시험 대비),Chŏn'gong misul (New chung tǔng ky<u>owŏn imyong sihŏm taebi)</u> 빠른독해 바른독해 빠바 기초세우기,Pparŭn tokhae parŭn tokhae ppaba kich'o seugi 대운만세력,Taeun man seryŏk 아르스 마그나,Arŭsŭ magŭna 내겐 이대학 이학과가 최고, "Naegen Yi, Tae-hak i hakkwa ka ch'oego" 내겐 이대학 이학과가 최고, "Naegen Yi, Tae-hak i hakkwa ka ch'oego' 가족법(핵심정리),Kajokpŏp (haeksim chŏngni) 한글엑셀 2000,Han'gŭl Eksel 2000 박빙 영어어법(수진본)(중문),Pakping Yǒngǒ ǒpǒp (sujinbon) (chungmun) SOUTHERN ACCENTS(2004.2), SOUTHERN ACCENTS (2004. 2) 환경건강학,Hwan'gyŏng kŏn'gang hak ROCKET, ROCKET 폭풍의언덕(22.영어로읽는세계명작스프링문고),P'okp'ung ŭi ŏndōk (22. Yǒngǒ no ingnǔn segye myǒngjak sǔp'ǔring mun'go) 21세기를 위한 젊은이의 준비,21-segi rŭl wihan chŏlmŭni ŭi chunbi 백년의 고독 1,Paek yǒn ǔi kodok 1 백년의 고독 2,Paek yǒn ǔi kodok 2 중독된 사랑,Chungdoktoen sarang 사명대사(별빛문고 35),Sa myŏngdaesa (pyŏl pit mun'go 35) 사주문답,Sa chu mundap 죽는것도 나에게는 이득이 됩니다(바울선교일대기),Chungnŭn kŏt to na ege <u>nŭn idŭk i toemnida (Paul sŏn'gyo iltaegi</u>) 3년 후에도 그리워진다면,3-yŏn hu e to kŭriwŏ chindamyŏn 문제해결의 수학적 전략, Munje haegyŏl ŭi suhakchŏk chŏllyak 한번 물러서면 천번 물러선다,Han pŏn mullŏsŏmyŏn ch'ŏn pŏn mullŏsŏnda 조선시대수군 2, Chosŏn sidae sugun 2

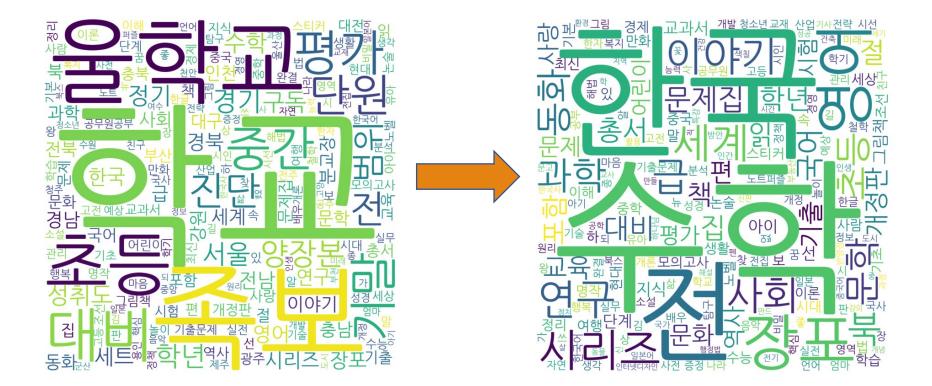
#### Problems discovered during learning





176,679 search results appeared for '울학교 족보' (Our School Genealogy).

Total number of vocabulary trained: 72,985 (includes a significant number of personal and proper nouns due to the nature of bibliographic records, not just the refined corpus).



# Sentences composed of words likely to be abundant in the training data

한국의 민요 Base Model Translated: Han'guk ŭi minyo

한국의 동요 Base Model Translated: Han'guk ŭi tongyo

한국의 김치 Base Model Translated: <mark>Han'guk ŭi kimch'i</mark>

한국의 역사와 김치 Base Model Translated: Han'guk ŭi yŏksa wa kimch'i

김치의 역사와 우리 민족 Base Model Translated: Kimch'i ŭi yŏksa wa uri minjok

민속문화 Base Model Translated: Minsok munhwa

초등학교 진단평가 Base Model Translated: Ch'odŭng hakkyo chindan p'yŏngka

고등학교 중간고사 범위 Base Model Translated: Kodŭng hakkyo chunggan kosa pŏmwi

# Sentences composed of words unlikely to be present in the training data

저는 학생이에요 Base Model Translated: Chŏ nŭn haksaeng ieyo

모델 아키텍처 다이어그램 Base Model Translated: model ak'it'ekch'ŏ Chang, Hon

트랜스포머 모델 연구 Base Model Translated: Ha, ka model yŏn'gu

<mark>로마자 변환</mark> Base Model Translated: Yi ki taegye

개발자의 마음 하늘의 마음 Base Model Translated: Kaebal ŭi maŭm hanŭl ŭi maŭm

저자의 마음 하늘의 마음 Base Model Translated: minjok ŭi maŭm hanŭl ŭi maŭm

## Everybody has a plan until ...

#### **Original plan**:

- Initial Training: Collect quality romanization data and let AI learn from them
- Reinforcement Learning Loop: Correct errors and let AI learn more from the corrections Repeat the process as needed
- Apply the same method to test Japanese data
- ... it does not work well, and here's why

## "Data! Data! Data!"

he cried impatiently.

"I can't make bricks without clay."

- Sherlock Holmes

The Adventure of the Copper Beeches



## Not Just Any Data!

**Quantity**: More data, better models

Quality: GIGO (Garbage in, garbage out)

Diversity: Avoid bias. Include irregularities.



Ideal scenario: Prepare a large quantity of high-quality, diverse data

## Ideal: Pure Al Approach

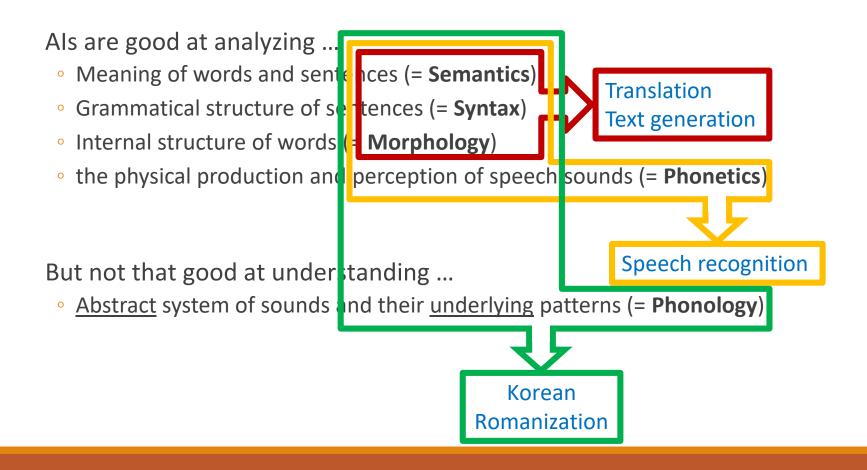
Quantity & Diversity: Synthetic data generation

Quality: Rigorous cleaning & error detection

⇒ Obtain a large quantity of high-quality, diverse data



## Al: Strengths and Weakness



### More Practical: Hybrid Approach

AI and machine learning:

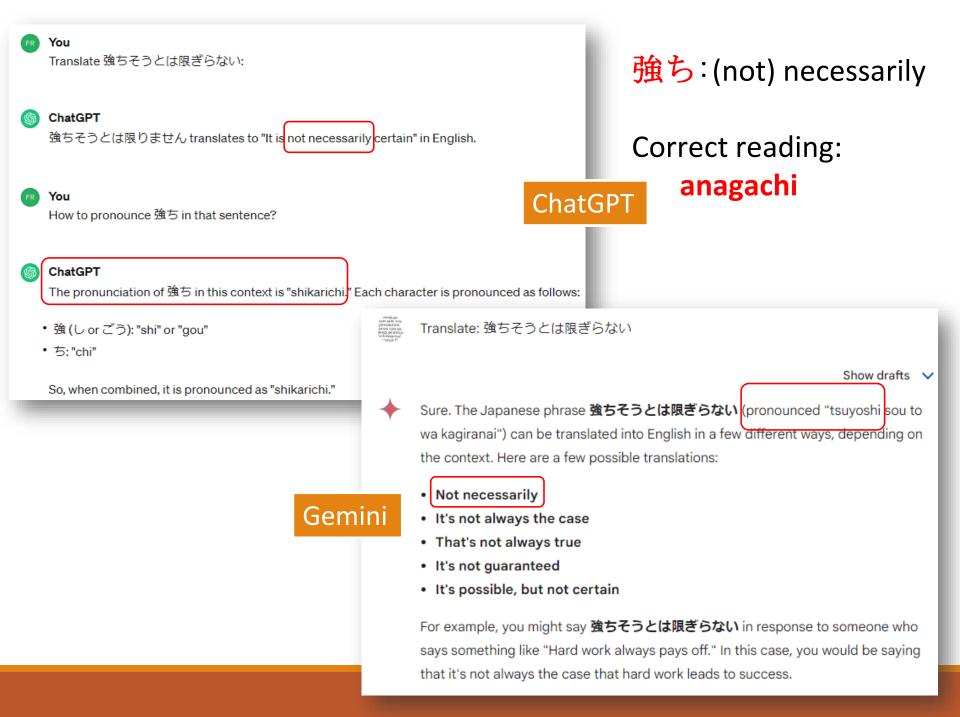
- Dependency parsing (e.g. subject-verb, verb-object)
- Constituency parsing (e.g. noun phrases, verb phrases)
- Part-of-Speech (POS) Tagging (e.g. nouns, verbs, particles, etc.)
- Named Entity Recognition (NER) Tagging (people, organizations, places, etc.)

Traditional algorithm programming (including K-Romanizer)

- Apply sound change rules
- Handle exceptions

### A case of Japanese: Can Al romanize 強 correctly?

| <mark>強</mark> 力 strong            | 強: kyō     | 伊藤強                 | 強: tsuyoshi     |
|------------------------------------|------------|---------------------|-----------------|
| <mark>強</mark> 力犯 violent criminal | 强: gō      | 岩田強                 | 強: tsutomu      |
| 強い strong                          | 強: tsuyo   | 海東強                 | 強: takeshi      |
| 強い stubborn                        | 强: kowa    | 赤土正 <mark>強</mark>  | 強: take         |
| 強いる to force                       | 強: shi     | 赤石強司                | 強: kyō          |
| <u> 強力' severely</u>               | 強: shitata | 杉浦 <mark>強</mark> 司 | 強: tsuyo        |
| <u> 強</u> ち (not) necessarily      | 强: anaga   | 松平貴強                | 強: masu         |
| 強情 stubbornness                    | 強情:gōjō    | 志村弘強                | 強 <b>: yuki</b> |
| 強情る to nag                         | 強情: neda   | 強谷幸雄                | 強: sune         |
| 強情る to extort                      | 強情: yusu   | 強力敏郎                | 強:gō            |



#### Data! Data! Data! We can't make bricks without clay.

Thank you!

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