People with no idea about AI, telling me my AI will destroy the world

Me wondering why my neural network is classifying a cat as a dog..



Source

# Welcome to Week 5 Lecture 2!

Data Science in Python & Machine Learning



# Assignments Due This Week by Sunday night

- Abalone Preprocessing Exercise (Core)
- 2. Pipelines Activity (Core)
- 3. Project 1 Part 5 (Core)

### Learning Goals:

- 1. Explain the three main types of variables we will be working with in this stack.
- 2. Convert data types into the correct format for machine learning
- 3. Implement data preparation in Python

# Warm Up

What are the 3 primary kinds of variable features we learned about in the last stack.

# Data Preprocessing

It must all be numbers.

#### **Data Preparation**

Interval/Continuous/Quantitative variables: Numbers that really mean numbers (int or float)

#### **Examples:**

- Values of length in cm: [2, 3.5, 8]
- Income in dollars: [57000, 38000]

**Solution:** No further preparation needed!

**Note**: Sometimes numeric data is in string form, such as '2'. This should be dealt with in data cleaning.

#### **Data Preparation**

Ordinal Categorical Variables: categories that can be ranked

#### Examples:

- low/medium/high,
- strongly disagree/disagree/agree/strongly agree
- \$5-10, \$11-15, \$16-20, >\$20

**Solution:** Convert to numbers representing the rank (sometimes your best judgment is required):

#### Example:

df[ordinal\_column].replace({'low':0, 'medium':1, 'high':2}, inplace=True)

#### **Data Preparation**

Categorical Nominal variables: different groups with no order.

NOTE: Sometimes these are represented as integers in the data, but should be interpreted as categories.

#### Examples:

- Yes/No,
- Male/Female/Nonbinary,
- Fat free/Regular
- Teacher A, Teacher B, Teacher C, Teacher D,

Solution: One-hot encoding

#### One hot encode

#### Each category becomes its own column

Genre		Comedy	Drama	Documentary	Scifi
Drama	-	0	1	0	0
Comedy	-	1	0	0	0
Drama		0	1	0	0
Documentary	-	0	0	1	0
SciFi		0	0	0	1

# One-Hot Encoding Can Cause Data Leakage

How can data from the test set leak into the training set with one-hot encoding?

Consider this:

What if when you split the data there are categories in a nominal column that are present in the test set, but not the training set?

If you encoded the column before splitting, there would be a column assigned to that category, even though it's not present in the training data!!

That's data leakage.

So, we encode after splitting.



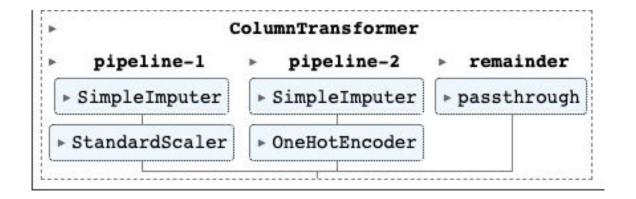
# Transforming Data in Python: SimpleImputer (imputes missing data)

Imputes missing values in many columns at once:

- 'mean', 'median', or 'mode' for numeric values
- 'most frequent', or constant for numeric and categorical values
- Can be used on a subset of rows.

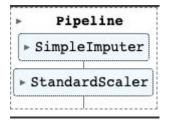
#### Make Column Transformer

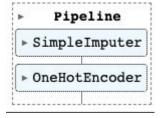
Make Column Transformer allows you to process numerical and categorical columns with the appropriate strategy. Can allow you to process different columns differently.

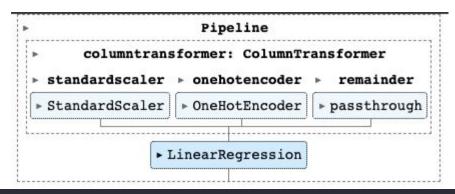


# Pipelines

Pipelines allow you to chain together steps after a train test split that would normally require several lines of code without it.







# Import Libraries

```
1 # Imports
2 import pandas as pd
3 from sklearn.model_selection import train_test_split
4 from sklearn.compose import make_column_selector, make_column_transformer
5 from sklearn.preprocessing import StandardScaler, OneHotEncoder
6 from sklearn.pipeline import make_pipeline
7 from sklearn.impute import SimpleImputer
```

#### Pipelines and Tuples

```
1 num selector = make column selector(dtype include='number')
 2 cat selector = make column selector(dtype include='object')
 3
 4 median imputer = SimpleImputer(strategy='median')
 5 freq imputer = SimpleImputer(strategy='most frequent')
 6
 7 scaler = StandardScaler()
 8 ohe encoder = OneHotEncoder(sparse=False, handle unknown='ignore')
 9
10 num pipe = make pipeline(median imputer, scaler)
11 cat pipe = make pipeline(freq imputer, ohe encoder)
12
13 num tuple = (num pipe, num selector)
14 cat tuple = (cat pipe, cat selector)
```

#### Column Transformer

make\_column\_transformer(cat\_tuple, remainder='passthrough')

By default ColumnTransformer **DROPS** any columns not specified in the steps.

```
[65] 1 col_trans = make_column_transformer(num_tuple, cat_tuple, remainder = 'passthrough')
```

# Now you know what you need to know to prepare a dataset for modeling!

### CodeAlong Notebook

CodeAlong DataSet

# Challenge!

# Tonight's Challenge Notebook

**Tonight's Challenge Dataset**