

POTENTIAL ENERGY SURFACES WITH ACTIVE LEARNING

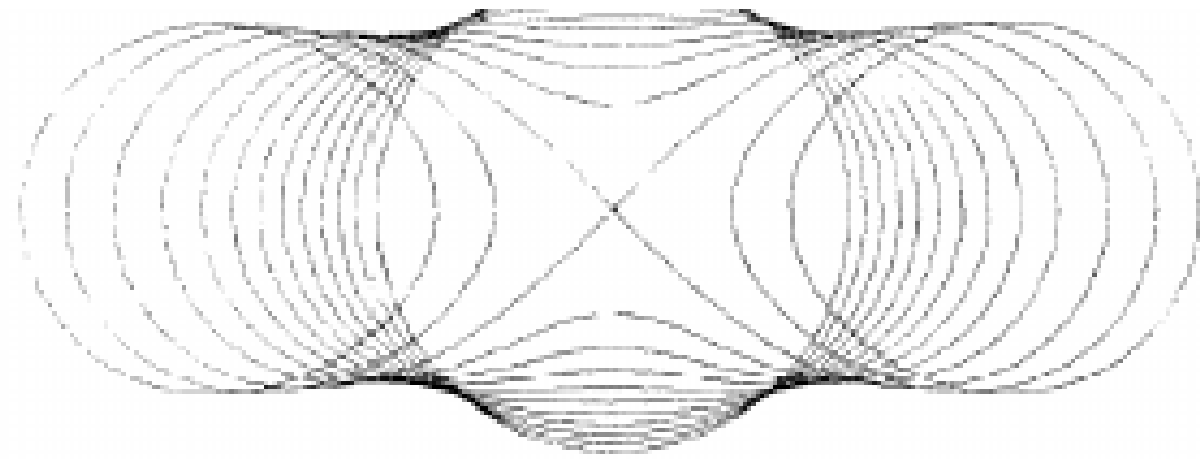
NPAC COMPUTER PROJECT

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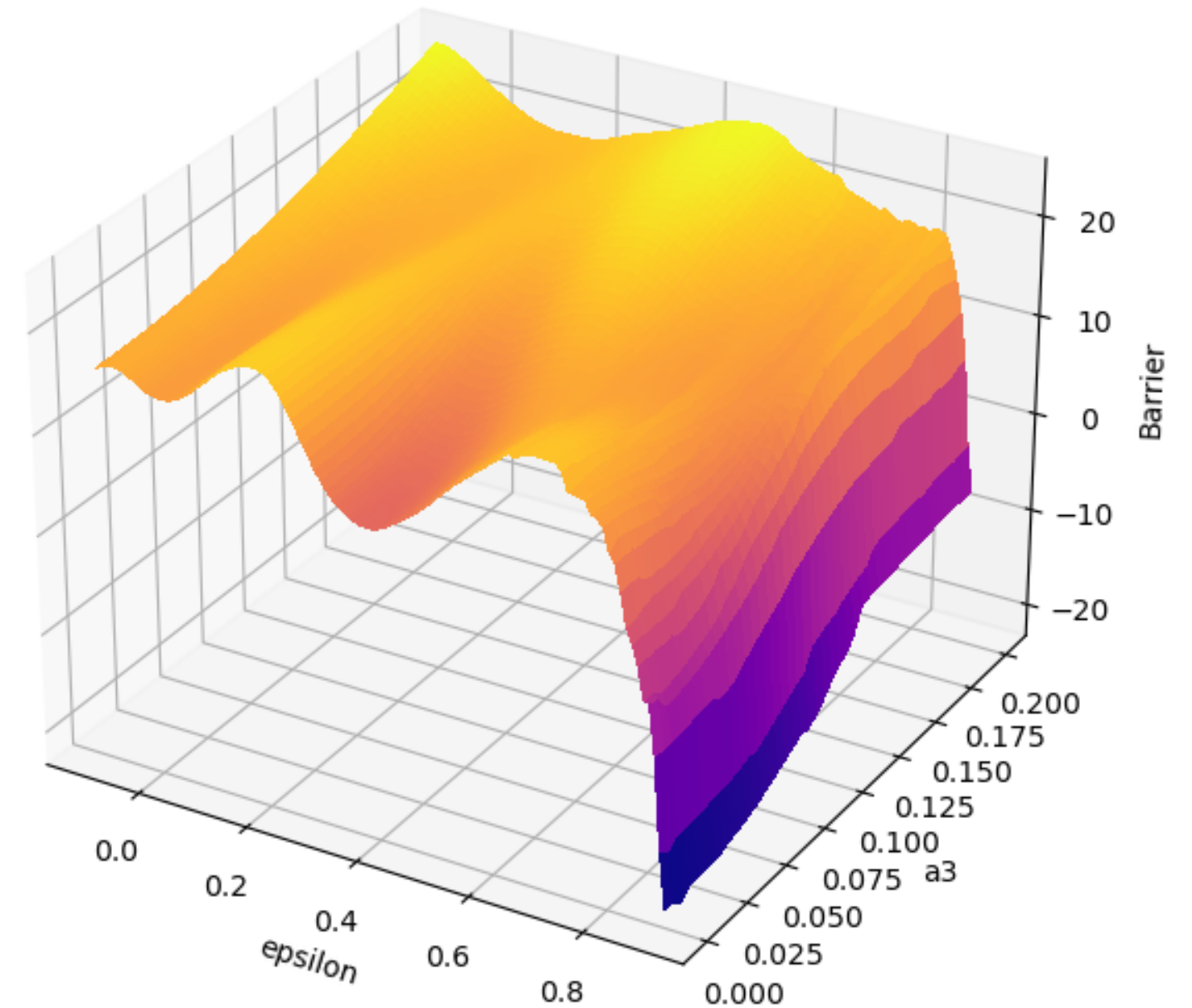
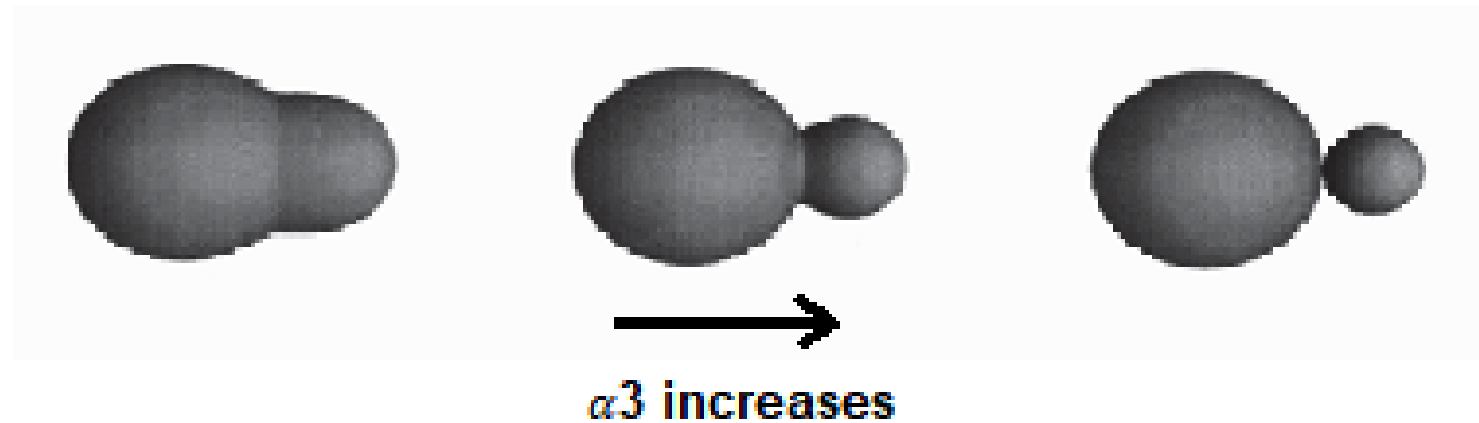
POTENTIAL ENERGY SURFACES

Fission barrier : activation energy required for a nucleus to undergo fission

2D PES: Nucleus energy as a function of elongation ϵ and left/right asymmetry α_3

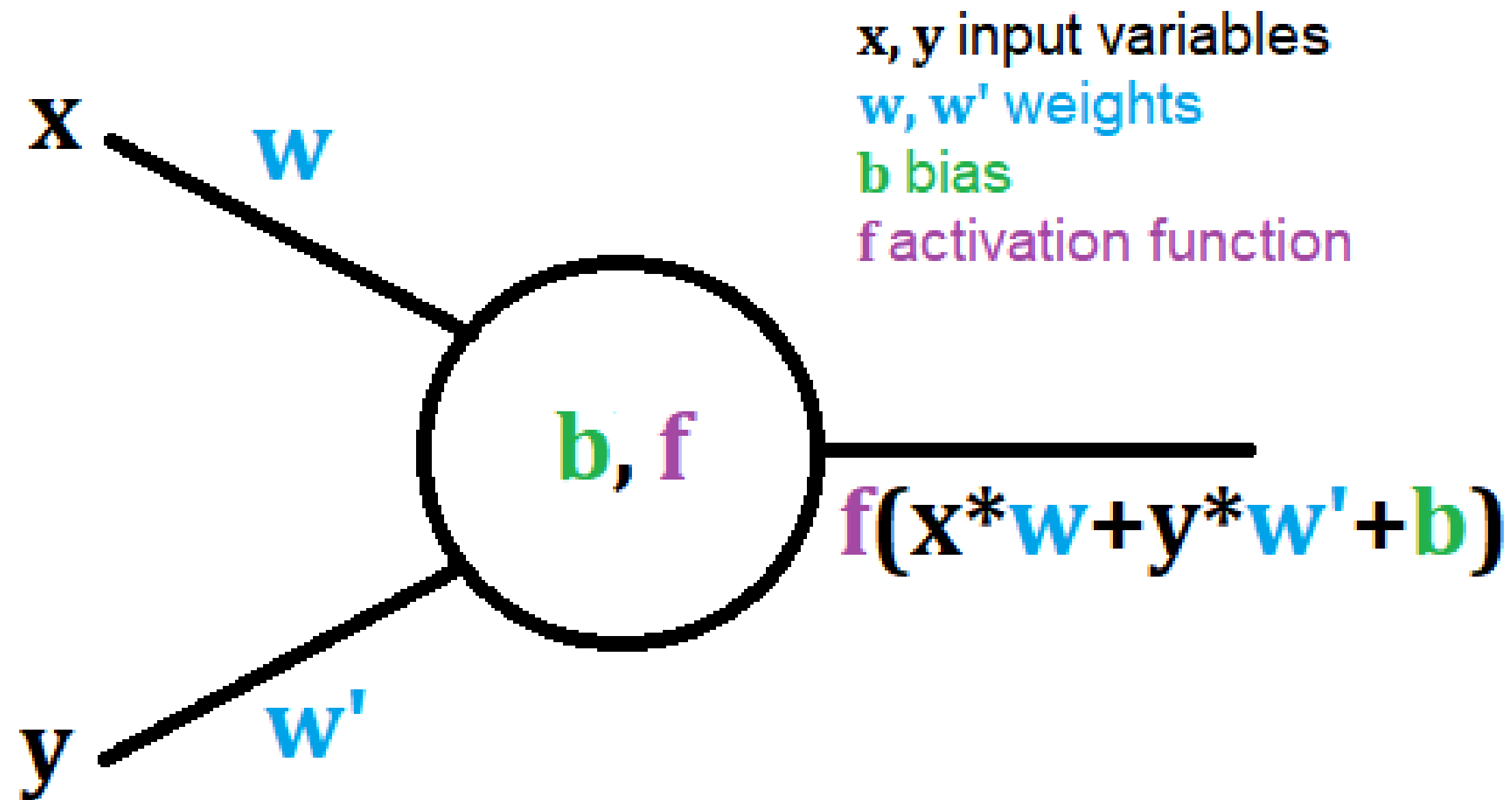


ϵ increases ϵ increases

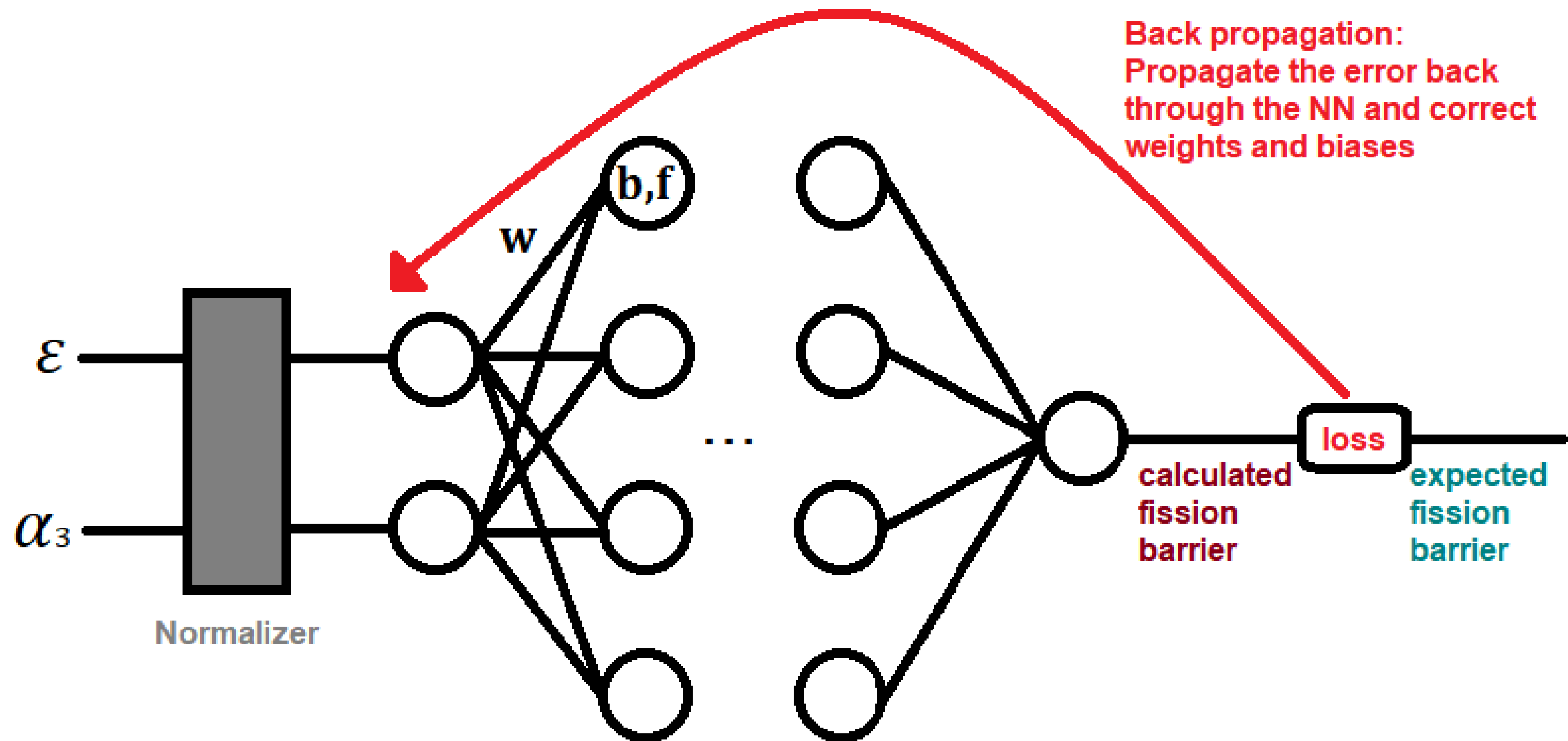


ϵ increase \rightarrow fission barrier falls
 α_3 tends to 0 \rightarrow fission barrier falls

REGRESSION WITH A NEURAL NETWORK - A NEURON



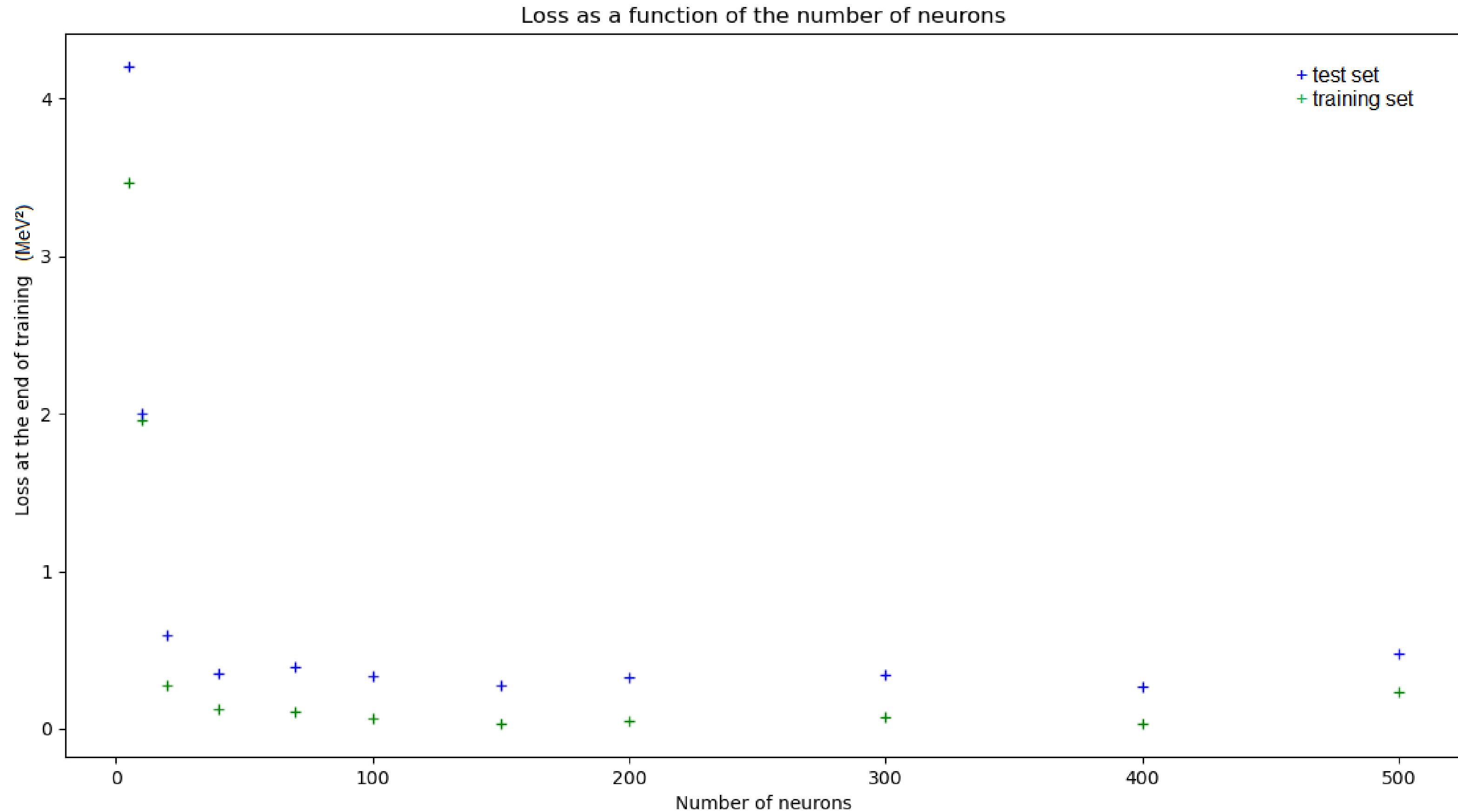
REGRESSION WITH A NEURAL NETWORK - NEURAL NETWORK



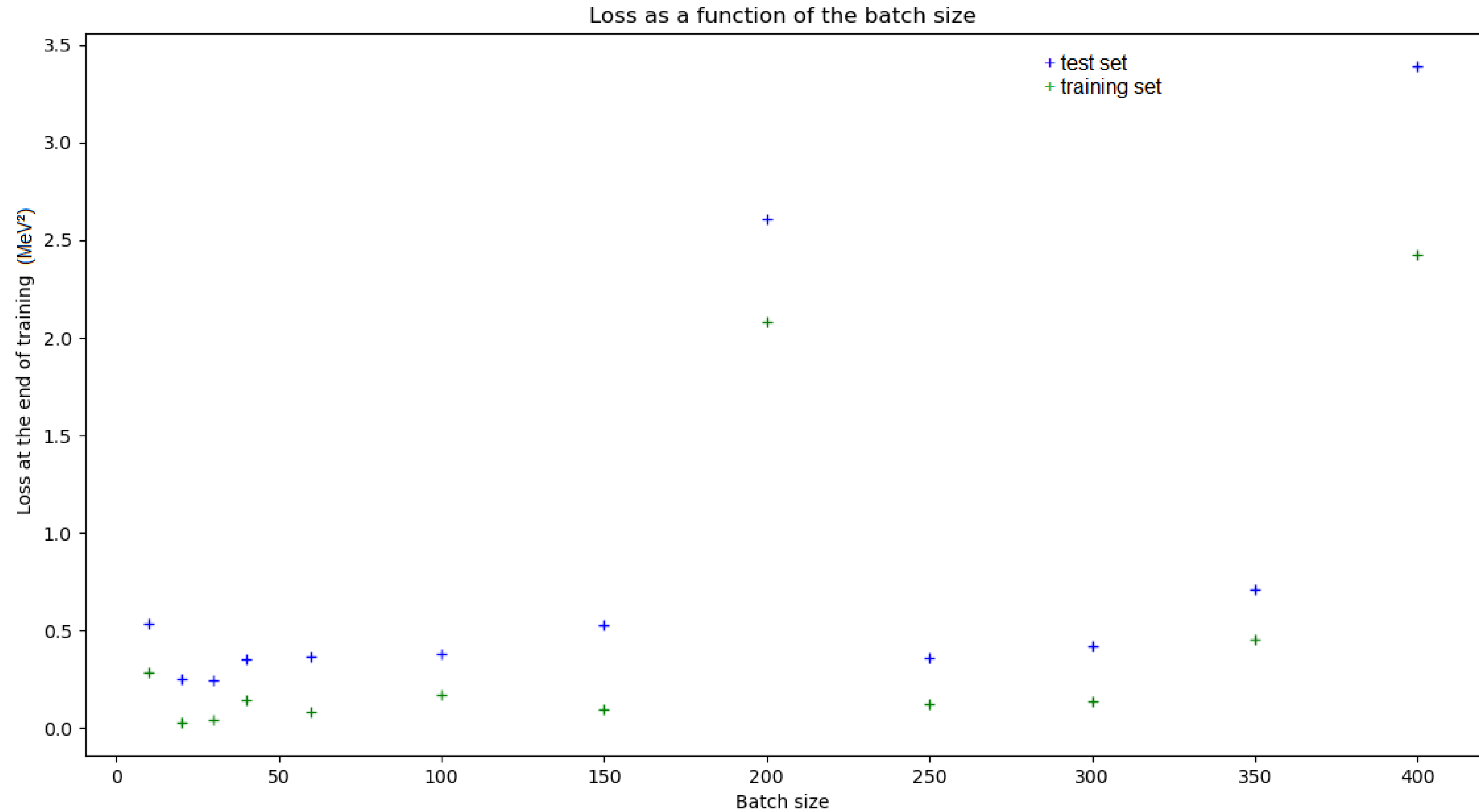
OPTIMIZATION OF HYPERPARAMETERS

1. **The number of hidden layers**
2. **The number of neurons per layer**
3. **The batch size**
4. **The number of epochs**
5. **The activation function**
6. **The loss optimizer**
7. **The loss function**

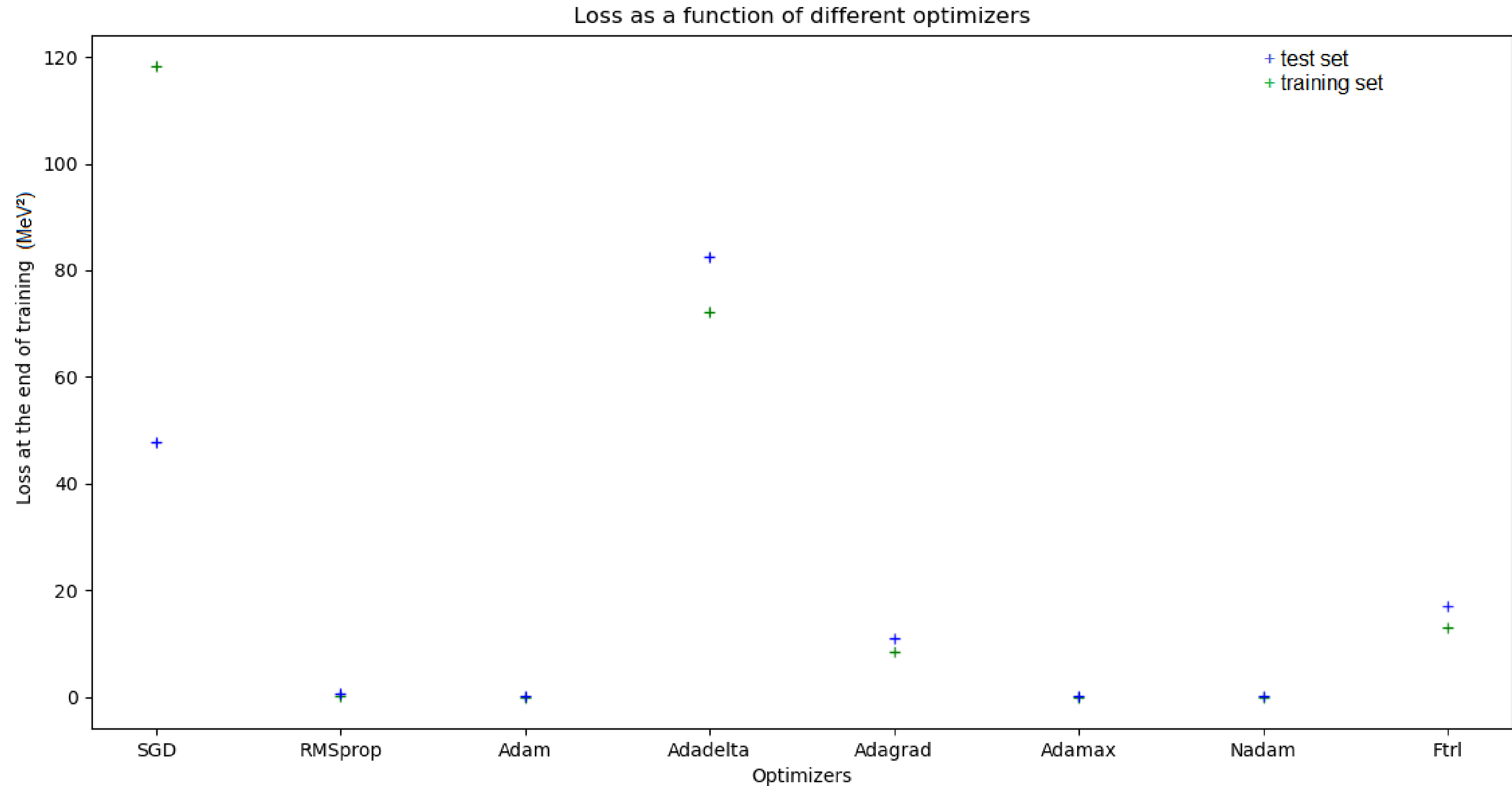
OPTIMIZATION OF HYPERPARAMETERS - NUMBER OF NEURONS



OPTIMIZATION OF HYPERPARAMETERS - BATCH SIZE



OPTIMIZATION OF HYPERPARAMETERS - OPTIMIZER



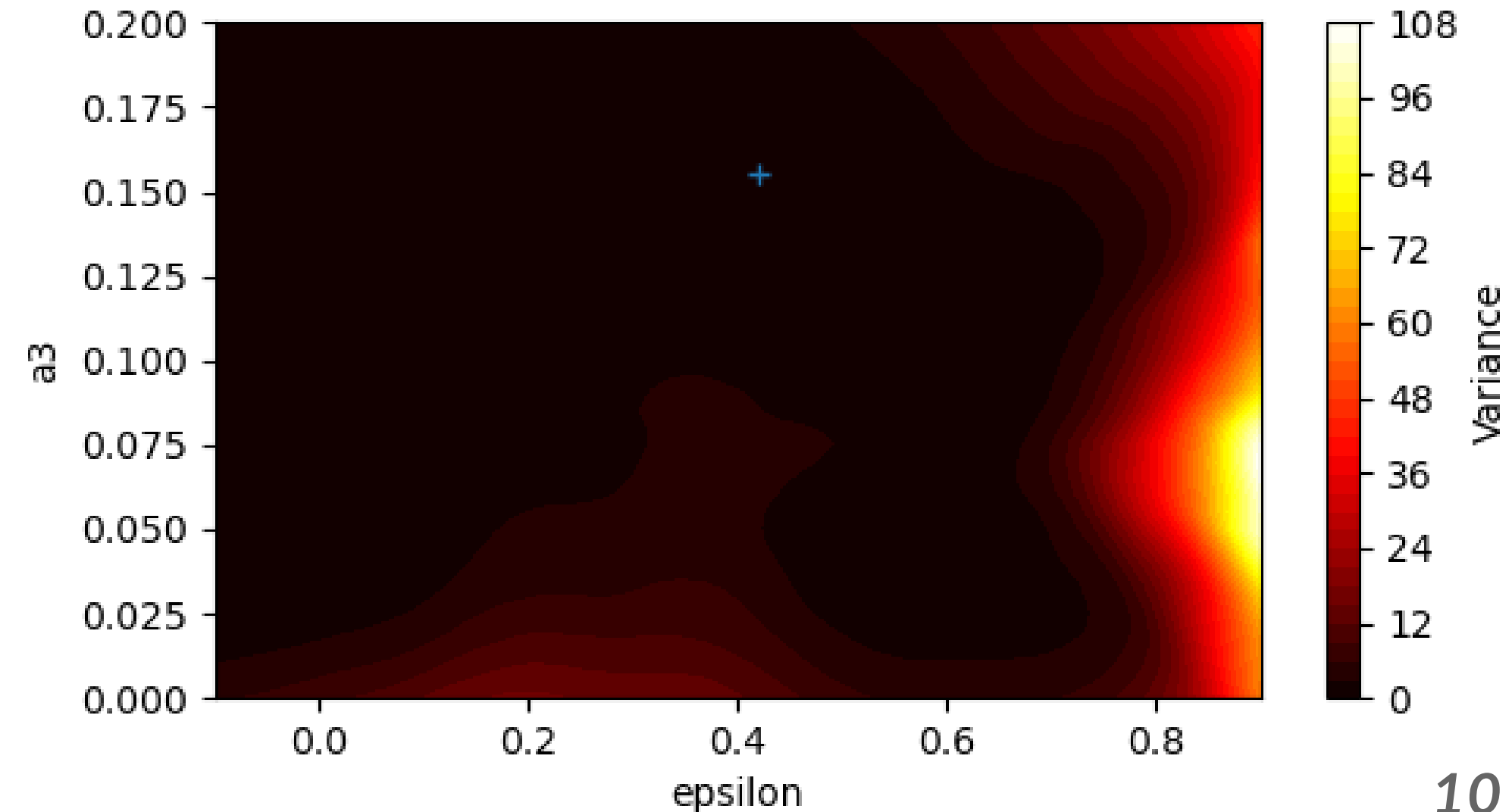
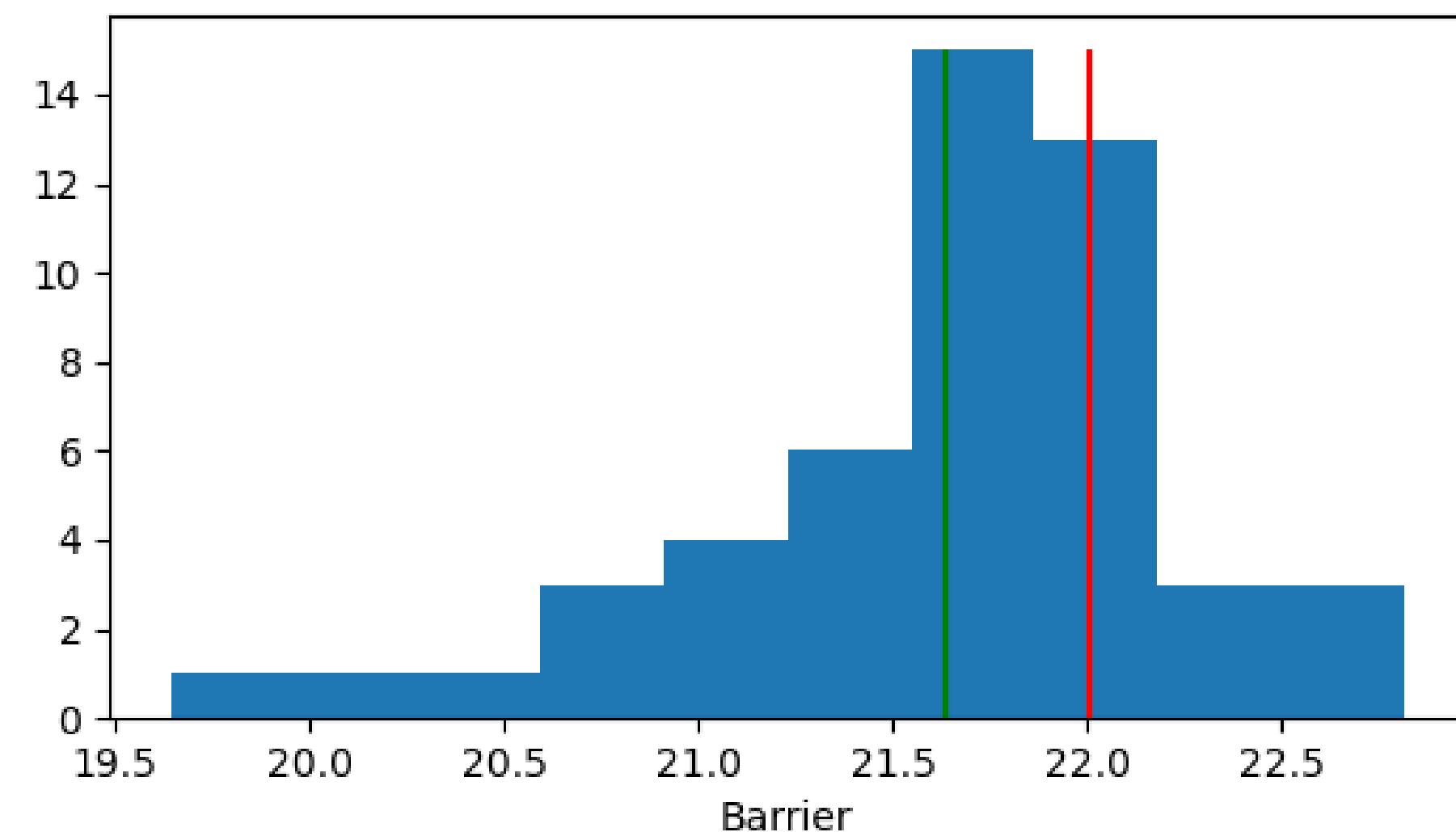
OPTIMIZATION OF HYPERPARAMETERS

- | | | |
|---|------------------------------------|-----------------------------------|
| 1. The number of hidden layers | 3 - 6 | |
| 2. The number of neurons per layer | 150 | |
| 3. The batch size | 10 - 60 | <i>< len(train_set)</i> |
| 4. The number of epochs | 2000 | |
| 5. The activation function | <i>'relu'</i> | |
| 6. The loss optimizer | <i>'Adamax'</i> | |
| 7. The loss function | <i>'mean_squared_error'</i> | |

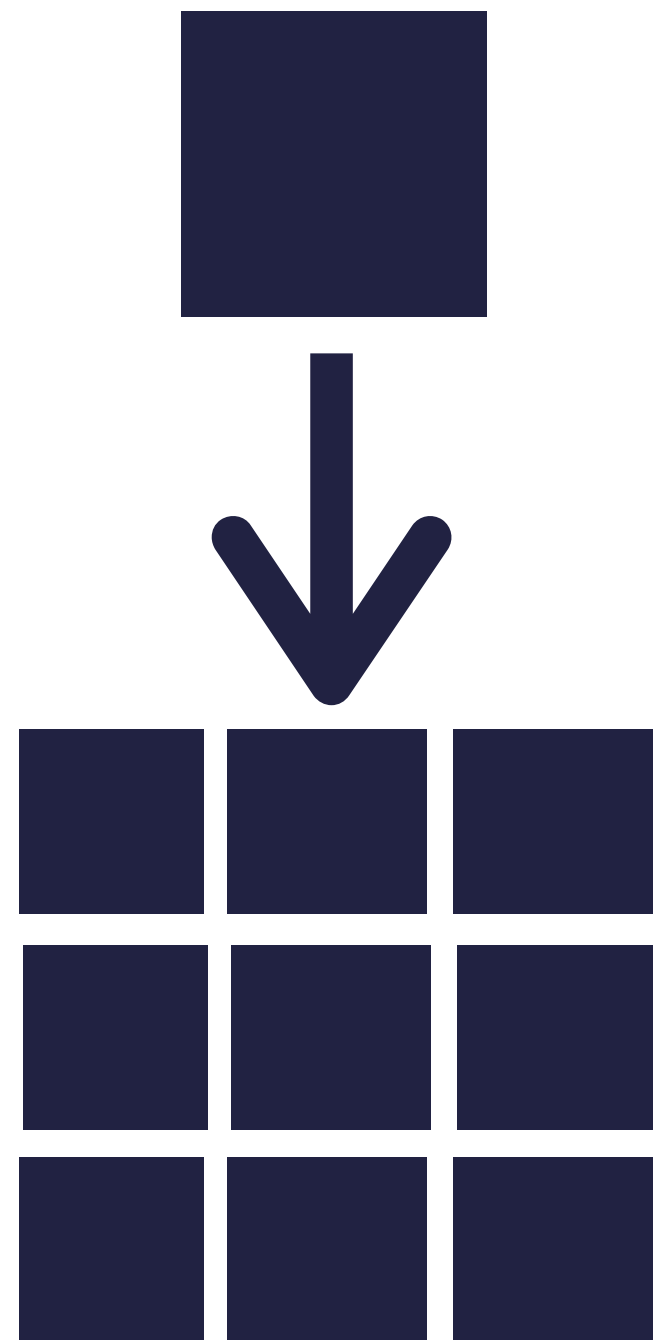
COMMITTEE - IDEA

A single neural network is initialized randomly and can converge to different minima of the loss function. These fluctuations can be removed when using a committee of models.

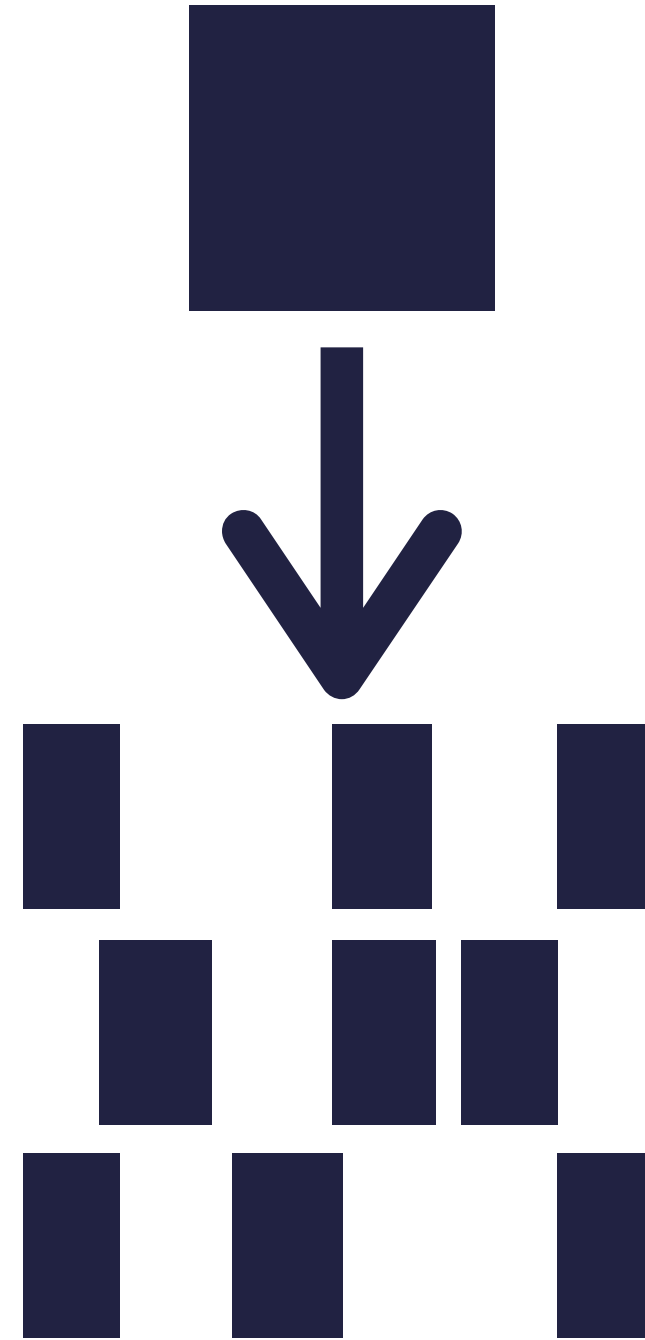
It also add a useful indicator: the variance.



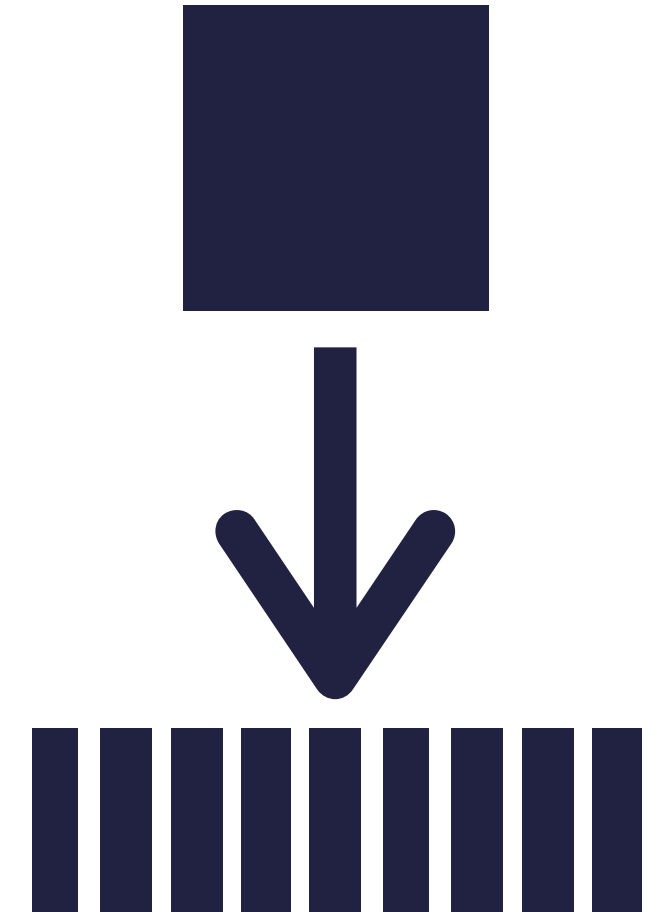
COMMITTEE - SHARING THE TRAINING SET



Sharing

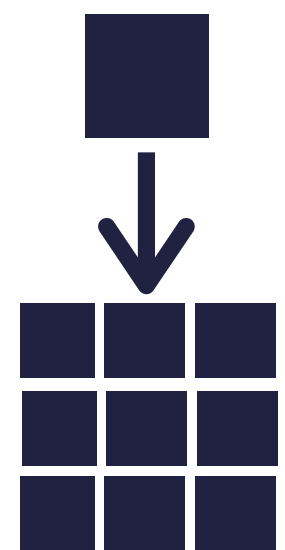


Bootstrapping

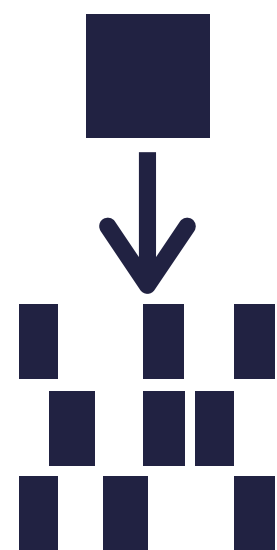


Splitting

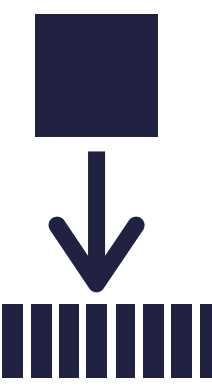
COMMITTEE - SHARING THE TRAINING SET



Sharing



Bootstrapping



Splitting

More systematic errors  **Less systematic errors**

Need less data



Need more data

Generating a PES without simulating a PES

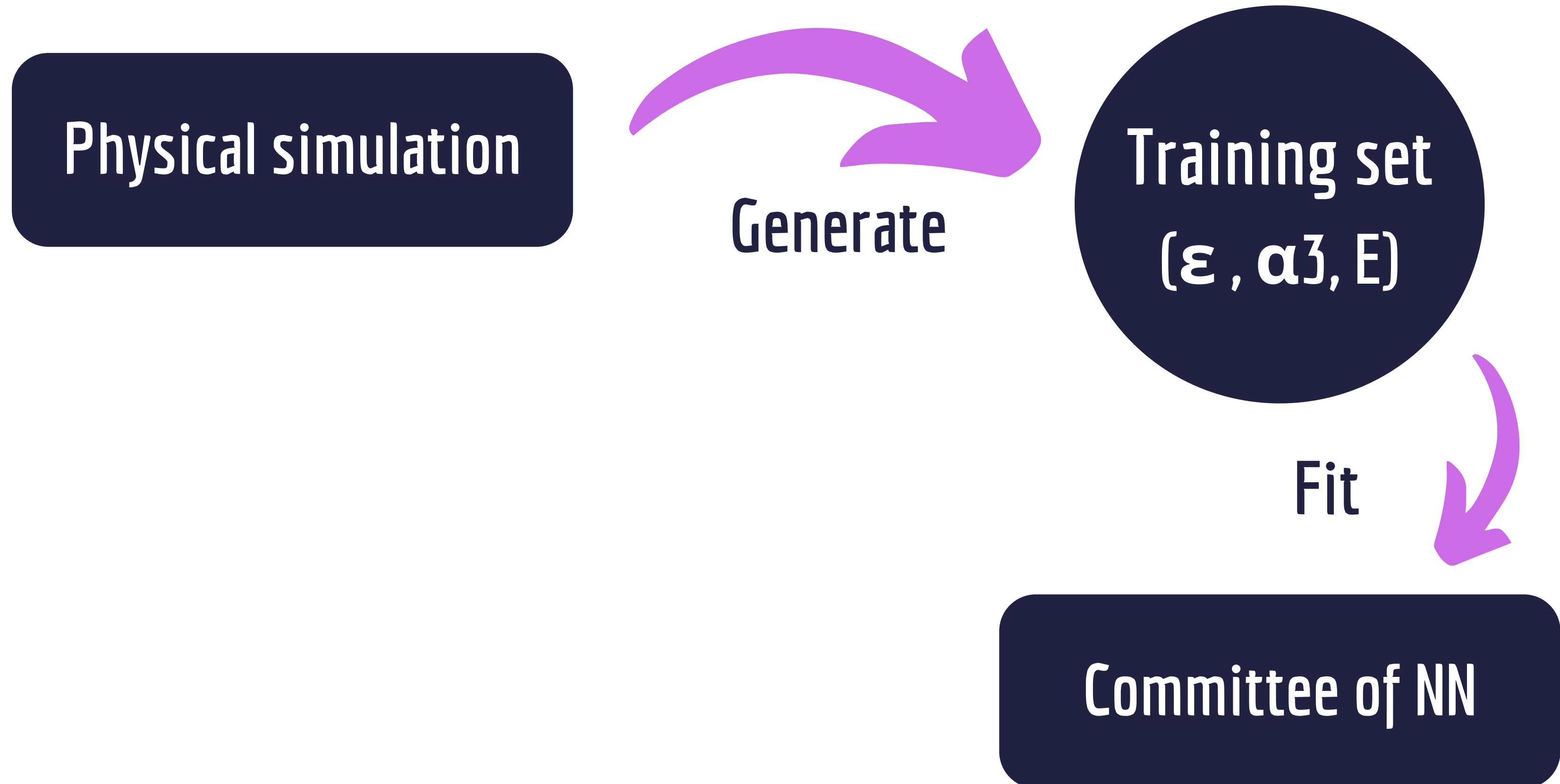
The Committee of models gets a an initial set of training points, fits and uses the variance to tell where to add more training data.

**Goal : maximize the reconstruction accuracy
minimize the size of the training set.**

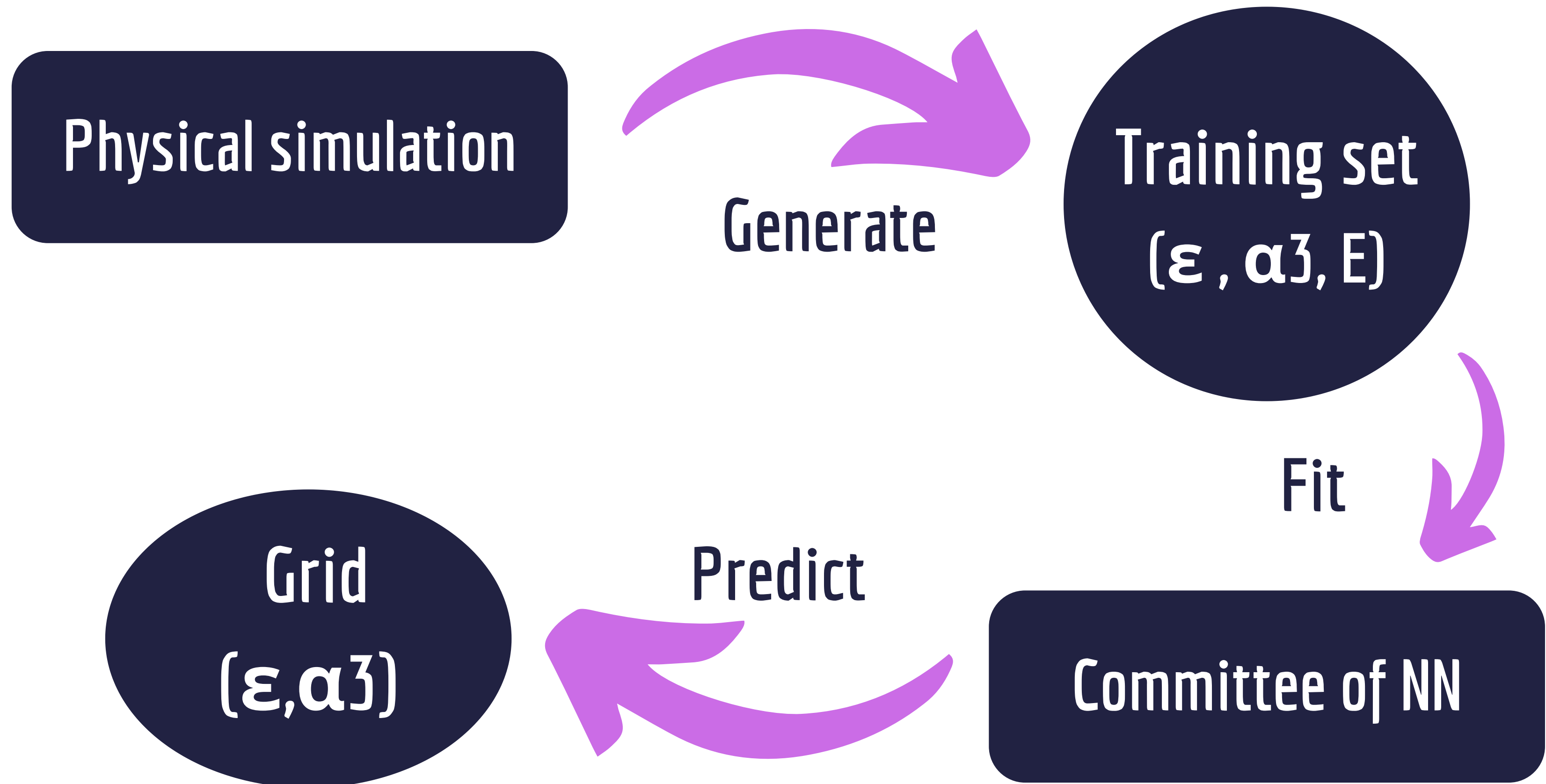
ACTIVE LEARNING - ALGORITHM



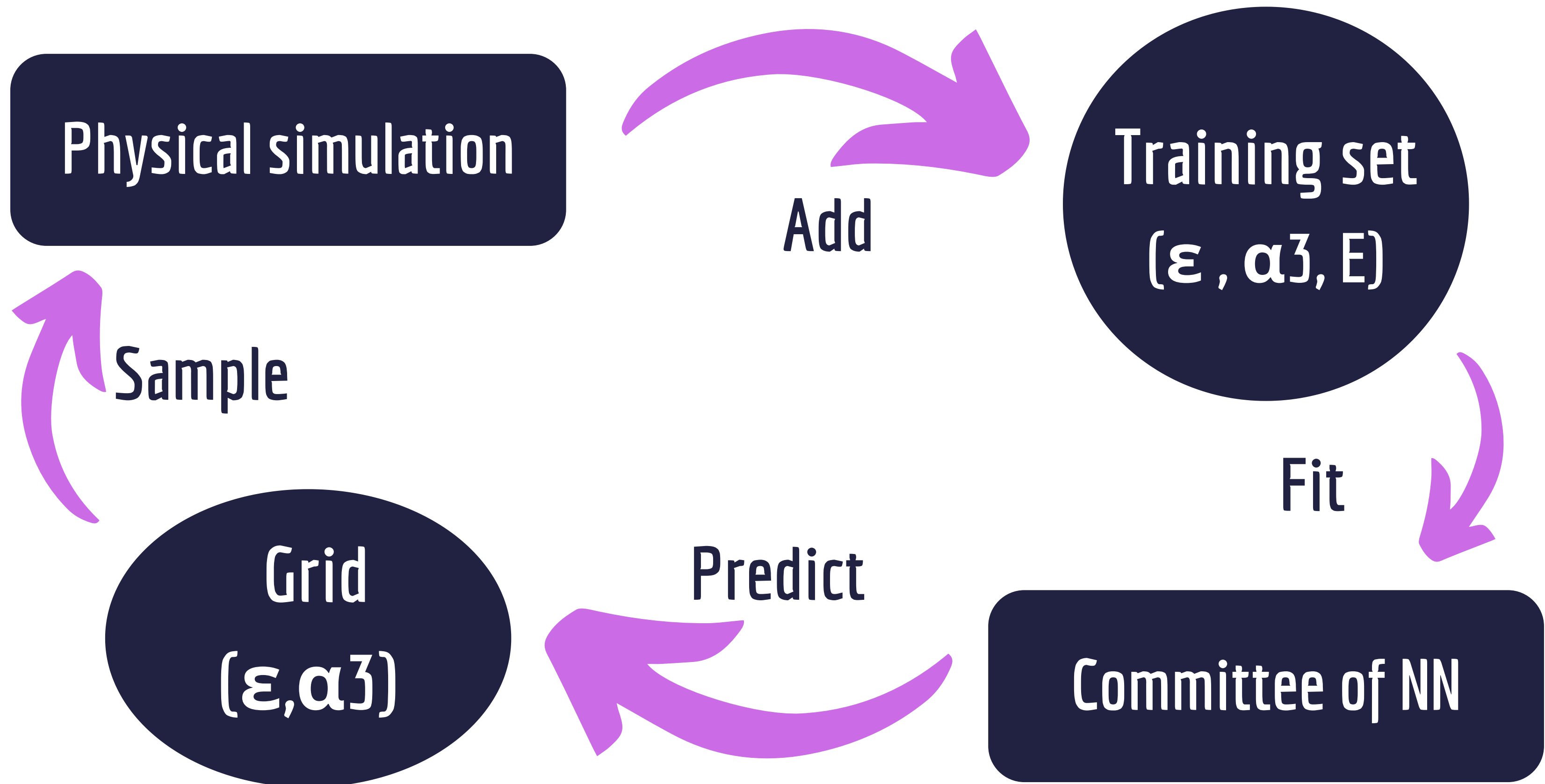
ACTIVE LEARNING - ALGORITHM



ACTIVE LEARNING - ALGORITHM



ACTIVE LEARNING - ALGORITHM



ACTIVE LEARNING - RESULTS

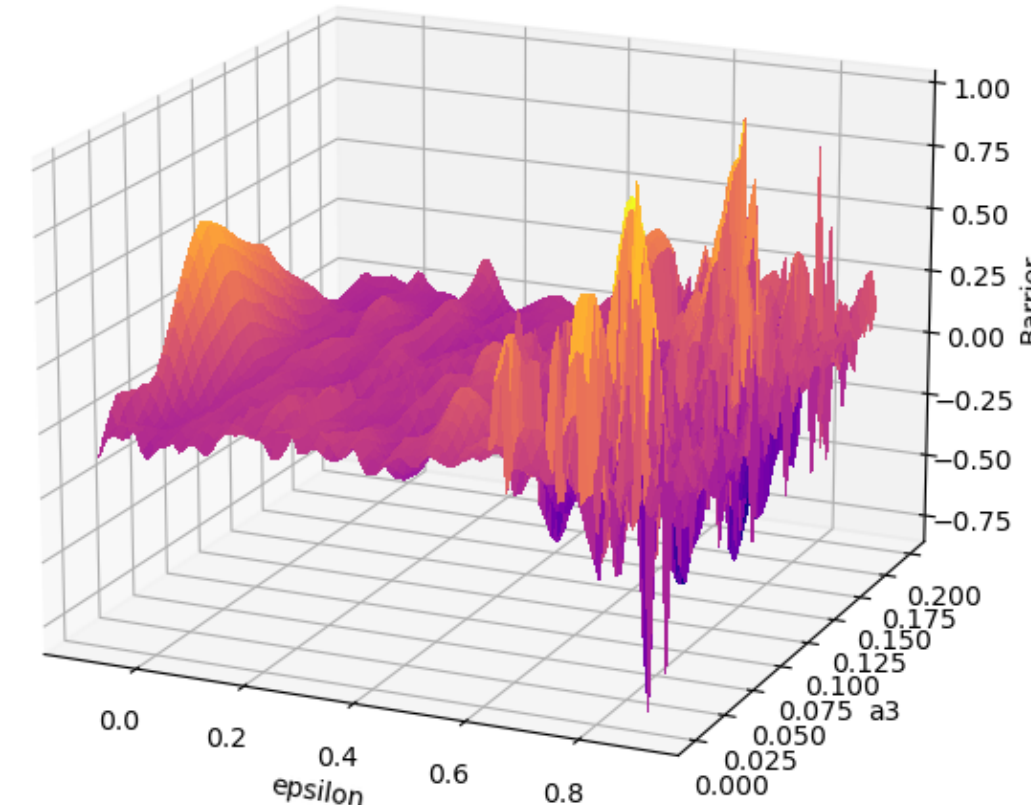
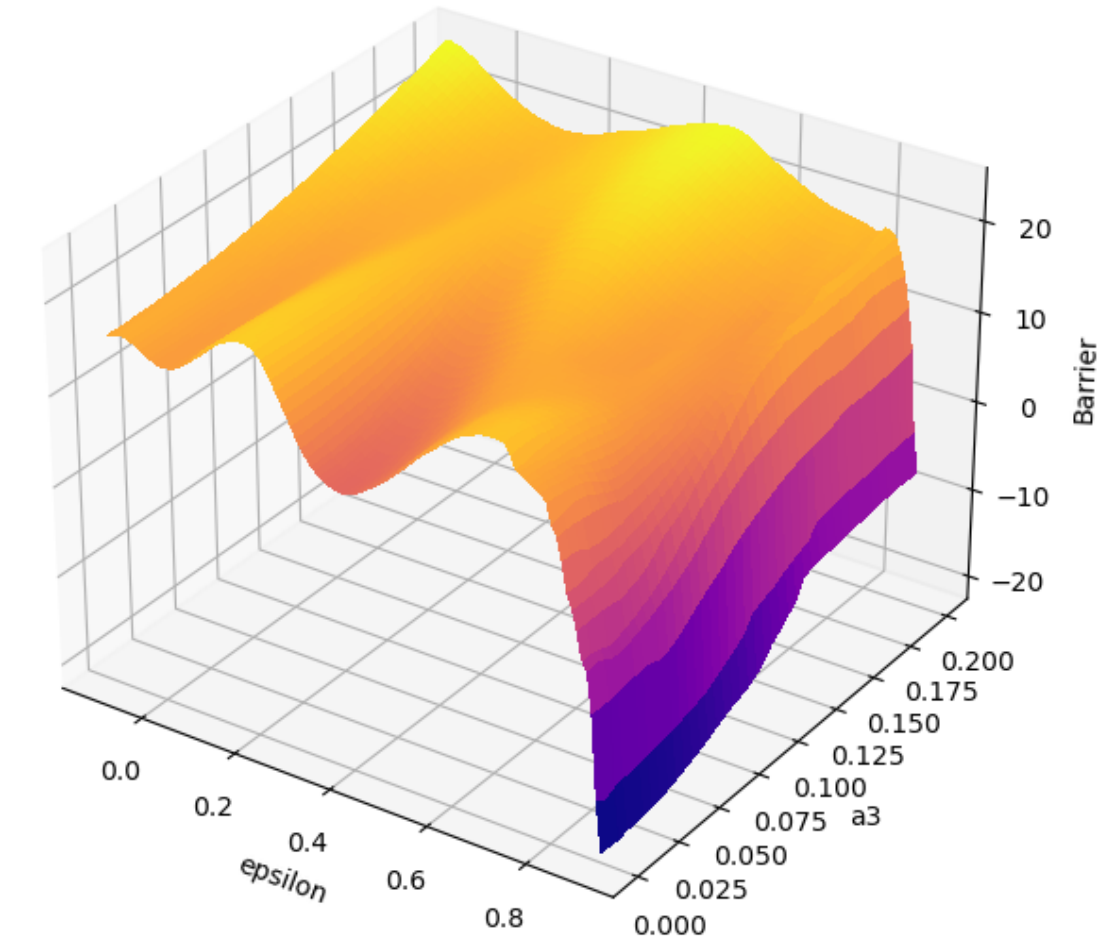
Starting with 3% of the dataset

**RUN 1: final rmse = 302 keV
 10% of the dataset**

**RUN 2: final rmse = 442 keV
 9.9% of the dataset**

**RUN 3: final rmse = 154 keV
 25% of the dataset**

Predicted PES
(Barrier in MeV)



Difference
between
predicted and
simulated PES
(Barrier in MeV)

ACTIVE LEARNING - FINAL THOUGHTS

No optimal hyperparameters for the active learning yet :

- Number of epochs per loop**
- Number of models**
- Sample size of the train dataset for each model**
- Size of the initial train dataset**

And most likely many other optimizations could be made



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