DNA classification using supervised deep learning

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Problem statement

• Deoxyribonucleic acid or short DNA, is the basis of how life works

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- Ancient DNA might reveal crucial information regarding past civilizations, past diseases or even extinct speecies
- Ancient DNA is subject to contamination with modern DNA
- Our aim: classify ancient and modern DNA
- Four data representation and two learning approaches



Computational approaches to DNA analysis

Challenge: find a comprehensive and robust representation for DNA.

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- One hot encoding
- Images
- Deep learning

Methodology

DNA sequence: [A,C,G,T]

Four different DNA representations :

 Nucleotides frequencies based representation

•
$$P(s) = \frac{frequence(s)}{l-s_l+1}$$

• 84 features

| DNA sub_sequence | Probability of occurrence | | | |
|------------------|---------------------------|--|--|--|
| A | 0.(2) | | | |
| C | 0.(3) | | | |
| G | 0.(1) | | | |
| Т | 0.(2) | | | |
| AA | 0 | | | |
| AC | 0.125 | | | |
| | | | | |
| TT | 0 | | | |
| AAA | 0 | | | |
| | | | | |
| ACT | 0.142 | | | |
| | | | | |
| CGC | 0.142 | | | |
| | | | | |
| TTT | 0 | | | |

Figure: DNA representation: example of features and their values for the illustrative sequence *ACTCGCTA*.

Methodology

TF-IDF based representation

• $TF - IDF(s) = \frac{frequence(s)}{l-s_l+1} * \log \frac{k}{n}$ Example of TF-IDF representation for the sequence ACGGTAACGGTG,considering the coprus ACGGTAACGGTG, TTGCCTGTGCATGA, ACCGGTTCAACGTGCAAAACGCG-CACCGC.

| DNA sub_sequence | TF-IDF weight | | | |
|------------------|---------------|--|--|--|
| AA | 0.0531 | | | |
| AC | 0.106 | | | |
| AG | 0.0 | | | |
| | | | | |
| CG | 0.106 | | | |
| GG | 0.106 | | | |
| | | | | |
| TA | 0.144 | | | |
| | | | | |
| AAC | 0.058 | | | |
| | | | | |
| ACG | 0.116 | | | |
| | | | | |
| CGG | 0.116 | | | |
| GGT | 0.116 | | | |
| GTA | 0.158 | | | |
| | | | | |
| TAA | 0.158 | | | |

Methodology

• Physical and chemical properties based representation

| Property name | Α | С | G | Т |
|--------------------------------|--------|-------|--------|--------|
| Molecular weight | 135.13 | 111.1 | 151.13 | 126.11 |
| Molecular density | 1.6 | 1.55 | 2.2 | 1.23 |
| Topological Polar Surface Area | 80.5 | 67.5 | 96.2 | 58.2 |
| Heavy Atom Count | 10 | 8 | 11 | 9 |
| Complexity | 127 | 170 | 225 | 195 |

TABLE I: Values representing measurable physical and chemical properties of nucleotides.

• One hot encoding

| DNA sub_sequence | Encoding | | | |
|------------------|---------------------|--|--|--|
| ACT | [1. 0. 0. 0. 0. 0.] | | | |
| CTC | [0. 0. 0. 1. 0. 0.] | | | |
| TCG | [0. 0. 0. 0. 0. 1.] | | | |
| CGC | [0. 1. 0. 0. 0. 0.] | | | |
| GCT | [0. 0. 0. 0. 1. 0.] | | | |
| CTA | [0. 0. 1. 0. 0. 0.] | | | |

Table 2: DNA representation: example of one hot encoding for the illustrative sequence *ACTCGCTA*. We used window length = 3 and slide value = 1.

Methodology

Classification problem

- Non linear models
- Multi-layer Perceptron
 - 2 hidden layers, ReLU, adaptive learning rate
- Convolutional neural network

| Layer (type) | Output | Shape | Param # |
|------------------------------|--------|--------------|---------|
| conv2d_1 (Conv2D) | (None, | 300, 62, 64) | 640 |
| max_pooling2d_1 (MaxPooling2 | (None, | 150, 31, 64) | 0 |
| conv2d_2 (Conv2D) | (None, | 74, 15, 32) | 18464 |
| max_pooling2d_2 (MaxPooling2 | (None, | 37, 7, 32) | 0 |
| conv2d_3 (Conv2D) | (None, | 18, 3, 16) | 4624 |
| flatten_1 (Flatten) | (None, | 864) | 0 |
| dense_1 (Dense) | (None, | 200) | 173000 |
| dense_2 (Dense) | (None, | 2) | 402 |
| activation_1 (Activation) | (None, | 2) | 0 |

Figure: Model summary of the convolutional neural network

Dataset

• Matter collected from Capidava archaeological site

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- 378.451 ancient sequences
- 115.218 modern sequences
- MLP k- folds cross validation, k=10
- CNN 66 % training, 33 % testing

Results

| Representation | Property type | ANN | | | CNN | | |
|--|------------------------------|----------|----------|--------|----------|----------|--------|
| | | Accuracy | F1-score | Recall | Accuracy | F1-score | Recall |
| Frequency based | - | 0.937 | 0.861 | 0.833 | - | - | - |
| TF-IDF | - | 0.912 | 0.944 | 0.960 | - | - | - |
| Physical and chemical properties | Molecular weight | 0.918 | 0.819 | 0.778 | - | - | - |
| | Density | 0.941 | 0.873 | 0.846 | - | - | - |
| | Topological polar surface | 0.935 | 0.859 | 0.829 | - | - | - |
| | Heavy atom count | 0.928 | 0.843 | 0.805 | - | - | - |
| | Complexity | 0.935 | 0.862 | 0.836 | - | - | - |
| One hot encoding | - | - | - | - | 0.9086 | 0.8697 | 0.8697 |

 $\ensuremath{\mathsf{Figure:}}$ Evaluation measures for the two supervised models and the considered representations

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Conclusion

• The aim was to find suitable representation and machine learning models to the goal of distinguishing between ancient and modern bacterial DNA

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- Obtained results are promising
- Acquire more modern DNA sequences
- Perform these experiments on other public dataset
- Other approaches

Thank you for your attention!