

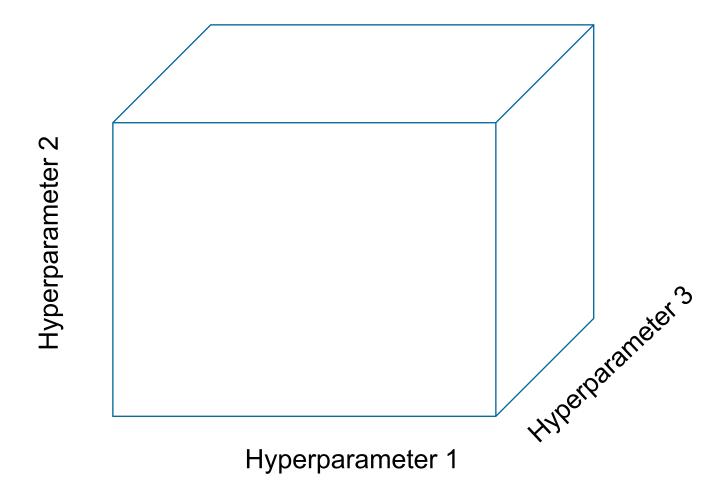
# **Applied Deep Learning**

Dr. Philippe Blaettchen Bayes Business School (formerly Cass) **Hyperparameter tuning – HParams** 

1. Define the hyperparameter space



## Hyperparameter space





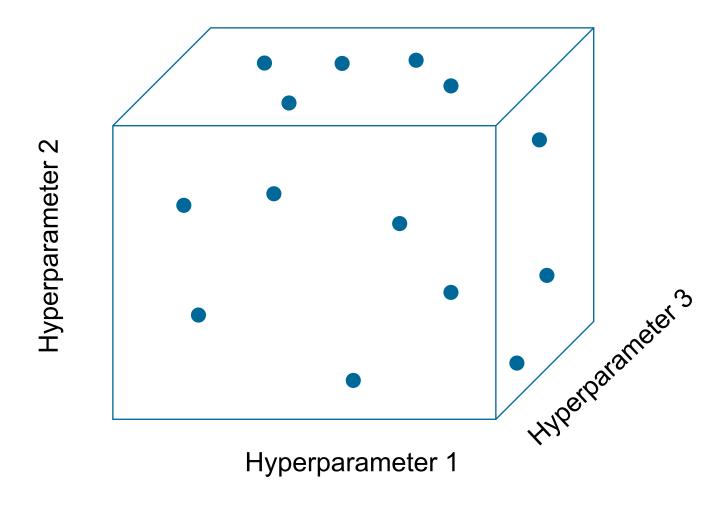
- 1. Define the hyperparameter space
- 2. Define how to keep track of our runs
- Define a function that, given a set of hyperparameters, trains and evaluates a model, then logs it

Always evaluate on the validation set!

1. Create a set of hyperparameters at random and run the above function with that



## Hyperparameter space

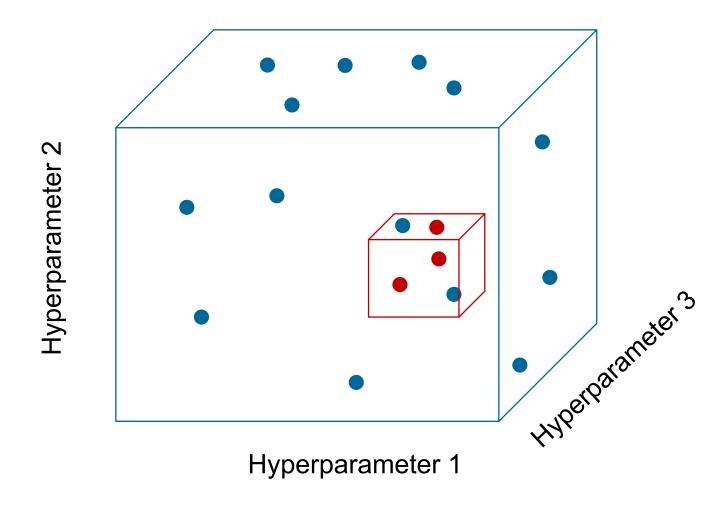




- 1. Define the hyperparameter space
- 2. Define how to keep track of our runs
- Define a function that, given a set of hyperparameters, trains and evaluates a model, then logs it
  - Always evaluate on the validation set!
- 4. Create a set of hyperparameters at random and run the above function with that
- 5. Look at the outcome, possibly search in a smaller grid



## Hyperparameter space





- 1. Define the hyperparameter space
- 2. Define how to keep track of our runs
- Define a function that, given a set of hyperparameters, trains and evaluates a model, then logs it

Always evaluate on the validation set!

- 4. Create a set of hyperparameters at random and run the above function with that
- 5. Look at the outcome, possibly search in a smaller grid
- 6. Evaluate your final model choice (on the test set)



# Let's try it together in Python





**Hyperparameter tuning – Keras Tuner** 

### Hyperparameter tuning process with Keras Tuner

- Define a function that, given a hyperparameter-setter, creates a model
   Within that function, using the hyperparameter-setter, we define the hyperparameter
   space
- 2. Define an instance of the Keras Tuner, specifying the type of hyperparameter search Can use RandomSearch, **Hyperband**, Sklearn, BayesianOptimization
- Let the tuner do its magic
   The hyperparameter-setter will automatically choose the "correct" hyperparameters
- 4. Based on the best parameters found, generate a model, train it, and evaluate it



# Try it out in Python





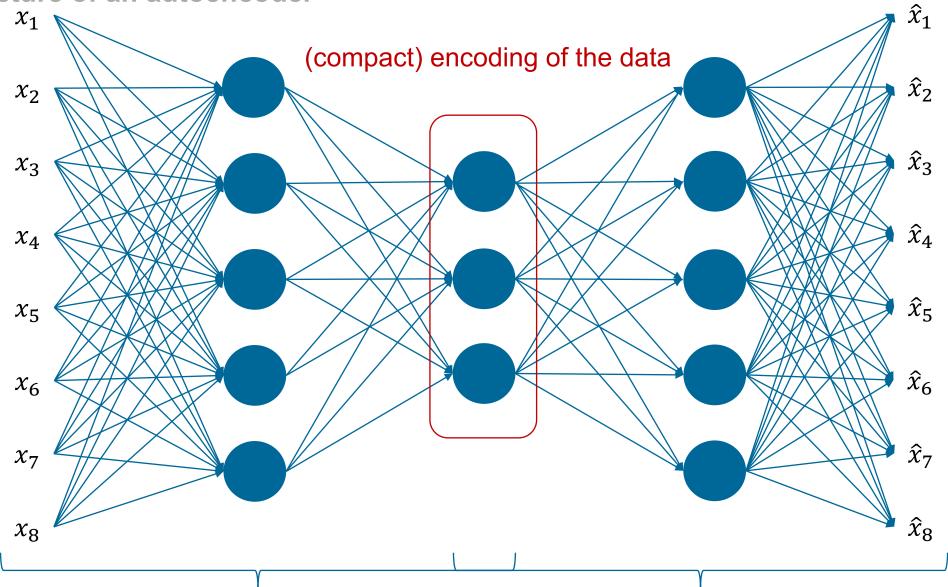
Autoencoders

#### What is an autoencoder

- A neural network that predicts its own inputs
  - → So that we can learn a (compact) representation of the data
- The type of tasks we can use it for:
  - Dimensionality reduction ("advanced PCA")
  - Denoising: train to "recover" data, after artificial noise has been added
  - Anomaly detection: train to represent normal data. When data cannot be predicted well, it
    is likely to be "anormal"
  - Generate new content (such as images): variational autoencoders



#### The structure of an autoencoder



"Encoder": find a representation f(x) "Decoder": unpack x = g(f(x))

# Let's try it together in Python





#### The process for denoising

- We create artificial noise on our data
- We build an autoencoder that takes the noisy data as input, and tries to build an accurate representation of the original
  - To do so with images, we need convolutional layers. Don't worry about how they work, we will get to them soon. You find all the code on using them in the notebook!
  - Training the autoencoder may take quite a bit of time!
- We then can run the autoencoder on new (noisy) data, to create non-noisy versions



# Try it out in Python





