

13.11.25



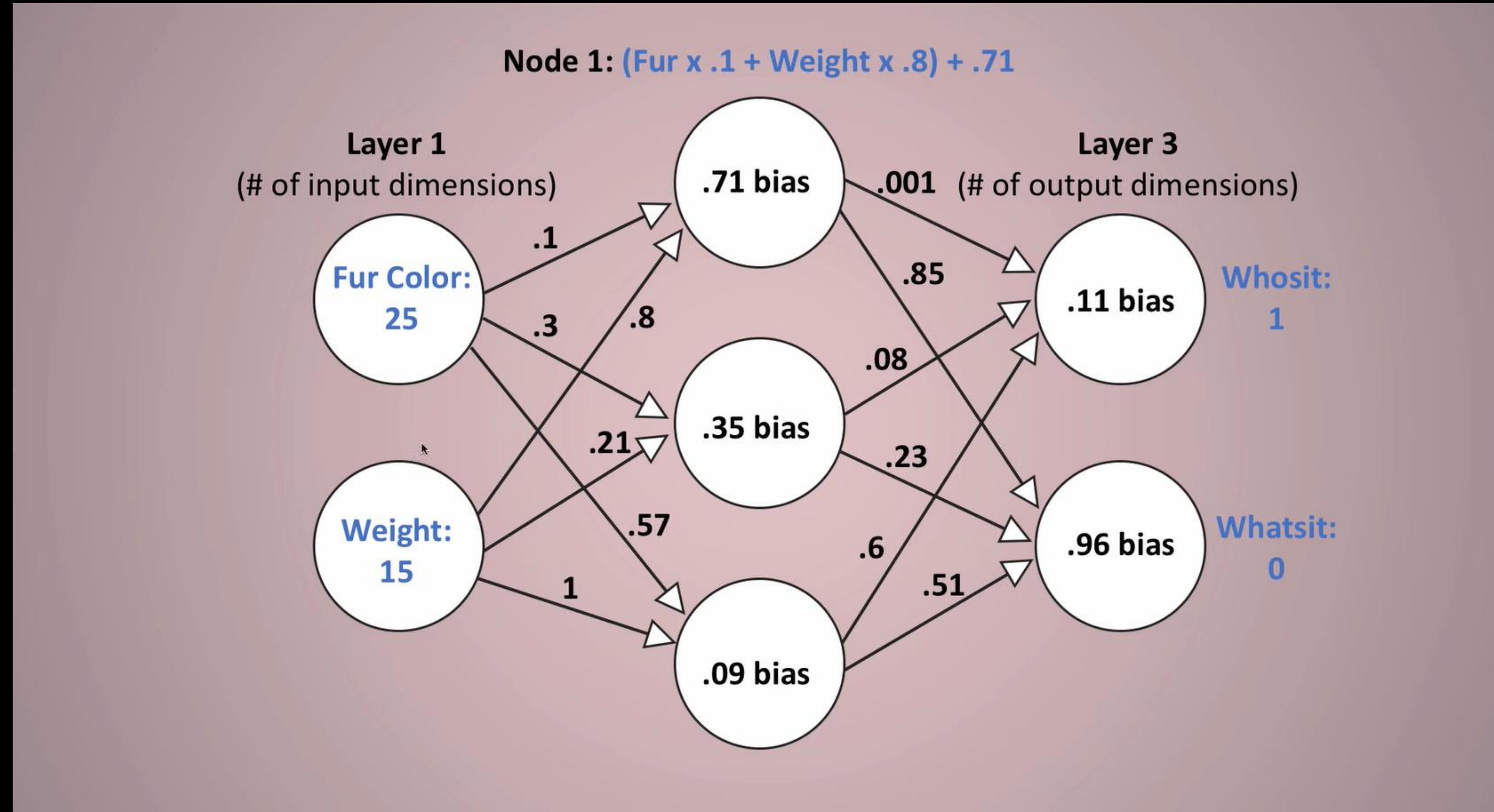
# Introduction to Data Science and Machine Learning

## NEURAL NETS

- **Quiz**
- **Frameworks for Implementing Neural Networks (NN)**
- **Data Preparation and Implementation of a NN with TensorFlow**
- **Additional Layer Types for NNs**

QUIZ





## **Learning Rate:**

**How much should this step outcome affect our weights and biases?**

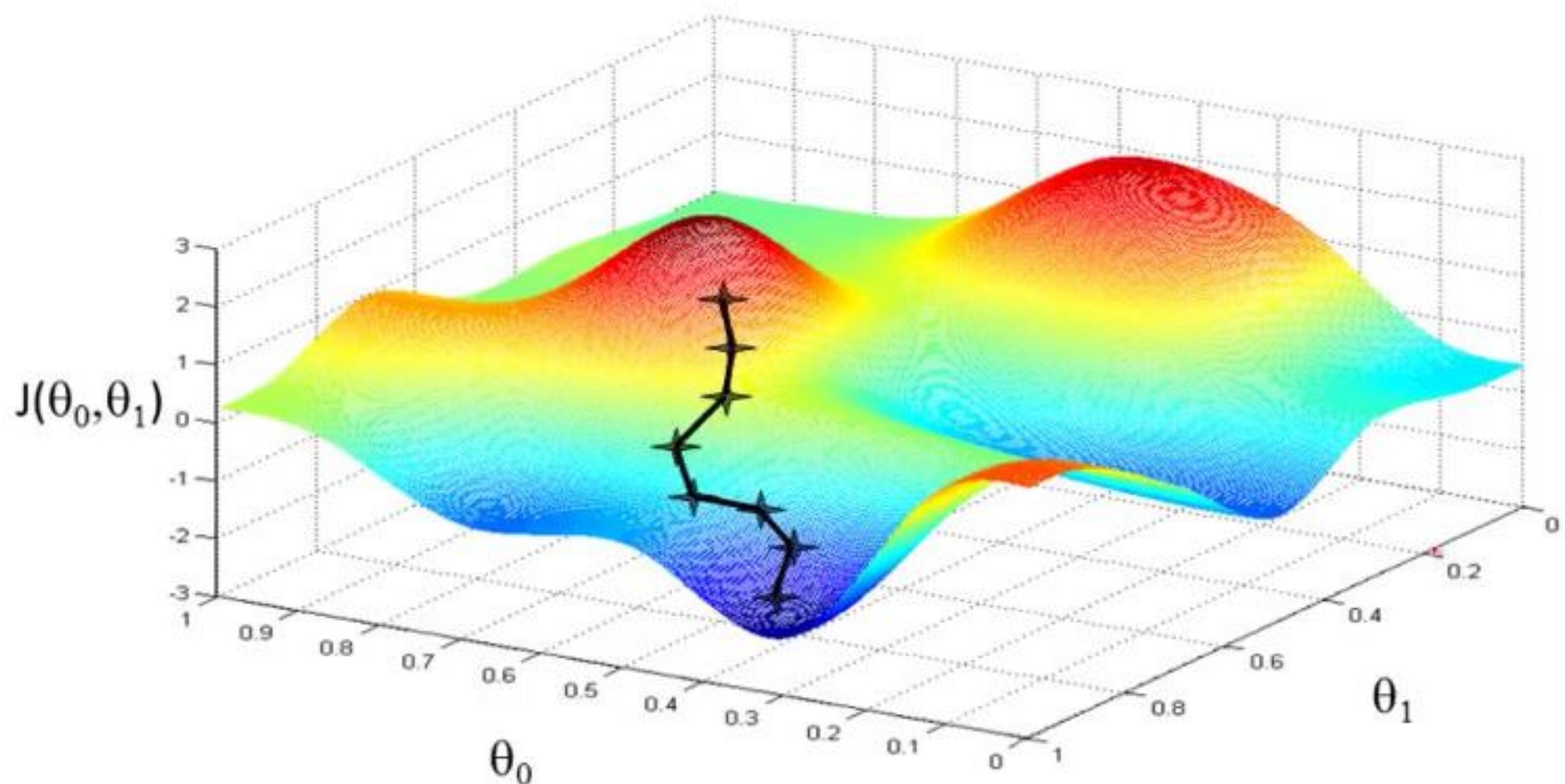
## **Momentum:**

**How much should past outcomes affect our weights and biases?**

**change = (learningRate \* delta \* value) + (momentum \* pastChange)**



# OPTIMIZER PARAMETERS



- Step size for approaching the cost minimum ("Learning Rate")
- Resistance to direction changes ("Momentum")

Quelle: <https://www.coursera.org/learn/machine-learning>

# PARAMETERS OF THE “ADAM” OPTIMIZER

- **Learning parameter for optimization:**  
**alpha (learning rate)**
- **Proportion of current gradient in calculating the next optimization step:**
  - **beta1 (decay rate for the direction) and**
  - **beta2 (decay rate for the magnitude of gradients)**

# (UN-) IMPORTANT HYPERPARAMETERS IN NEURAL NETS

- **Architecture Definition:**
  - **Number of hidden layers in the network**
  - **Types of hidden layers**
  - **Number of neurons per hidden layer**
  - **Selection of activation function**
- **Selection of the optimizer**
- **Setting of the learning rate for the optimizer**

# **LIBRARIES FOR NEURAL NETS**



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Transformers



- **TensorFlow 0.1 (Nov 2015):** Released as open source software by Google; developed by the Google Brain Team for internal research and production)
- **TensorFlow 1.4 (Nov 2017):** Development of the Keras API as a high-level API for TensorFlow and other ML libraries, to increase user-friendliness for commonly used models.
- **TensorFlow 2.0 (Sep 2019):** Keras is integrated as a high-level API into TensorFlow.
- **TensorFlow 2.3 (October 2020):** Significant performance improvements, distributed training, quantized training, and improved mobile deployments.
- **Keras 3.0 (Dec 2023):** Major release that extends support for multiple backends, including TensorFlow, JAX, and PyTorch, making Keras a versatile framework for various deep learning needs.



# Transformers

- **Transformers 0.1 (November 2018):** Initial open-source release providing an implementation of BERT in Pytorch.
- **Transformers 1.0 (July 2019):** Inclusion of other models beyond BERT
- **Transformers 2.0 (July 2019):** Inclusion of the TensorFlow framework
- **Transformers 3.0 (June 2020):** New Tokenizer API, enhanced documentation & tutorials
- **Transformers 4.0 (November 2020):** Fast tokenizers and file re-organization
- **Transformers 4.x Series (2020-Present):**
  - 56 minor releases until today
  - supporting state-of-the-art architectures: LLaMA, Mistral, Gemma, Qwen, Phi, and many others
  - advanced multimodal capabilities (vision-language, audio-text, video understanding)

# DATA PREPARATION

**For every modeling, the data must have the following properties:**

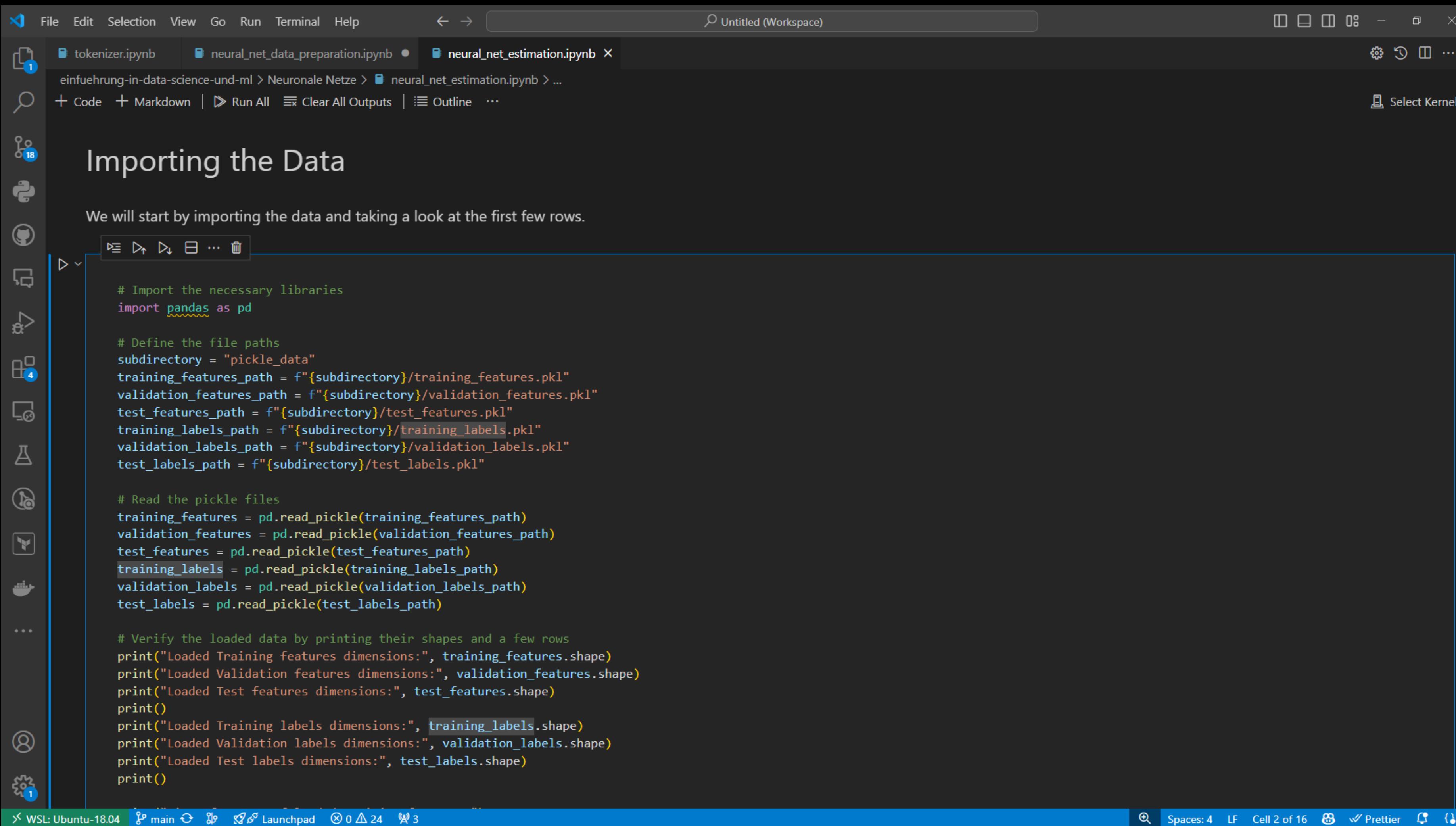
- 1. There must be no missing values.**
- 2. All values must be numbers.**
- 3. Categorical variables are one-hot encoded.**

# EXAMPLE DATA PREPARATION FOR THE TRAINING OF A NEURAL NET

The screenshot shows a Jupyter Notebook interface with the following details:

- File Bar:** File, Edit, Selection, View, Go, Run, Terminal, Help.
- Title Bar:** Untitled (Workspace).
- File List:** tokenizer.ipynb, neural\_net\_data\_preparation.ipynb (selected), neural\_net\_estimation.ipynb.
- Toolbar:** Code, Markdown, Run All, Restart, Clear All Outputs, Variables, Outline, Python 3.8.0.
- Section Header:** Import Libraries and Data.
- Code Cell:** Contains Python code for importing libraries (pandas, numpy, os) and data from a CSV file. It also includes a call to `data.head()` to verify the data.
- Data Preview:** Shows the first 5 rows of the imported data frame. The columns include id, date, bedrooms, bathrooms, sqft\_living, sqft\_lot, floors, waterfront, view, condition, sqft\_above, sqft\_basement, yr\_built, yr\_renovated, zipcode, lat, long, and sqft\_living.
- Section Header:** Data Preparation.
- Code Cell:** Contains Python code for defining categorical features and inspecting their data types and unique values.
- Bottom Status Bar:** WSL:Ubuntu-18.04, main, Launchpad, Spaces: 4, LF, Cell 6 of 8.

# EXAMPLE DEFINITION AND OPTIMIZATION OF A NEURAL NET



The screenshot shows a Jupyter Notebook workspace titled "Untitled (Workspace)". The top bar includes File, Edit, Selection, View, Go, Run, Terminal, Help, and a search bar. The left sidebar lists files: tokenizer.ipynb (1), neural\_net\_data\_preparation.ipynb, and neural\_net\_estimation.ipynb (marked with a red dot). Below the sidebar, a section titled "Importing the Data" contains the following text:

We will start by importing the data and taking a look at the first few rows.

```
# Import the necessary libraries
import pandas as pd

# Define the file paths
subdirectory = "pickle_data"
training_features_path = f"{subdirectory}/training_features.pkl"
validation_features_path = f"{subdirectory}/validation_features.pkl"
test_features_path = f"{subdirectory}/test_features.pkl"
training_labels_path = f"{subdirectory}/training_labels.pkl"
validation_labels_path = f"{subdirectory}/validation_labels.pkl"
test_labels_path = f"{subdirectory}/test_labels.pkl"

# Read the pickle files
training_features = pd.read_pickle(training_features_path)
validation_features = pd.read_pickle(validation_features_path)
test_features = pd.read_pickle(test_features_path)
training_labels = pd.read_pickle(training_labels_path)
validation_labels = pd.read_pickle(validation_labels_path)
test_labels = pd.read_pickle(test_labels_path)

# Verify the loaded data by printing their shapes and a few rows
print("Loaded Training features dimensions:", training_features.shape)
print("Loaded Validation features dimensions:", validation_features.shape)
print("Loaded Test features dimensions:", test_features.shape)
print()
print("Loaded Training labels dimensions:", training_labels.shape)
print("Loaded Validation labels dimensions:", validation_labels.shape)
print("Loaded Test labels dimensions:", test_labels.shape)
print()
```

The bottom status bar shows WSL: Ubuntu-18.04, main, Launchpad, 0 △ 24, 3, Spaces: 4, LF, Cell 2 of 16, Prettier, and other icons.

# BREAKOUT

- Load the example notebooks from below into your Codespace and run them once unchanged.
- Supplement your data preparation with the steps performed in [this example notebook](#):
  - 1) One-hot encoding of categorical variables
  - 2) Removing cases with missing values
  - 3) Export of training and validation data as pickle files
- Estimate a first neural network based on [this example notebook](#).

# BATCHES, STEPS AND EPOCHS

## *Batch*

- **The entire set of training data is divided into separate subgroups of equal size.**
- **The standard batch size in TensorFlow is 32.**

## *Step*

- **A single iteration of gradient descent performed on one batch of data, during which all model weights are updated once.**

## *Epoch*

- **Optimization of the model using the complete training data:  
Number of Steps × Batch Size = Training Sample Size**
- **Depending on the model, very few epochs may suffice, or several hundred or thousand may be needed for optimization.**

# NORMALIZATION

**Definition:**

- **Subtracting the mean and dividing by the standard deviation.**

**Ensures all input features are on similar scales, which stabilizes training and speeds up convergence.**

# BATCH NORMALIZATION

- **Performing normalization at the batch level**

**Additional optimization parameters:**

- **Exactly identical means and standard deviations are not necessarily optimal for modeling purposes**  
→ **Normalization parameters are incorporated as trainable parameters**

# LEARNING RESSOURCES

- Watch [this video](#) (7 minutes) to better understand the properties of dropout layers.
- Watch [this video](#) (5 minutes) to better understand the benefits of normalization.
- Complete the first chapter of [this course](#) on DataCamp to learn about identifying missing values.

# TASKS

- **Examine all your model variables for the existence of missing and implausible values.**
- **Prepare your dataset by correctly encoding all categorical features and removing all rows with missing values.**
- **Train a first neural network for your dataset.  
(Delete all rows with missing values.)**