

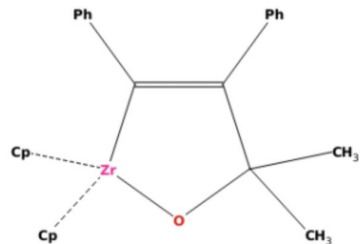
# ML in Chemistry!

## Data Analytics

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We use tech to connect human potential and  
opportunity with dignity & humility

# How does this look like?



(0D) Features

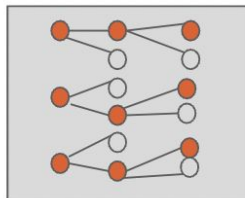
**X1**

**X2**

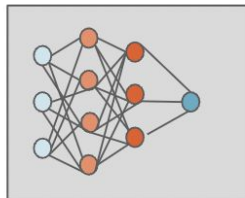
**X3**

**Xn**

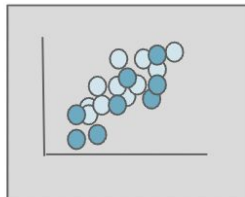
**Tree\_based**



**PyTorch\_NN**

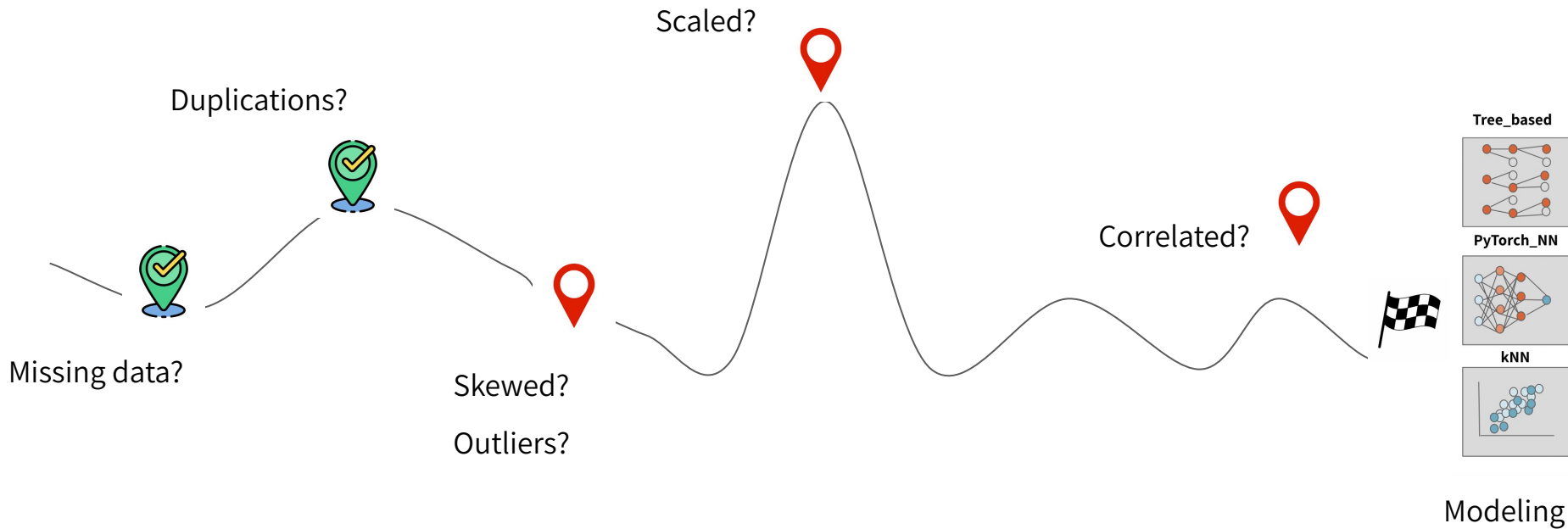


**kNN**

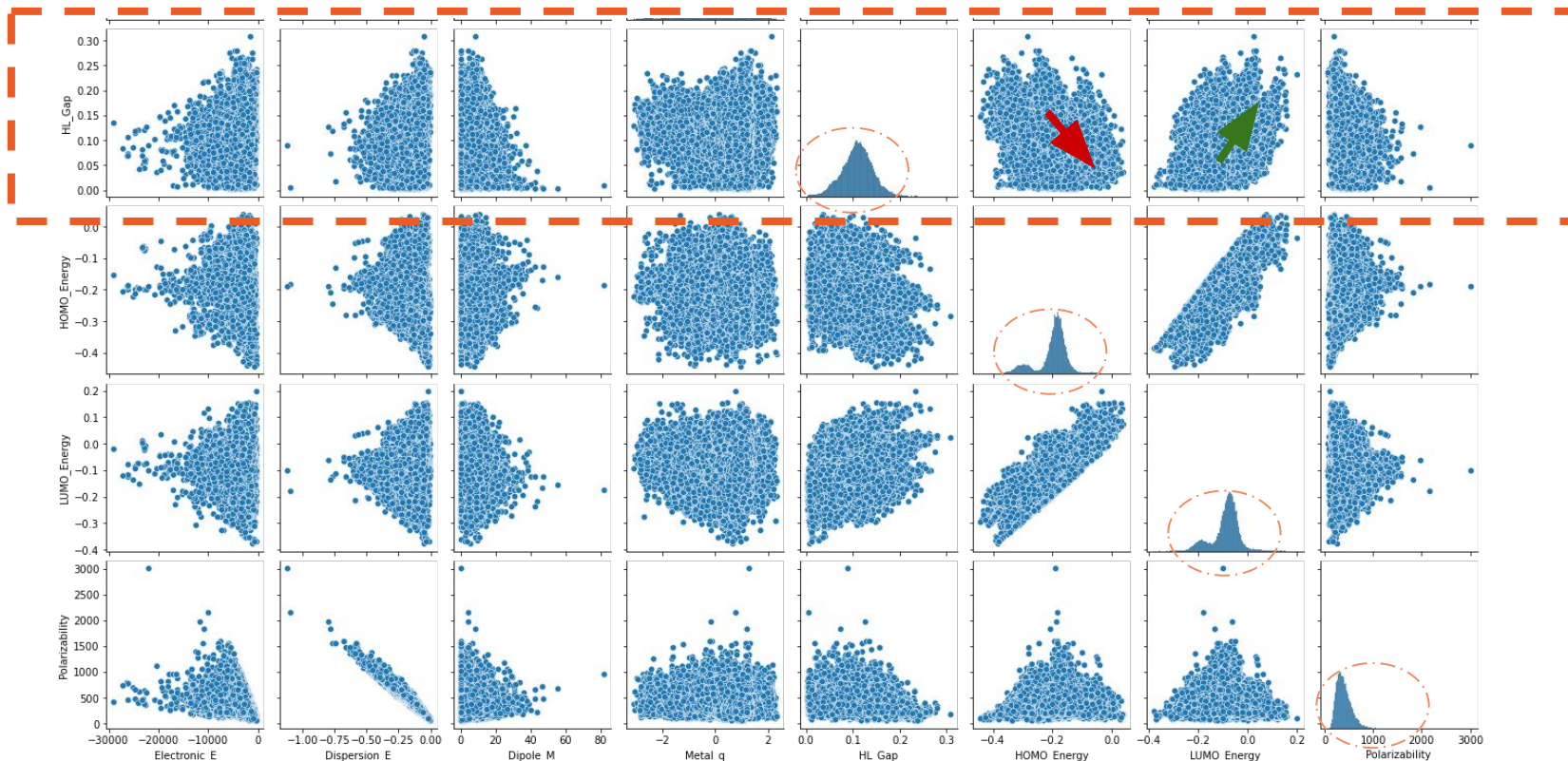


Target  
**y**

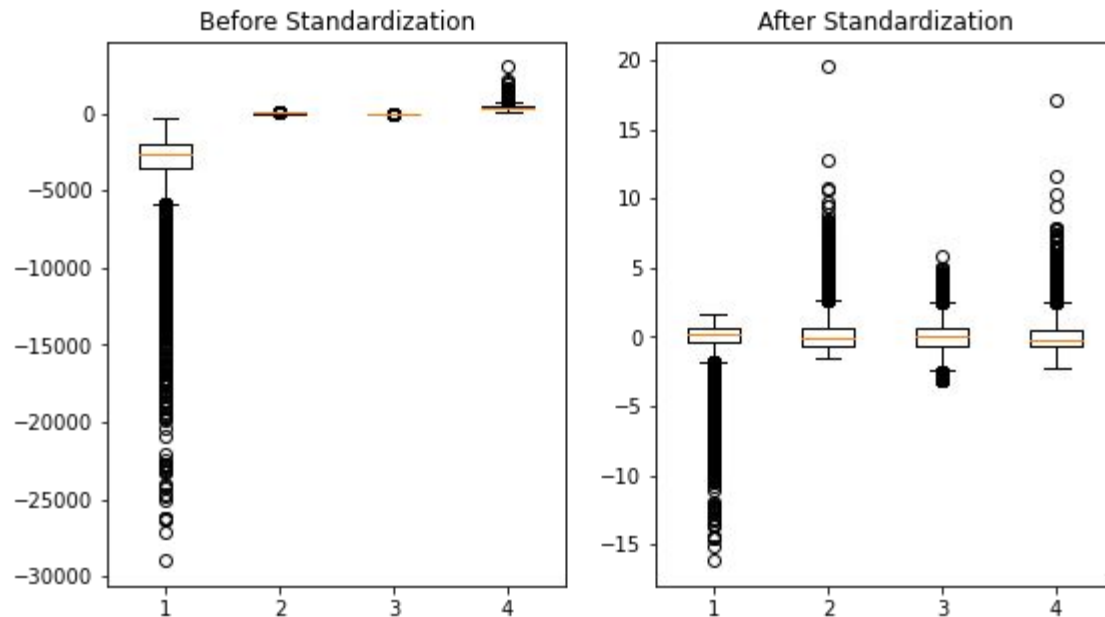
# Data Checkpoints!



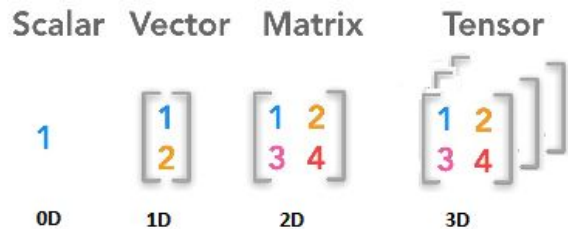
# Zooming in!



# Scaled!, Should we really bother?



# Is that all? Happy with the Dimensions?



<https://www.databricks.com/blog/2021/04/14/7-reasons-to-learn-pytorch-on-databricks.html>

```
# Convert arrays to PyTorch tensors  
x = torch.tensor(x, dtype=torch.float32)  
y = torch.tensor(y, dtype=torch.float32)
```

# Fitting | GridSearch-CV Time!

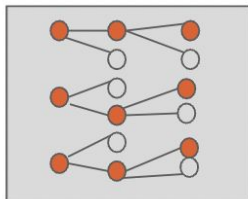
2D array  
(86665, 7)

80 20

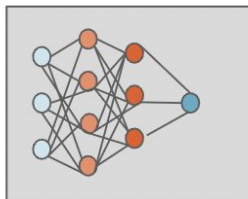
Scaled tensor-ed  
(86665, 7)

Scaled 2D array  
(86665, 7)

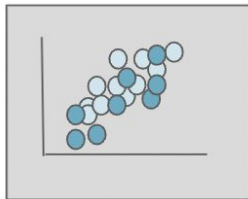
## Tree\_based



## PyTorch\_NN



## kNN



RF: n\_trees in [50, 100, 300, 500, 1000, 2000, 5000, 10000]

GBoost: {'learning\_rate': [0.1, 0.01, 0.001],  
'subsample': [0.5],  
'max\_features': [6, 7]}



Optimizer: Adam, SGD  
Lr: 0.01, 0.001

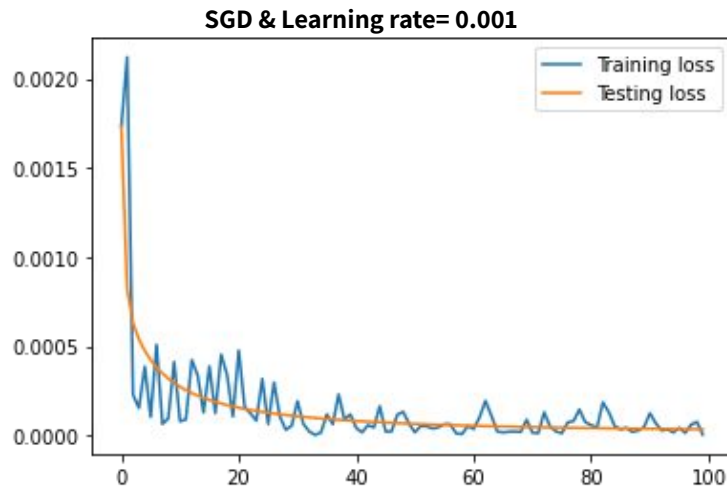
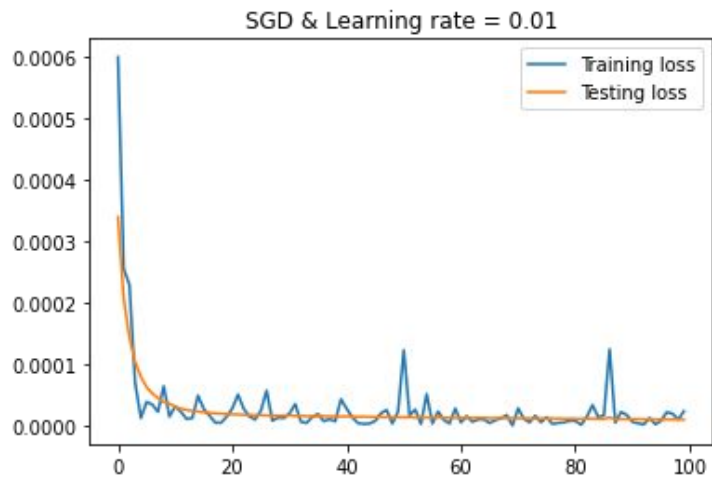
k=1, ..., 5, ..., 10

# Numbers?... Error Evaluation!

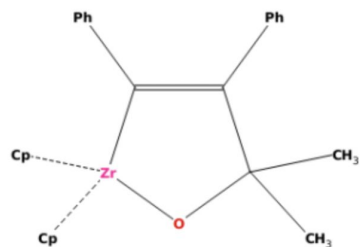
	RF	GBoost	NN_PyTorch	kNN
MAE	0.0375	0.0009		0.0064
MSE	5.362	1.389	Epoch 100 4.5907e-14	8.266
RMSE	0.0007	0.0011		0.0090
R2	0.9995	0.9987		0.927



# Graphs?



# Conclusion!



(0D) Features

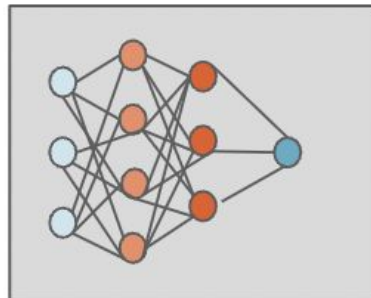
**X1**

**X2**

**X3**

**Xn**

**PyTorch\_NN**



Target  
**y**  
(HL\_GAP)



# Thanks a lot!



**Contact**



[Nesma Mousa](#)

