

Learning Meters of Arabic and English poems

With Recurrent Neural Networks

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Introduction

فَقُولُ رَسُولِ اللَّهِ أَزْكَى وَأَشْرَحُ

وَدَعْ عَنْكَ آرَاءَ الرِّجَالِ وَقَوْلَهُمْ

But ... What is poetry?

General Definition:

- **Poetry** is a piece of writing or speaking, which **MUST** follow specific **Patterns**.

Example, *English verse*:

That **time** of **year** thou **mayst** in **me** behold

To detect poems' meters, we need to learn those **Patterns**.

- **Foot** التفعيلة: is a sequence of vowels and consonants.

Feet	Scansion
فَعُولُنْ	0/0//
فَاعِلُنْ	0//0/
مُسْتَفْعِلُنْ	0//0/0/
مَفَاعِلُنْ	0/0/0//
مَفْعُولَات	0//0///
فَاعِلَاتُنْ	0/0//0/
مُفَاعِلَتُنْ	0///0//
مُتَفَاعِلُنْ	0//0///

Arabic Patterns/Meters بحور الشعر:

- **Meter** البحر: is a sequence of **feet**.

Meter Name	Meter <i>feet combination</i>
<i>al-Wafer</i>	مُفَاعَلَتْنِ مُفَاعَلَتْنِ فَعُولُنْ
<i>al-Taweel</i>	فَعُولُنْ مَفَاعِيلُنْ فَعُولُنْ مَفَاعِيلُنْ
⋮	⋮
<i>al-Moktadib</i>	مَفْعُولَاتُ مُسْتَفْعِلُنْ مُسْتَفْعِلُنْ
<i>al-Modar'e</i>	مَفَاعِيلُنْ فَاعِلَاتُنْ مَفَاعِيلُنْ

Arabic Prosody, example!

From بحر الوافر:

ويسأل في الحوادث ذو صوابٍ		
ويسأل فل	حوادث ذو	صوابين
0///0//	0///0//	0/0//
مفاعلتن	مفاعلتن	فعولن

English Prosody

English Meters Building Blocks:

- Syllables: /'wɔ:tə/ = /'wɔ:/ + /tə(r)/.
 - **stressed** + unstressed.
- Foot: is a combination of stressed and unstressed syllables.

Feet	Stresses Combination
<i>Iamb</i>	×/
<i>Trochee</i>	/×
<i>Dactyl</i>	/××
<i>Anapest</i>	××/
<i>Pyrrhic</i>	××
<i>Amphibrach</i>	×/×
<i>Spondee</i>	//

Meter: is repeating a foot n times; where $n \in [1, 8]$.

Iambic pentameter verse:

That **time** of **year** thou **mayst** in **me** be**hold**.
Iambic Foot 2nd 3rd 4th 5th

Literature Review

Abuata and Al-Omari:

- Five-step Algorithm
 1. Getting the input, carrying full diacritics.
 2. Metrical scansion rules are applied to the Arud writing. 0/0/..
 3. Grouping zero and ones to feet **تفعيلات**.
 4. A class is assigned to the input.
- **Results:** 82.2% of 417 verses.

Alnagdawi et al, similar approach; Context-Free Grammar; 75% correctly classed from 128.

ويسأل في الحوادث ذو صوابٍ
ويسأل فل حوادث ذو صوابين
0///0// 0///0// 0///0//
مفاعلتن مفاعلتن مفاعلتن

Issues;

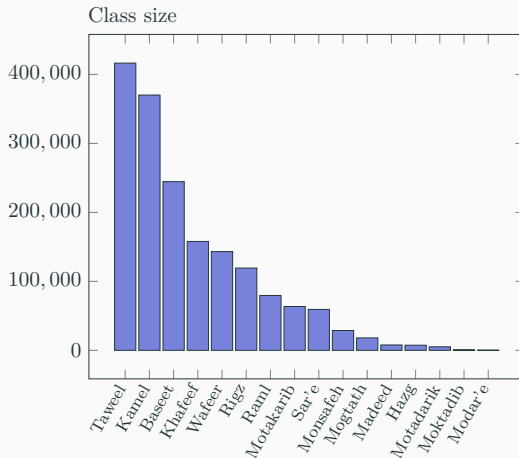
- A huge constrain. **Diacritics** are a must.
- Converting the text into pronounced text is **probabilistic**.
 - اثبات الحروف المحذوفة خطأً
 - التصرف فى التقاء الساكنين

Binary Classification; Metric or Free-Verse:

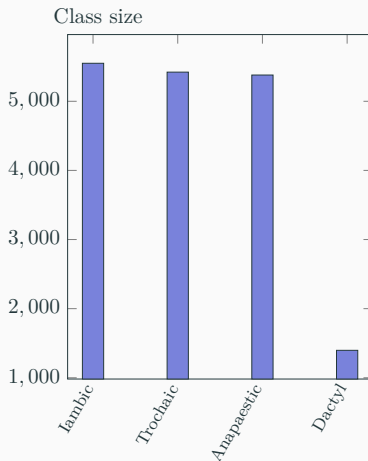
- verses are represented as vectors of statistical features.

Datasets

Arabic Dataset:



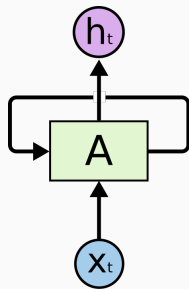
English Dataset:



Methodology

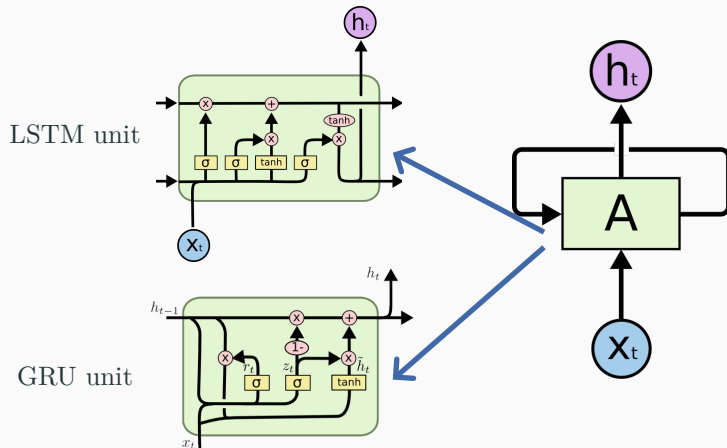
Which Network!

- **Pattern:** is a sequence of characters.
- Unlike feedforward neural networks, RNNs can use their internal state (memory) to process sequences of inputs.



Rolled Rnn unit

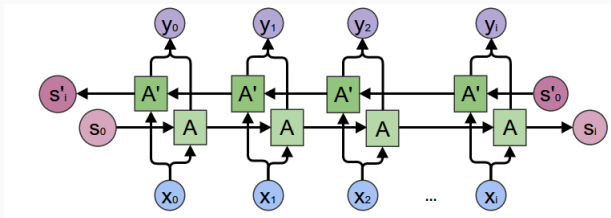
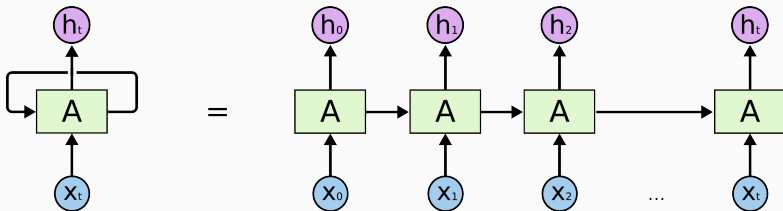
RNN, Architectures



- Two variants of unidirectional *recurrent units*.

RNN, Architectures

Unidirectional & Bidirectional RNN:



An Issue:

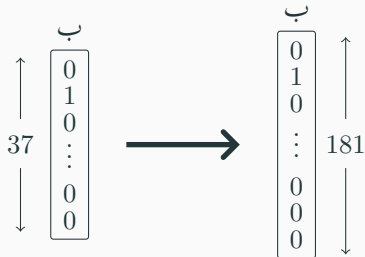
- Diacritics are standalone characters!
 - $\text{len مرحبا} \neq \text{len مَرْحَبًا}$
 - We have represented the letter and its diacritic as a **one character**.

Benefits:

1. Verse's length is fixed, regardless the diacritic states.
2. Saving more space, by shorten the length of full diacritic verses.
3. Models can be tested on both diacritic or non-diacritic data.

Encoding Techniques

1. One-Hot
2. Binary
3. Two-Hot (new technique)

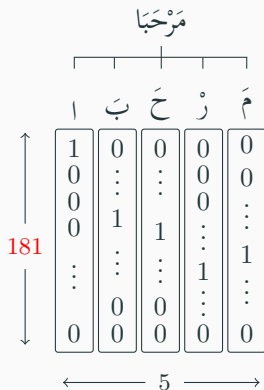


One-Hot Vector: from 37×1 to 181×1

181 is the number of all combination between letters and diacritics.

$$181 = 36 + 36 \times 4 + 1$$

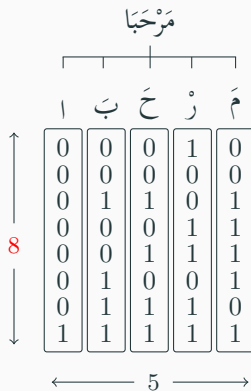
One-Hot, example



Binary

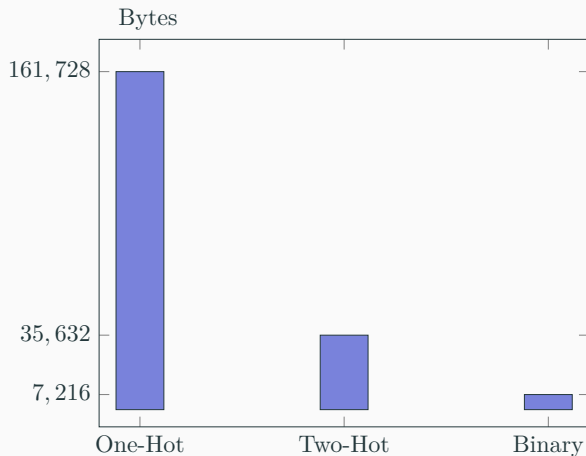
Let n be the vector length.

$$n = \lceil \log_2 l \rceil \quad l \in \{181, 28\}$$



$$\begin{array}{c}
 \begin{array}{c} \updownarrow \\ 37 \\ \updownarrow \end{array} \begin{array}{|c|} \hline \text{ح} \\ \hline 0 \\ 1 \\ 0 \\ 0 \\ \vdots \\ 0 \\ \hline \end{array} \\
 m
 \end{array}
 +
 \begin{array}{c}
 \begin{array}{c} \updownarrow \\ k \\ \updownarrow \end{array} \begin{array}{|c|} \hline \text{و} \\ \hline 1 \\ 0 \\ 0 \\ 0 \\ \hline \end{array} \\
 k
 \end{array}
 =
 \begin{array}{c}
 \begin{array}{c} \updownarrow \\ 41 \\ \updownarrow \end{array} \begin{array}{|c|} \hline \text{ح} \\ \hline 1 \\ 0 \\ 0 \\ 0 \\ 0 \\ 1 \\ 0 \\ \vdots \\ 0 \\ \hline \end{array} \\
 \left. \begin{array}{l} \text{4} \times 1 \text{ diacritic vector} \\ \text{which represents } \text{و} \end{array} \right\} \\
 \left. \begin{array}{l} \text{37} \times 1 \text{ letter vector} \\ \text{which represents } \text{ح} \end{array} \right\}
 \end{array}$$

Space Comparison



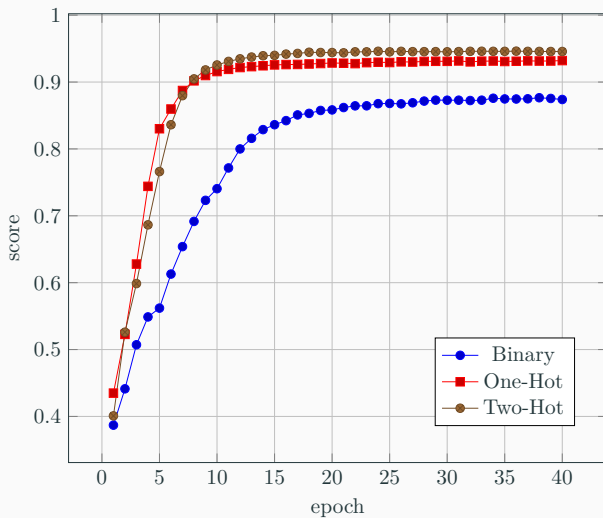
Results

Arabi Results

#	data size	encoding	diacritic	archit.	f1
1	full data	two-hot	Yes	7L, 50U, 0	95.79%
2	full data	two-hot	No	7L, 50U, 0	95.43%
3	full data	binary	Yes	7L, 81U, 0	95.51%
4	full data	binary	No	10L, 30U, 0	93.2%
5	full data	one-hot	Yes	7L, 50U, 1	95.32%
6	full data	one-hot	No	7L, 82U, 0	93.94%
7	eliminated	two-hot	Yes	7L, 81U, 1	95.88%
8	eliminated	two-hot	No	4L, 50U, 1	96.29%
9	eliminated	binary	Yes	7L, 81U, 1	94.87%
10	eliminated	binary	No	4L, 82U, 0	96.38%
11	eliminated	one-hot	Yes	7L, 75U, 0	95.65%
12	eliminated	one-hot	No	7L, 50U, 0	95.04%

id	encoding	cell type	f1 test
1	one-hot	GRU	81.35%
2	one-hot	LSTM	80.34%
3	binary	LSTM	75.43%
4	binary	GRU	75.04%

Encoding Effect



Binary Encoding Problem

ع	م
0	0
0	0
1	1
0	1
1	1
1	1
0	0
1	1

Questions?