

Translating through Transliteration The Case of Tajik Persian

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Background

• The Persian language is written in two different scripts



- Mutual intelligibility between standard dialects is high in spoken form, but falls to zero in written form
- Tajikistan, a country of ~10 million, cannot access written media from the greater Persian-speaking world (~100 million people
- Roughly 2.2% of the Internet is written in Persian
- Less than 0.1% is written in Cyrillic, the rest is in Arabic
- The scripts do not have a simple one-to-one correspondence, obfuscating typical transliteration
- Can a model be trained to "translate" between the two dialects through transliteration?

Method

- Previous Work:
- Proposed a statistical model for machine transliteration, but lacked a true parallel corpus with which to fully verify model performance (Davis, 2012)
- Model:
- Neural network-based Grapheme-to-Phoneme (G2P)
- Why G2P:

Challenges

• Script Comparison

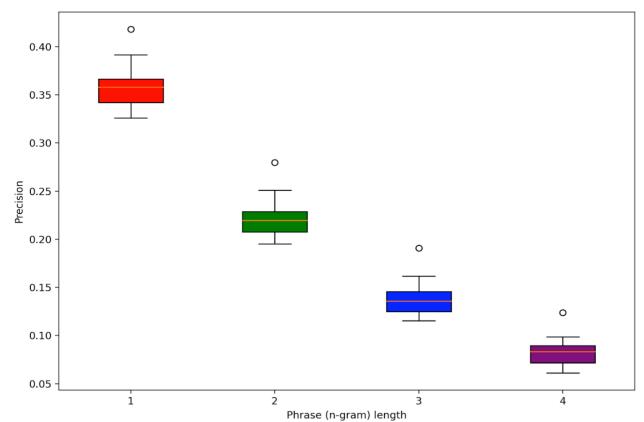
- The Perso-Arabic script is an abjad
- Vowels are often unwritten, and sometimes ambiguous
- The speaker must know how to pronounce the word already
- The Tajik-Cyrillic script is an alphabet
- All sounds are (generally) written as they are pronounced
- The speaker does not require prior knowledge to learn how to pronounce a word
- Case Sensitivity
- The Arabic script does not implement case, while Cyrillic does
- When converting from Arabic to Cyrillic, case must be inferred

• Unwritten Grammatical Particle: "Ezâfe"

- The "Ezâfe" links two words together, and can be used to denote: possession, adjective-noun relationships, noun linkage, and given name
- Despite being so common, it is often unwritten in Perso-Arabic text, but always written in Tajik-Cyrillic
- When transliterating from Arabic to Cyrillic, the location of the "Ezâfe" must be inferred and inserted where necessary
- Non-bijective Alignment and Letter Ambiguity
- Several syllables and letters have one rendering in Cyrillic, but several in Arabic
- G2P models are typically used in Text-to-Speech (TTS) systems, converting graphemes (letters) to phonemes (pronunciations)
- Typical transliteration models do exist, but G2P may be more suited to this task, as it greatly resembles TTS
- The Arabic standard does not accurately represent pronunciation, but the Cyrillic standard does
- We seek to apply such a model (Yolchuyeva et al., 2020) in one direction: Arabic (Grapheme) to Cyrillic (Phoneme)
- **Corpus:** the <u>very first</u> aligned digraphic Persian corpus, manually collected from blogs and articles online
- ~5400 sentences, ~42,000 words

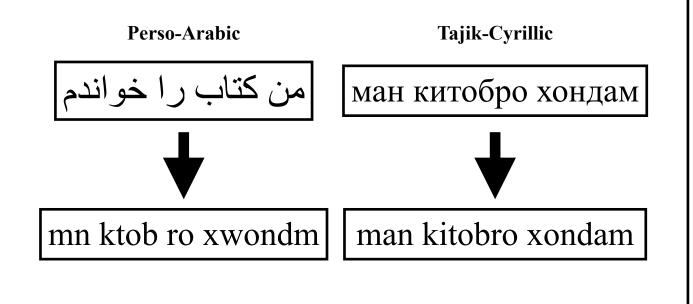
Results and Conclusion

- Model Hyperparameters:
- Learning Rate: 0.00044, Dropout Rate: 0.2
- Individual Word Error Evaluation:
- 39.2% of words predicted correctly
- When including predictions 1 and 2 edit-distances away, this becomes 66.7% and 82.2%, respectively



BLEU Score Evaluation

• When transliterating from Cyrillic to Arabic, the correct option must be chosen



| | Script | Example Sentence (errors marked in red) |
|---|---|--|
| | Arabic | امروز ناظران بین المللی کنفرانس مطبو عاتی در شهر دوشنبه برگزار مینمایند |
| | Cyrillic (Expected) | Имруз нозирони байналмиллали конфронси матбуоти дар шахри Душанбе баргузор менамоянд |
| y | Cyrillic (Predicted) | имруз нозарон беин лмлй канфаронс матбуъотй дар шахр душанбе баргузор маинаминд |
| | Analy So V N | heir counterpart in the reference translation ysis ome vowels successfully predicted, others unsuccessfully owel insertion partially successful Iodel proves unable to detect ezâfe |
| | | 1 |
| | • F | Plusion 2P approach presents a viable approach to transliterating ersian from Arabic to Cyrillic urther improvements required before our model becomes sable |

• Supplement corpus with manually-added "ezâfe" tags

• Continue hyperparameter testing



Chris Irwin Davis. 2012. Tajik-Farsi Persian Transliteration Using Statistical Machine Translation. In *Proceedings of the Eighth International Conference on Language Resources and Evaluation (LREC'12)*, pages 3988–3995, Istanbul, Turkey. European Language Resources Association (ELRA).

Yolchuyeva, S., Németh, G., & Gyires-Tóth, B. 2020. Transformer based grapheme-to-phoneme conversion.

Papineni, K., Roukos, S., Ward, T., & Zhu, W. J. 2002. BLEU: A Method For Automatic Evaluation Of Machine Translation. In Proceedings of the 40th annual meeting on Association for Computational Linguistics.

