Exploratory Data Analysis

Plan:

- For each subtask we have training, validation and test sample sets with code, language, label and generator.
- We want to find interesting information about each programming language to see if we should treat the separately or if they are informative piece of data.
- We will use simple metrics to see if they are significant differences between them or not.
- After analysis we will get some conclusions

In [1]:
```
import pandas as pd
from pathlib import Path
import matplotlib.pyplot as plt

path = Path("SemEval-2026-Task13")
dir_paths = {name: path / f"task_{name}" for name in ["a", "b", "c"
datasets = {
    name: {
        "training": dir_paths[name] / f"task_{name}_training_set.pa
        if name == "b"
        else dir_paths[name] / f"task_{name}_training_set_1.parquet
        "validation": dir_paths[name] / f"task_{name}_validation_se
        "test": dir_paths[name] / f"task_{name}_test_set_sample.par
    }
    for name in ["a", "b", "c"]
}

data = pd.read_parquet(datasets["b"]["training"])
print(data.head())
data["word_count"] = data["code"].str.split().str.len()
```

```
                                                code generator  labe
l language
0  def load(config, filepath, token):\n    if con...     Human
0   Python
1  n = int(input())\narr = list(map(int, input()....     Human
0   Python
2  using Aow.Infrastructure.Domain;\nusing Aow.In...    GPT-4o     1
0       C#
3  def save_data(bot, force=False):\n    if bot.d...     Human
0   Python
4  def parse_metadata(metaurl, progress=1e5):\n  ...     Human
0   Python
```

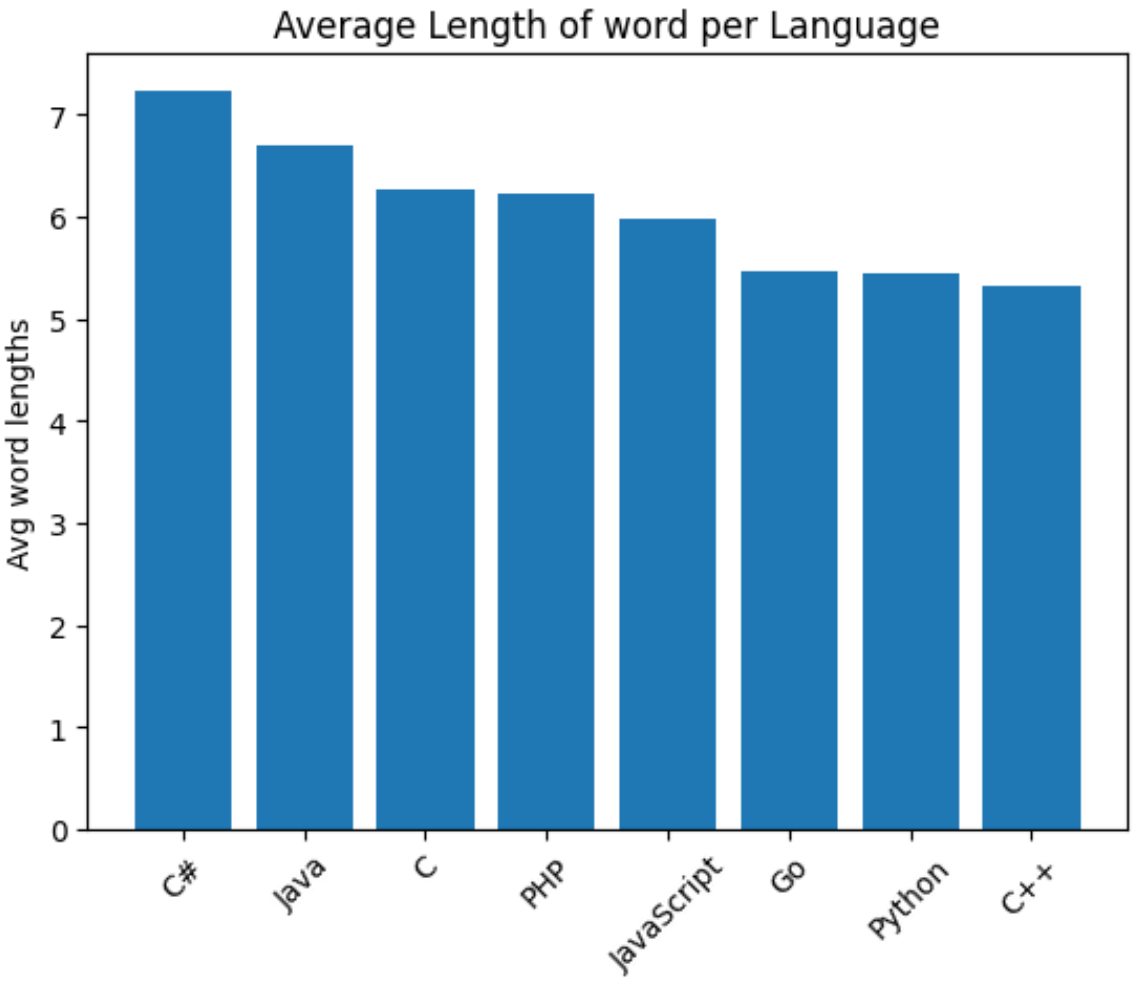
In [2]:
```
data_cut = data.head(10_000)
words = (
    data_cut.assign(word=data_cut["code"].str.lower().str.replace(r'
```

```
        .explode("word")
        .dropna(subset=["word"])
)

words = words[words["word"].str.strip() != ""]

top_words = (
    words.groupby(["language", "word"])
         .size()
         .reset_index(name="count")
)

top5 = (
    top_words.sort_values(["language", "count"], ascending=[True, Fa
             .groupby("language")
             .head(5)
)

print(top5)
```

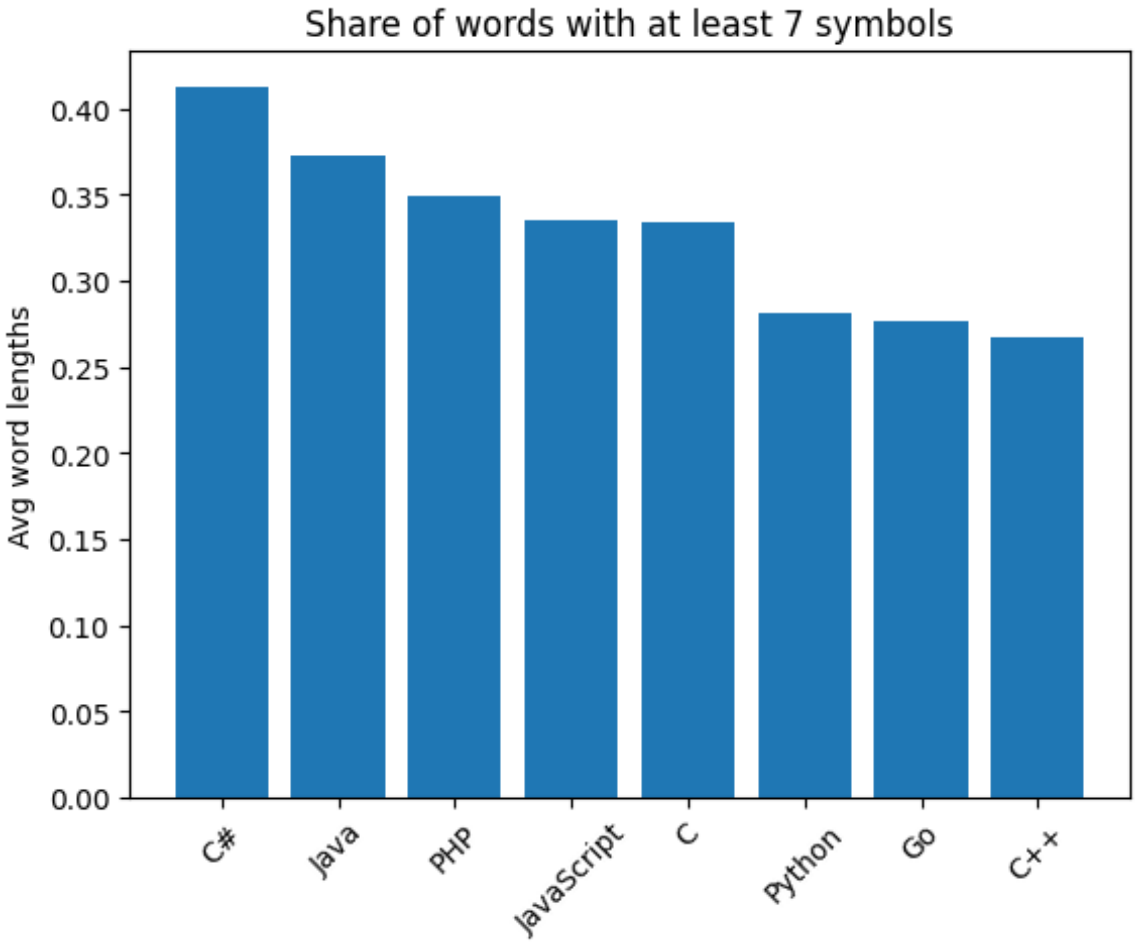
	language	word	count
4664	C	if	1603
0	C	0	1380
107	C	1	1046
8572	C	return	1024
4856	C	int	903
25487	C#	public	3937
20152	C#	if	2864
23453	C#	new	2575
26470	C#	return	2547
28266	C#	string	2515
36512	C++	i	3570
36864	C++	int	3068
30922	C++	0	2983
31083	C++	1	2187
36580	C++	if	2164
46526	Go	err	2441
47831	Go	if	1983
50896	Go	return	1885
49529	Go	nil	1616
47094	Go	func	1165
85778	Java	public	9974
95026	Java	the	9472
88100	Java	return	8640
93444	Java	string	8297
73858	Java	if	7997
110524	JavaScript	this	2690
102246	JavaScript	const	1982
105112	JavaScript	if	1686
108876	JavaScript	return	1673
104238	JavaScript	function	1365
124347	PHP	this	3491
114256	PHP	class	2345
122279	PHP	return	2207
116898	PHP	function	2113
121645	PHP	public	1812
151826	Python	self	9951
126008	Python	1	7314
140341	Python	if	6736
125750	Python	0	6139
140584	Python	in	5495

In [3]:
```
words = words.assign(word_length=words["word"].str.len())

avg_word_length = (
    words.groupby("language")["word_length"]
         .mean()
         .reset_index(name="avg_word_length")
         .sort_values("avg_word_length", ascending=False)
)

avg_word_length["avg_word_length"] = avg_word_length["avg_word_leng
plt.bar(avg_word_length["language"], avg_word_length["avg_word_leng
plt.xticks(rotation=45)
plt.title("Average Length of word per Language")
plt.ylabel("Avg word lengths")
```

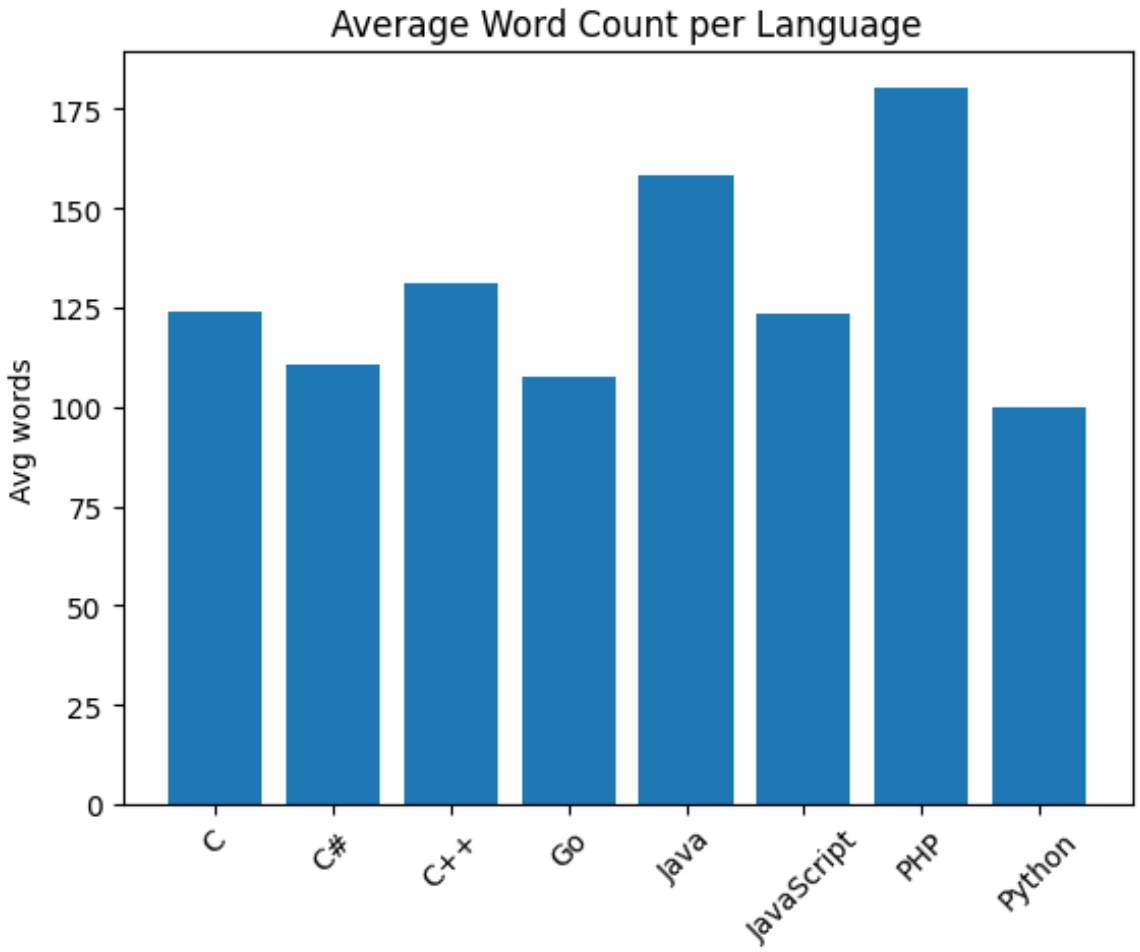
```
plt.show()
```



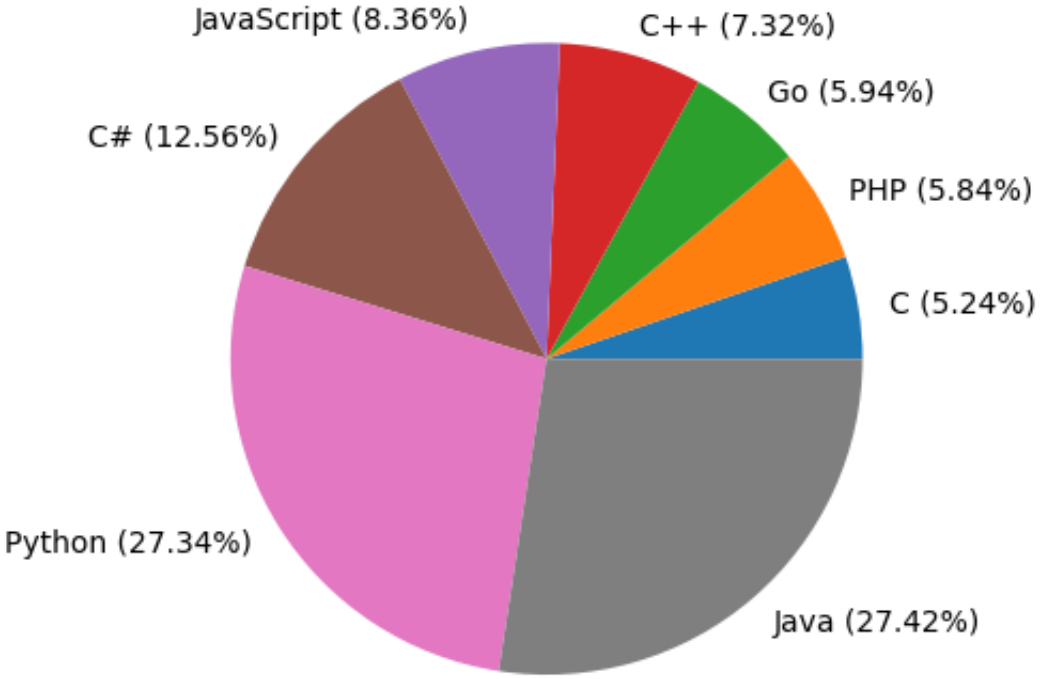
In [4]:
```
words = words.assign(word_length=words["word"].str.len())

words = words.assign(is_4plus = words["word_length"] >= 7)

avg_nplus = (
    words.groupby("language")["is_4plus"]
         .mean()
         .reset_index(name="avg_share_nplus")
         .sort_values("avg_share_nplus", ascending=False)
)

plt.bar(avg_nplus["language"], avg_nplus["avg_share_nplus"])
plt.xticks(rotation=45)
plt.title("Share of words with at least 7 symbols")
plt.ylabel("Avg word lengths")
plt.show()
```



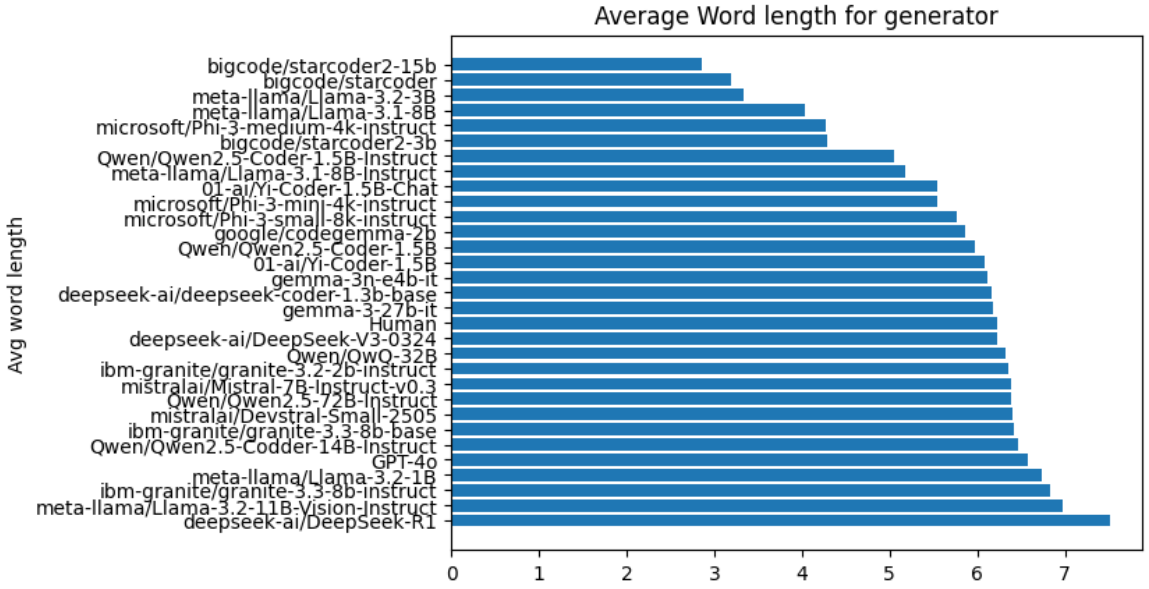
In [5]:
```
avg_words = data.groupby("language")["word_count"].mean().reset_ind

plt.bar(avg_words["language"], avg_words["word_count"])
plt.xticks(rotation=45)
plt.title("Average Word Count per Language")
plt.ylabel("Avg words")
plt.show()
```



In [6]:
```
language_count = data.groupby("language").count().sort_values("code'
total = language_count['code'].sum()
labels = [f'{name} ({value/total*100:.2f}%)' for name, value in zip
plt.pie(language_count['code'], labels=labels)
plt.show()
```



In [7]:
```
words = words.assign(word_length=words["word"].str.len())

avg_word_length_gen = (
    words.groupby("generator")["word_length"]
         .mean()
         .reset_index(name="avg_word_length")
         .sort_values("avg_word_length", ascending=False)
)

plt.barh(avg_word_length_gen["generator"], avg_word_length_gen["avg
plt.title("Average Word length for generator")
plt.ylabel("Avg word length")
plt.show()
```



Conclusions

Basic metrics showed that there is visible difference between language when it comes to average number of words or just a number of words in code snippet.

Because of size of dataset, it was more reasonable to load one dataset at a time and python garbage collection is debatable, so graphs from above won't tell all information gathered because of this EDA, try changing the dataset parameter from above to see how where the counts in other datasets.

But in summary we can see:

- In dataset A human used wrote smaller words than AI models
- Many languages have different word count and word lengths than other so it would be reasonable to use this information later on
- The average lengths of words used by different generators is valuable information, so if we would get large enough batches of code where each batch has the same generator we should be able to with relatively high accuracy determine what generator was used
- Different languages uses different vocabulary, so there are different popular words