

```

import os

import numpy as np

import pandas as pd

from pathlib import Path

from tensorflow.python.keras.utils.np_utils import to_categorical


def read_painmonit_np(label):
    """ Function to read the UzL dataset in form of numpys created by script "create_np_files".

    Parameters
    -----

    label: String. Either 'covas' or 'heater' depending on the label you want to use.


    Returns
    -----

    X, y, subjects: np.
    """
    np_dir = Path("datasets", "painmonit", "np-dataset")

    return np.load(Path(np_dir, "X.npy")), np.load(Path(np_dir, "y_{}.npy".format(label))),
    np.load(Path(np_dir, "subjects.npy"))


def read_biovid_np():
    """ Function to read the BioVid dataset in form of numpys created by script "create_np_files".

    Parameters
    -----

    label: String. Either 'covas' or 'heater' depending on the label you want to use.


    Returns

```

```

-----

X, y, subjects: np.

'''

np_dir = Path("datasets", "biovid", "np-dataset")

return np.load(Path(np_dir, "X.npy")), np.load(Path(np_dir, "y.npy")), np.load(Path(np_dir,
"subjects.npy"))

```

```
def get_initials(file_name, n= 4):
```

```
    """
```

Returns the initials in a given filename.

Returns the first 'n' characters after the last os directory seperator (windows: '\\') inside the string.

Example: '\\data-files\\Ad Wa-2019-12-19_11-03-14' -> 'Ad Wa' with n=5

Parameters

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```

file_name: String. Example: "\\synchronised-data-files\\Ad Wa-2019-12-19_11-03-14"

n: Int. Number of characters to extract after last os seperator.

Returns

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```

string: Initials. "Ad Wa"

```
    """
```

```
# Find last occurrence of "\\"
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```
index = file_name.rfind(os.path.sep)
```

```
# Get initials -> assumes that the first 5 letters of the filename describe the initials (e.g. "***\\Ad
Wa***")
```

```
initials = file_name[index + 1: index + 1 + n]
```

```
return initials
```

```
def from_categorical(x):
```

```
    """Returns the class vector for a given one-hot vector.
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```
    Parameters
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```

```
    x: np. One-hot vector.
```

```
    Returns
```

```
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```

```
    np: np.argmax(x, axis= 1)
```

```
    """
```

```
    return np.argmax(x, axis= 1)
```

```
def unison_shuffled(x):
```

```
    """ Shuffles several numpy/pandas arrays (given as list) unison.
```

```
    Parameters
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```
    -----
```

```
    x: list of np.
```

```
    Returns
```

```
    -----
```

```
    y: list of np.
```

```

'''
# Check all elements have same length
assert len({len(i) for i in x if i is not None}) == 1

# Create a random permutation
p = np.random.permutation(len(x[0]))

# Apply permutation to all elements
return [i[p] if isinstance(i, np.ndarray) else i.reset_index().reindex(p).drop(["index"], axis = 1) if
isinstance(i, pd.DataFrame) else None for i in x]

```

```

def pick_classes(data, y, classes, input_is_categorical= False):

```

```

    """Function to pick certain classes for given data.

```

```

    Parameters

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```

```

    data: list. Data to pick from.

```

```

    y: np. Label of the dataset.

```

```

    classes: list/np. List of lists with class labels. For example, [[0], [4]], or [[0], [3, 4]]

```

```

    input_is_categorical: bool. Should be set True if y is given as categorical. Otherwise False.
    Default is False.

```

```

    Returns

```

```

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```

```

    list: data for the two class problem

```

```

    np: label for the two class problem

```

```

    """

```

```

    if input_is_categorical:

```

```

        y = from_categorical(y)

```

```

    # Create a dict to convert class labels: picked_classes [[0], [4]] -> 0:0, 4:1 or [[0], [3, 4]] -> 0:0,
    3:1, 4:1

```

```

d = {}
for class_idx, class_group in enumerate(classes):
    for class_name in class_group:
        d[class_name] = class_idx

# Convert initial class label to new ones and set not used ones to '-1'
y = np.array([d[x] if x in d else -1 for x in y])

# Create index list for elements according the picked classes
indices = [False] * len(y)
for c in np.unique(list(d.values())):
    indices = indices | (y == c)

# Add label vector to the data list
data.append(y)

# Select elements absed on the index list for all dataframes in 'data'
data = pick_data(data= data, condition= indices)

# if input was categorical, convert 'y' to categorical again
if input_is_categorical:
    data[-1] = to_categorical(data[-1])

return data

```

```

def pick_data(data, condition):

```

```

    """Function to pick certain data for given data under given condition.

```

```

    Parameters

```

data: list. Data to pick from

condition: list of bool.

Returns

list: selected data -> return [i[condition] for i in data]

"""

return [i[condition] if i is not None else None for i in data]

def min_max_scaling(x):

return (x - x.min()) / (x.max() - x.min())

def normalize_features(x):

if not isinstance(x, pd.DataFrame):

raise NotImplementedError("normalize_features' expects pandas dataframe but received type: '{}'.format(type(x))")

for column in x.columns:

x[column]= min_max_scaling(x[column])

return x

def normalize(x):

"""Function to normalize the data of given to a CNN.

Expects a np dataframe in shape (num samples, time series data, sensors, channels).

Normalizes each 'times series data' for each sample independent.

Parameters

x: np.

Returns

normalized data.

"""

if x is None:

 return None

for i, data in enumerate(x):

 for y in np.arange(data.shape[1]):

 max_value = np.max(data[:, y])

 min_value = np.min(data[:, y])

 if max_value != min_value:

 data[:, y] = (data[:, y] - min_value) / (max_value - min_value)

 else:

 if max_value > 1:

 data[:, y] = 1

 x[i] = data

return x

def resample_by_interpolation(signal, input_fs, output_fs):

 scale = output_fs / input_fs

 # calculate new length of sample

 n = round(len(signal) * scale)

```

# use linear interpolation

# endpoint keyword means than linspace doesn't go all the way to 1.0

# If it did, there are some off-by-one errors

# e.g. scale=2.0, [1,2,3] should go to [1,1.5,2,2.5,3,3]

# but with endpoint=True, we get [1,1.4,1.8,2.2,2.6,3]

# Both are OK, but since resampling will often involve

# exact ratios (i.e. for 44100 to 22050 or vice versa)

# using endpoint=False gets less noise in the resampled sound

resampled_signal = np.interp(
    np.linspace(0.0, 1.0, n, endpoint=False), # where to interpret
    np.linspace(0.0, 1.0, len(signal), endpoint=False), # known positions
    signal, # known data points
)

return resampled_signal

```

```
def resample_axis(X, input_fs, output_fs, axis= 1):
```

```
    """Resamples an axis of a given numpy to have a given sample size. Uses resample from scipy
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    Parameters

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```
    X: numpy. The given numpy to resample.
```

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    sample_size: int. The sample size to resample the axis.

```

```

    axis: int. The axis to downsample.

```

```

    Returns

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```


np: Resampled numpy. False if not possible.

"""

```
X = np.apply_along_axis(func1d= resample_by_interpolation, axis= axis, arr=X, input_fs=
input_fs, output_fs= output_fs)
```

return X