# Interview-talk for Student Assistant position

at XAI Group at Fraunhofer HHI

## **Ivan Timofeev**

#### About me

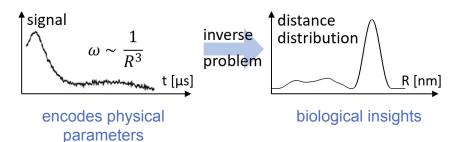
- 4<sup>th</sup> semester MSc Data Science (Freie Universität Berlin)
- BSc + MSc Physics
- 6 years in experimental research

## Plan for the talk

- Previous research in bio-chemical physics
- Data Science experience
- Recent project on XAI

## Research in bio-chemical physics

Studied properties of biomolecules
using electron magnetic resonance
spectroscopy

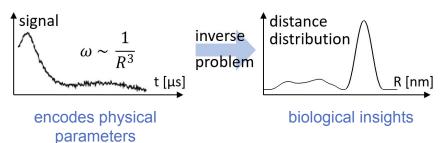


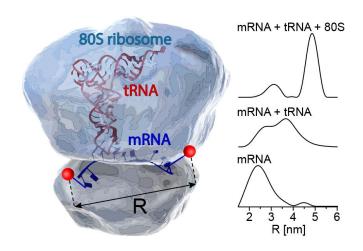
# Research in bio-chemical physics

Studied properties of biomolecules
using electron magnetic resonance
spectroscopy



- Ribosomal complexes
- Boosting reliability (regression, Monte Carlo simulations, neural network)
- Method development
- 11 papers (4 as primary contributor)





# Experience in data science

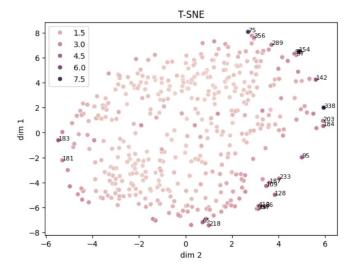
- 3 years of studying (classic ML, DL, tool proficiency)
- Implemented architectures:
  - Developed NN framework from scratch using numpy
  - Implemented ResNet for defect detection in solar cell images
  - Implemented self-supervised CV models for representation learning (BYOL, contrastive learning)
  - Implemented Transformer for text generation (Attention Is All You Need)
- Tool proficiency:
  - Tools: Unix/Linux, VCS Git, GitHub CI/CD, Docker
  - Languages: Python (PyTorch, TensorFlow), C++, MATLAB,
     Mathematica, SQL, bash

# XAI project: part 1/3

Task: identify anomalies in tabular dataset in unsupervised manner and explain using LRP

• Kernel density estimation 
$$\tilde{p}(m{x}) = \frac{1}{N} \sum_{k=1}^N \exp(-\gamma \| m{x} - m{u}_k \|^2)$$

• x' – anomaly threshold  $\tilde{p}(x) < \tilde{p}(x')$ 



# XAI project: part 1/3

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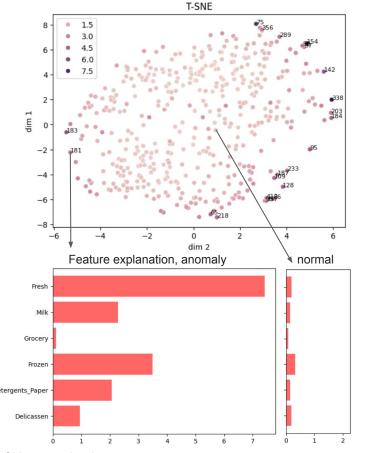
x' – anomaly threshold

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Layer-wise relevances

$$R_k^{(j)} = \frac{\exp(-\gamma z_{jk})}{\sum_{k \neq j} \exp(-\gamma z_{jk})} \cdot y_j$$

$$R_i^{(j)} = \sum_{k \neq j} \frac{[x_k - x_j]_i^2}{\|x_k - x_j\|^2} \cdot R_k^{(j)}$$



[ref] G. Montavon, J. R. Kauffmann, W. Samek, and K.-R. Müller. Explaining the Predictions of Unsupervised Learning Models, volume 13200 of Lecture Notes in Computer Science, pages 117-138. Springer, 2020

## XAI project: part 2/3

Task: predict atomization energies of molecules based on

- atomic numbers
- 3D coordinates

- Linear model with pair-encoded representation
  - $\circ$  Explanation: energy(molecule) =  $\Sigma$  energy(atomic pair)
  - Good metrics, but physically incorrect pair energies

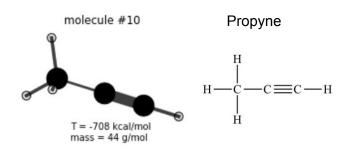
- atomization energies

# XAI project: part 2/3

Task: predict atomization energies of molecules based on

- Linear model with pair-encoded representation
  - $\circ$  Explanation: energy(molecule) =  $\Sigma$  energy(atomic pair)
  - Good metrics, but physically incorrect pair energies
- Physical constraints meaningful pair energies
- Energy of C-C bond -200regression result iterature -175-150E\_pair [kcal/mol] -125-50-251.8 2.0 2.2 2.4 2.6 2.8 3.0 3.2 3.4 3.6 distance [a.u.]

- atomic numbers
- 3D coordinates
- atomization energies

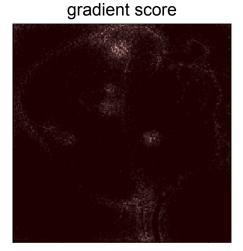


# XAI project: part 3/3

#### Task: gain insights into healthy/diseased leaf dataset

- Pretrained VGG16 + simple binary classifier
- Robust gradient-based explanation prioritizes excitatory effect over inhibitory effects in the network

label: diseased



### Final remarks

How I can contribute to the work of the XAI Group at Fraunhofer HHI:

- DL proficiency
- Expertise in multiple scientific domains
- Physics+math background
- Academic background
- + Overall highly motivated to grow in data science
- + Interested in applying and developing XAI methods
- + Open to discussing any available research topic in your group

Thank you for your attention!

# XAI project: part 1/3

$$\tilde{p}(\boldsymbol{x}) = \frac{1}{N} \sum_{k=1}^{N} \exp(-\gamma \|\boldsymbol{x} - \boldsymbol{u}_k\|^2) \qquad \tilde{p}(\boldsymbol{x}) < \tilde{p}(\boldsymbol{x}') \qquad o(\boldsymbol{x}) \triangleq -\frac{1}{\gamma} \log \tilde{p}(\boldsymbol{x})$$

$$\begin{pmatrix} h_k = \|\boldsymbol{x} - \boldsymbol{u}_k\|^2 & (\text{layer 1}) \\ o(\boldsymbol{x}) = \text{LME}_k^{-\gamma} \{h_k\} & (\text{layer 2}) \end{pmatrix}$$

$$\text{LME}_k^{\alpha} \{h_k\} = \frac{1}{\alpha} \log \left(\frac{1}{N} \sum_{k=1}^{N} \exp(\alpha h_k)\right)$$

$$R_i = \sum_k \frac{[{m x} - {m u}_k]_i^2}{\epsilon + \|{m x} - {m u}_k\|^2} R_k$$

$$R_k = \frac{\exp(-\beta h_k)}{\sum_k \exp(-\beta h_k)} \cdot o(\boldsymbol{x})$$