

wasm-js-energy-study-analysis

September 28, 2023

```
[1]: pip install pandas matplotlib seaborn numpy scipy statsmodels scikits.bootstrap
```

Collecting pandas

Downloading

pandas-2.1.1-cp39-cp39-manylinux_2_17_aarch64.manylinux2014_aarch64.whl (14.9 MB)

14.9/14.9 MB

31.6 MB/s eta 0:00:0000:0100:01

Collecting matplotlib

Downloading

matplotlib-3.8.0-cp39-cp39-manylinux_2_17_aarch64.manylinux2014_aarch64.whl (11.4 MB)

11.4/11.4 MB

34.2 MB/s eta 0:00:0000:0100:01

Collecting seaborn

Downloading seaborn-0.12.2-py3-none-any.whl (293 kB)

293.3/293.3 KB

43.1 MB/s eta 0:00:00

Collecting numpy

Downloading

numpy-1.26.0-cp39-cp39-manylinux_2_17_aarch64.manylinux2014_aarch64.whl (14.2 MB)

14.2/14.2 MB

33.0 MB/s eta 0:00:0000:0100:01

Collecting scipy

Downloading

scipy-1.11.3-cp39-cp39-manylinux_2_17_aarch64.manylinux2014_aarch64.whl (33.0 MB)

33.0/33.0 MB

25.7 MB/s eta 0:00:0000:0100:01

Collecting statsmodels

Downloading

statsmodels-0.14.0-cp39-cp39-manylinux_2_17_aarch64.manylinux2014_aarch64.whl (9.8 MB)

9.8/9.8 MB

5.6 MB/s eta 0:00:0000:0100:01m

Collecting scikits.bootstrap

Downloading scikits.bootstrap-1.1.0-py2.py3-none-any.whl (14 kB)

```

Requirement already satisfied: python-dateutil>=2.8.2 in
/opt/conda/lib/python3.9/site-packages (from pandas) (2.8.2)
Collecting tzdata>=2022.1
  Downloading tzdata-2023.3-py2.py3-none-any.whl (341 kB)
                                341.8/341.8 KB
33.1 MB/s eta 0:00:00
Requirement already satisfied: pytz>=2020.1 in
/opt/conda/lib/python3.9/site-packages (from pandas) (2022.1)
Requirement already satisfied: importlib-resources>=3.2.0 in
/opt/conda/lib/python3.9/site-packages (from matplotlib) (5.6.0)
Collecting cyclor>=0.10
  Downloading cyclor-0.11.0-py3-none-any.whl (6.4 kB)
Collecting pillow>=6.2.0
  Downloading Pillow-10.0.1-cp39-cp39-manylinux_2_28_aarch64.whl (3.5 MB)
                                3.5/3.5 MB
24.7 MB/s eta 0:00:0000:0100:01
Requirement already satisfied: packaging>=20.0 in
/opt/conda/lib/python3.9/site-packages (from matplotlib) (21.3)
Requirement already satisfied: pyparsing>=2.3.1 in
/opt/conda/lib/python3.9/site-packages (from matplotlib) (3.0.7)
Collecting contourpy>=1.0.1
  Downloading
contourpy-1.1.1-cp39-cp39-manylinux_2_17_aarch64.manylinux2014_aarch64.whl (286
kB)
                                286.4/286.4 KB
30.2 MB/s eta 0:00:00
Collecting fonttools>=4.22.0
  Downloading
fonttools-4.42.1-cp39-cp39-manylinux_2_17_aarch64.manylinux2014_aarch64.whl (4.5
MB)
                                4.5/4.5 MB
14.1 MB/s eta 0:00:00a 0:00:01
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MB)
                                1.4/1.4 MB
28.0 MB/s eta 0:00:00a 0:00:01
Collecting patsy>=0.5.2
  Downloading patsy-0.5.3-py2.py3-none-any.whl (233 kB)
                                233.8/233.8 KB
29.3 MB/s eta 0:00:00
Collecting typing-extensions
  Downloading typing_extensions-4.8.0-py3-none-any.whl (31 kB)
Collecting pyerf
  Downloading pyerf-1.0.1-py3-none-any.whl (11 kB)
Requirement already satisfied: zipp>=3.1.0 in /opt/conda/lib/python3.9/site-
packages (from importlib-resources>=3.2.0->matplotlib) (3.7.0)

```

Requirement already satisfied: six in /opt/conda/lib/python3.9/site-packages (from patsy>=0.5.2->statsmodels) (1.16.0)
Installing collected packages: pyperf, tzdata, typing-extensions, pillow, numpy, kiwisolver, fonttools, cyclor, scipy, scikits.bootstrap, patsy, pandas, contourpy, statsmodels, matplotlib, seaborn
Successfully installed contourpy-1.1.1 cyclor-0.11.0 fonttools-4.42.1 kiwisolver-1.4.5 matplotlib-3.8.0 numpy-1.26.0 pandas-2.1.1 patsy-0.5.3 pillow-10.0.1 pyperf-1.0.1 scikits.bootstrap-1.1.0 scipy-1.11.3 seaborn-0.12.2 statsmodels-0.14.0 typing-extensions-4.8.0 tzdata-2023.3
Note: you may need to restart the kernel to use updated packages.

```
[2]: import os
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import numpy
from scipy import stats
import scipy
import helper.cliffs_delta as cliff
import statsmodels.api as sm
import itertools
import scikits.bootstrap as boot
```

1 Table of contents

1. RQ1: JS vs. WASM Energy Consumption
2. RQ2: JS Browsers Energy Consumption
3. RQ2: WASM Browsers Energy Consumption

2 Prepare Data

```
[3]: sns.set(rc = {'figure.figsize':(10,10)})
confidence_level = 0.95

df = pd.read_csv('./data/energy.csv', sep=';', quotechar='"', names=["browser", "language", "algorithm", "device", "energy"])
df['energy_total'] = df['energy'].astype(float)
df['implementation'] = numpy.where(df['language'] == 'js', 'js', 'wasm')

## unique values in columns
browsers = numpy.sort(df['browser'].unique())
browserpairs = list(itertools.combinations(browsers, 2))
languages = df['language'].unique()
implementations = numpy.sort(df['implementation'].unique())
implementationpairs = list(itertools.combinations(implementations, 2))
languagepairs = list(itertools.combinations(languages, 2))
```

```

algorithms = df['algorithm'].unique()
devices = df['device'].unique()

df.head()

```

```

[3]:  browser language          algorithm  device  energy  energy_total  \
0  firefox      c          fibonacci SM-G991B  6.531442      6.531442
1   chrome      c    humblenumbers SM-G991B  5.787359      5.787359
2   chrome     js matrixmultiplication SM-G991B  9.739240      9.739240
3   chrome     js    seqnonsquares SM-G991B  4.516185      4.516185
4   chrome     js         nqueens SM-G991B  5.935186      5.935186

    implementation
0           wasm
1           wasm
2            js
3            js
4            js

```

2.1 Count samples

```

[4]: data = []

for browser in browsers:
    for language in languages:
        for algorithm in algorithms:
            for device in devices:
                #print(browser,device,numpy.mean(df[(df['browser'] == browser) &
                ↪ (df['language'] == language) & (df['algorithm'] == algorithm) &
                ↪ (df['device'] == device)][ 'energy']))
                data.append(
                    [browser, language, algorithm, device, df[(df['browser'] ==
                    ↪ browser) & (df['language'] == language) & (df['algorithm'] == algorithm) &
                    ↪ (df['device'] == device)][ 'energy_total'].count()
                ])

# Create the pandas DataFrame
count = pd.DataFrame(data, columns = ['browser', 'language', 'algorithm',
    ↪ 'device', 'count'])
display(count)

```

```

    browser language          algorithm  device  count
0   chrome      c          fibonacci SM-G991B     30
1   chrome      c          fibonacci  Nexus 5     32
2   chrome      c    humblenumbers SM-G991B     31
3   chrome      c    humblenumbers  Nexus 5     34
4   chrome      c matrixmultiplication SM-G991B     31

```

..
147	firefox	js	happynumbers	Nexus 5	31
148	firefox	js	pancakesort	SM-G991B	31
149	firefox	js	pancakesort	Nexus 5	33
150	firefox	js	quicksort	SM-G991B	32
151	firefox	js	quicksort	Nexus 5	33

[152 rows x 5 columns]

2.2 Calculations

```
[5]: data = []
for implementation in implementations:
    x = df[(df['implementation'] == implementation)]
    cnt = x.size
    sum = numpy.round(numpy.sum(x['energy']), 2)

    data.append(
        [implementation, cnt, sum]
    )

# Create the pandas DataFrame
stat = pd.DataFrame(data, columns = ['implementation', 'count', 'sum'])
print(stat.to_string())
```

	implementation	count	sum
0	js	16877	24490.76
1	wasm	16905	18937.81

3 Violinplot (Algorithms)

```
[6]: for device in devices:
    for browser in browsers:
        data = df[(df['device'] == device) & (df['browser'] == browser)].
        ↪sort_values(by=['algorithm'])
        plt.figure()
        plt.xticks(rotation=90)
        plt.ylim(0, 35)
        sns.violinplot(x='algorithm', y='energy', hue='implementation',
        ↪hue_order=implementations, data=data, palette='colorblind', split=True).
        ↪set_title(device + " - " + browser)
```

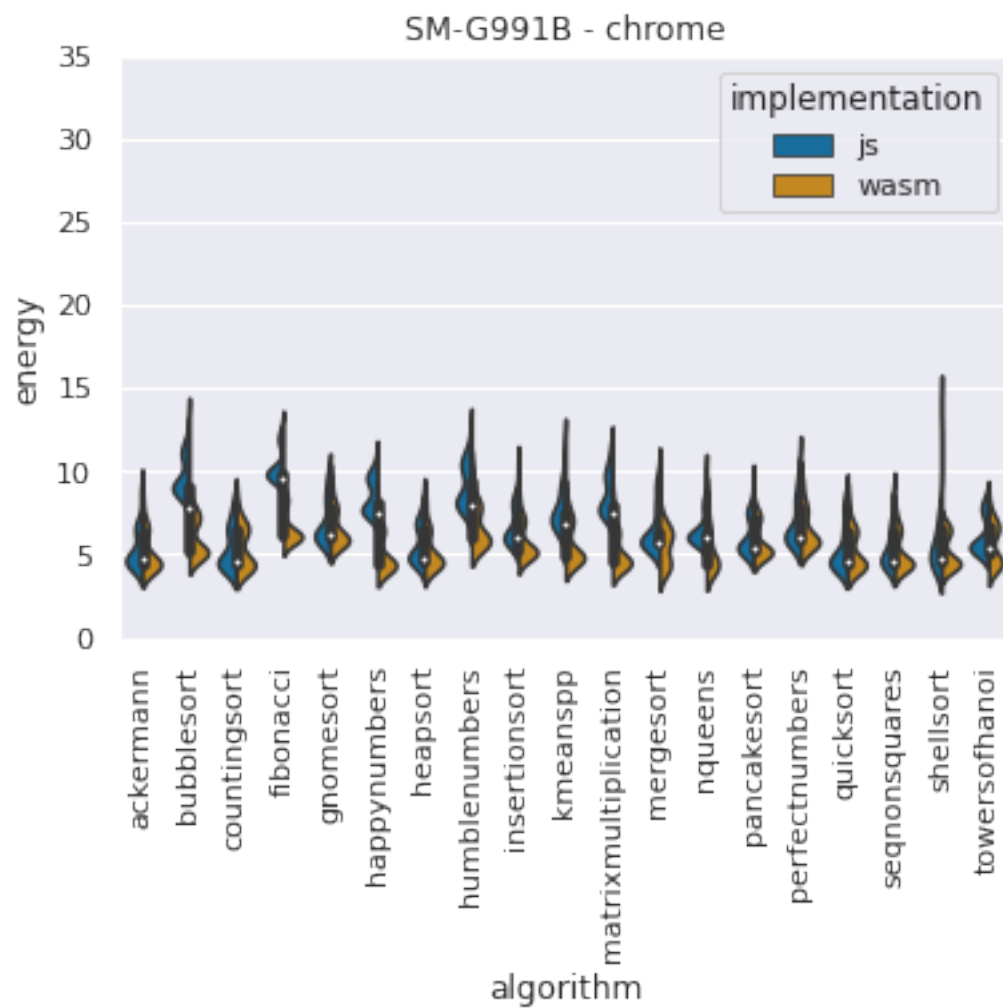
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning: is_categorical_dtype is deprecated and will be removed in a future version. Use isinstance(dtype, CategoricalDtype) instead

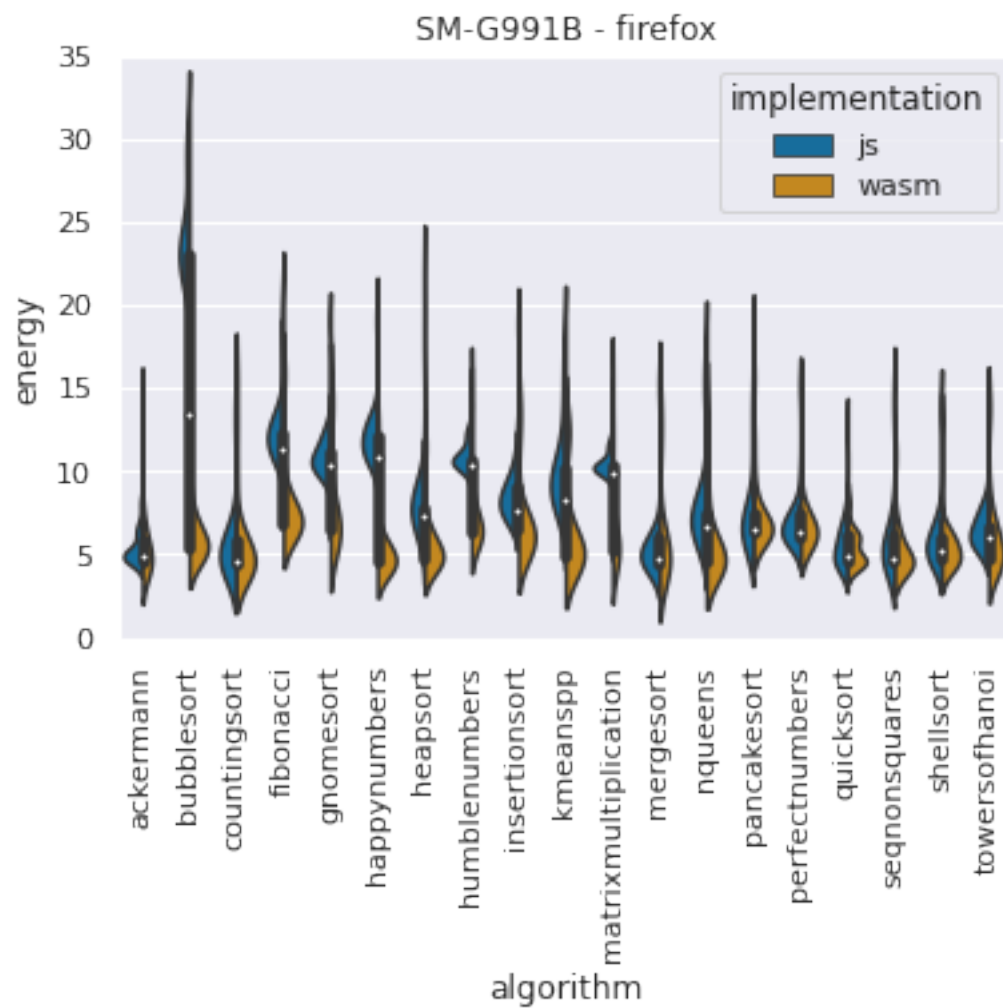
```
if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
```

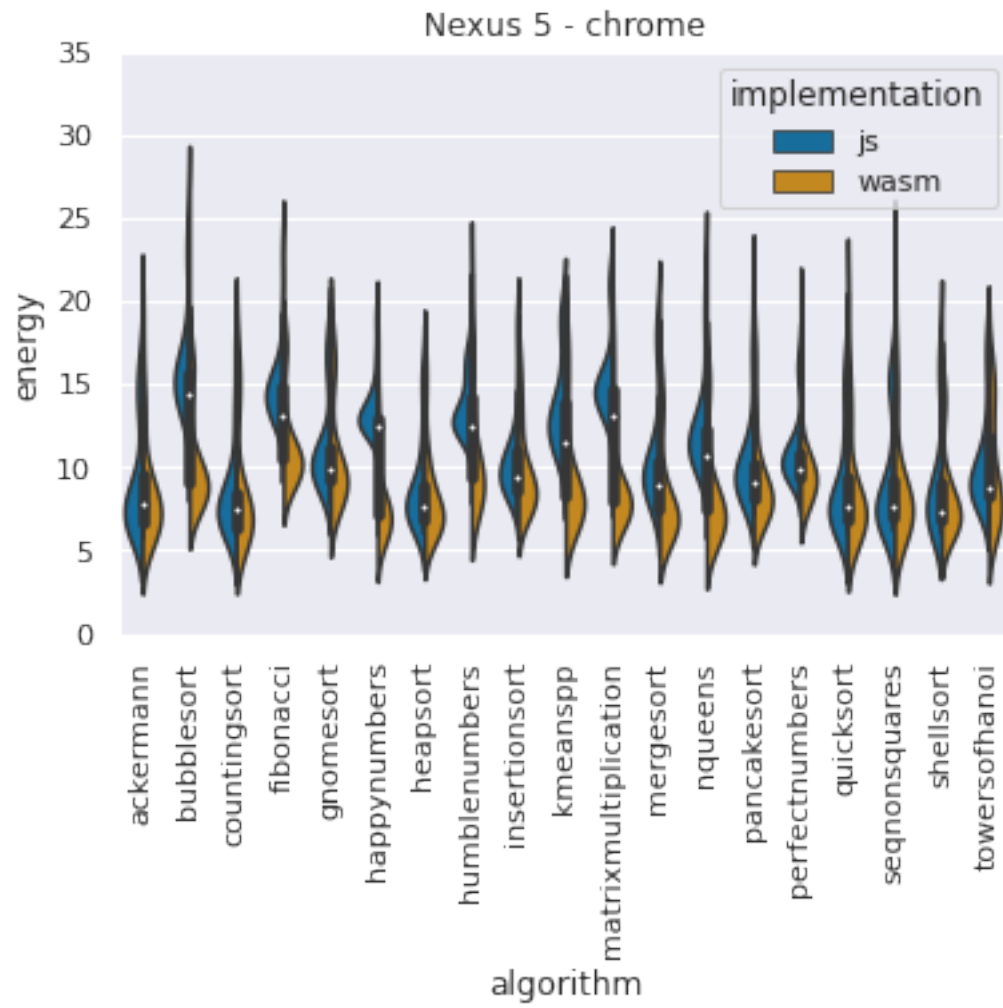
```

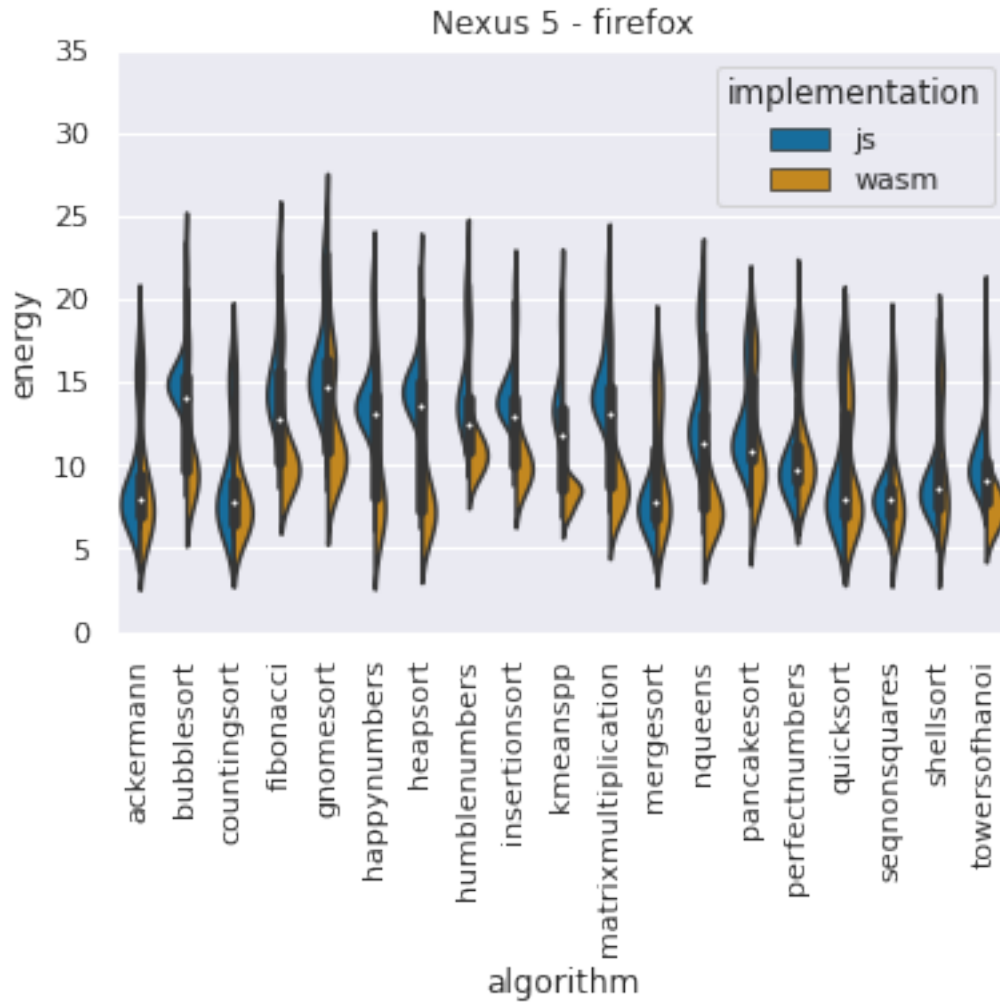
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
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/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
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/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
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/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
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is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):

```









4 RQ1: JavaScript vs. WebAssembly

4.1 Shapiro Wilk Test

```
[7]: data = []
non_normal = 0

for implementation in implementations:
    energy = df[(df['implementation'] == implementation)]['energy']

    if len(energy) >= 3:
        shapiro_test = stats.shapiro(energy)

        non_normal += (1 if shapiro_test.pvalue <= 0.05 else 0)
```

```

        data.append(
            [implementation,
             shapiro_test.statistic,
             shapiro_test.pvalue
            ]
        )

# Create the pandas DataFrame
swt = pd.DataFrame(data, columns = ['implementation', 'w', 'p'])
#print(swt.to_string())
display(swt)

print("\n{n} non-normally distributed samples".format(non_normal))
print("{n} normally distributed samples".format(len(swt) - non_normal))
print("{:.2f}% non-normally distributed samples".format(non_normal/
    ↪len(swt)*100))

```

	implementation	w	p
0	js	0.927419	1.147179e-32
1	wasm	0.863110	1.110529e-41

```

2 non-normally distributed samples
0 normally distributed samples
100.00% non-normally distributed samples

```

4.2 Shapiro Wilk Test (By Device)

```

[8]: data = []
non_normal = 0

for device in devices:
    for implementation in implementations:
        energy = df[(df['implementation'] == implementation) & (df['device'] == ↪
            ↪device)][['energy']]

        if len(energy) >= 3:
            shapiro_test = stats.shapiro(energy)

            non_normal += (1 if shapiro_test.pvalue <= 0.05 else 0)

        data.append(
            [
                device,
                implementation,
                shapiro_test.statistic,
                shapiro_test.pvalue
            ]
        )

```

```

    ]
)

# Create the pandas DataFrame
swt = pd.DataFrame(data, columns = ['device', 'implementation', 'w', 'p'])
#print(swt.to_string())
display(swt)

print("\n{n} non-normally distributed samples".format(non_normal))
print("{ normally distributed samples".format(len(swt) - non_normal))
print("{:.2f}% non-normally distributed samples".format(non_normal/
↳len(swt)*100))

```

	device	implementation	w	p
0	SM-G991B	js	0.771749	1.073380e-37
1	SM-G991B	wasm	0.747010	3.835847e-39
2	Nexus 5	js	0.960188	8.760830e-18
3	Nexus 5	wasm	0.857640	6.419627e-32

```

4 non-normally distributed samples
0 normally distributed samples
100.00% non-normally distributed samples

```

4.3 Mann-Whitney-U-Test

```

[9]: data = []

for implementationpair in implementationpairs:
    impl1_energy = df[(df['implementation'] == implementationpair[0])]['energy']
    impl2_energy = df[(df['implementation'] == implementationpair[1])]['energy']
    eff = cliff.cliffs_delta(impl1_energy, impl2_energy)

    u = stats.mannwhitneyu(impl1_energy, impl2_energy, alternative='two-sided')

    data.append(
        [
            implementationpair[0] + ' vs. ' + implementationpair[1],
            u.statistic,
            u.pvalue,
            eff[0],
            eff[1]
        ]
    )

# Create the pandas DataFrame
ut = pd.DataFrame(data, columns = ['implementation', 'u', 'p', 'eff', 'interp'])
display(ut)

```

```

interp = ut['interp'].value_counts()
interp = pd.DataFrame(interp, columns = ['interp', 'percent'])
interp['percent'] = (interp['interp'] / interp['interp'].sum()) * 100

display(interp)

```

```

implementation      u          p      eff  interp
0      js vs. wasm  3904820.0  1.173223e-93  0.341271  medium

```

Empty DataFrame

Columns: [interp, percent]

Index: []

4.4 Violinplot

```

[10]: data = df
plt.ylim(0, 35)
vp = sns.violinplot(x='implementation', y='energy', hue='implementation',
    ↪ hue_order=implementations, data=data, palette='colorblind',dodge=False)
vp.set_title("Energy Consumption by Implementations - " + device + " - " +
    ↪ browser)
vp.set_ylabel("Energy (Joules)")
plt.show()

```

/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning: is_categorical_dtype is deprecated and will be removed in a future version. Use isinstance(dtype, CategoricalDtype) instead

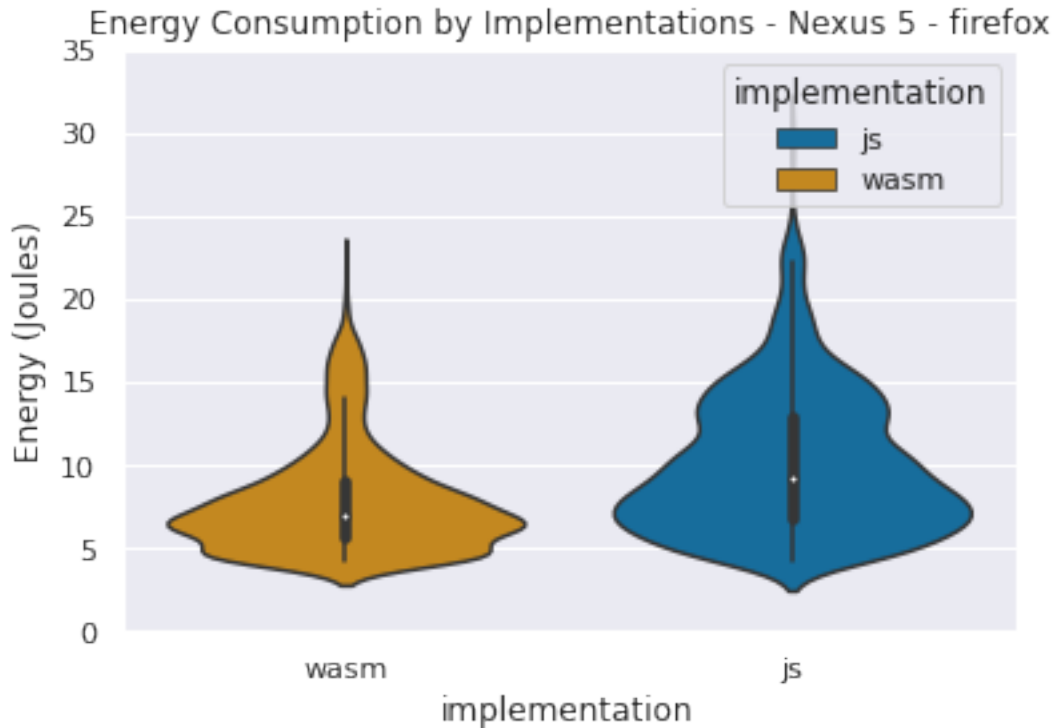
```
if pd.api.types.is_categorical_dtype(vector):
```

/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning: is_categorical_dtype is deprecated and will be removed in a future version. Use isinstance(dtype, CategoricalDtype) instead

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if pd.api.types.is_categorical_dtype(vector):
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```
if pd.api.types.is_categorical_dtype(vector):
```



4.5 Violinplot (By Browser)

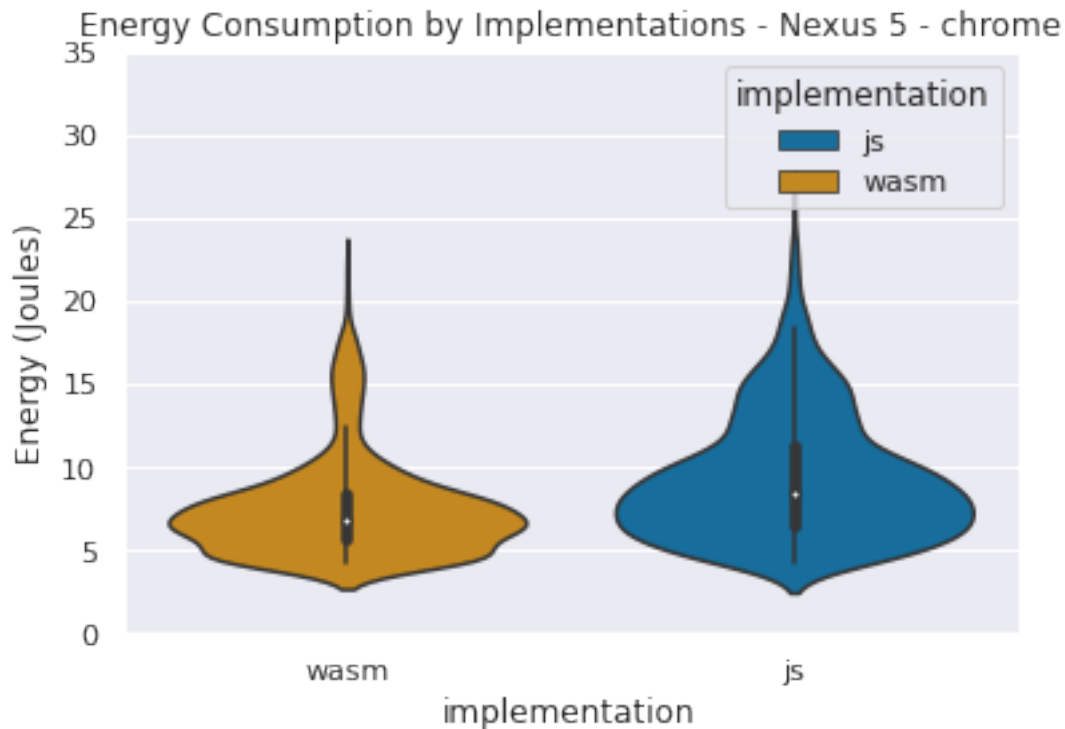
```
[11]: data = []
for browser in browsers:
    data = df[(df['browser'] == browser)]
    plt.ylim(0, 35)
    vp = sns.violinplot(x='implementation', y='energy', hue='implementation',
        hue_order=implementations, data=data, palette='colorblind',dodge=False)
    vp.set_title("Energy Consumption by Implementations - " + device + " - " +
        browser)
    vp.set_ylabel("Energy (Joules)")
    plt.show()
```

```
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
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    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
```

```

isinstance(dtype, CategoricalDtype) instead
if pd.api.types.is_categorical_dtype(vector):

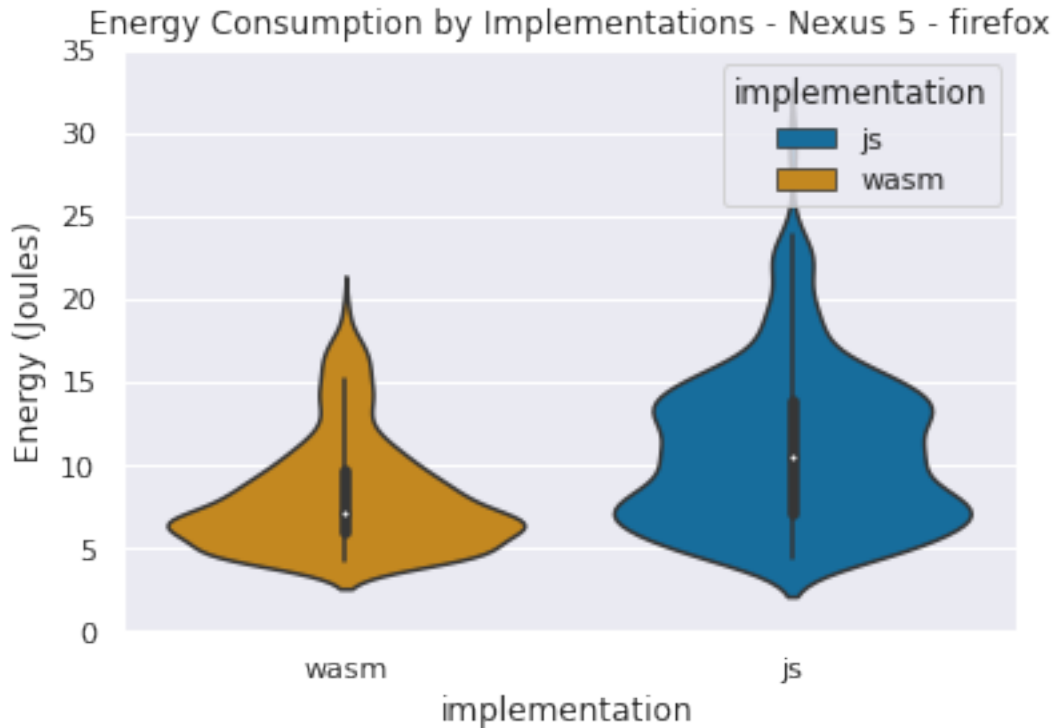
```



```

/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
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is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):

```



4.6 Violinplot (By Browser & By Device)

```
[12]: data = []
for device in devices:
    for browser in browsers:
        data = df[(df['browser'] == browser) & (df['device'] == device)].
        ↪sort_values(by=['implementation'])
        plt.ylim(0, 35)
        vp = sns.violinplot(x='implementation', y='energy',
        ↪hue='implementation', hue_order=implementations, data=data,
        ↪palette='colorblind', dodge=False)
        vp.set_title(device + " - " + browser)
        vp.set_ylabel("Energy (Joules)")
        plt.show()
```

/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning: is_categorical_dtype is deprecated and will be removed in a future version. Use isinstance(dtype, CategoricalDtype) instead

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isinstance(dtype, CategoricalDtype) instead
```

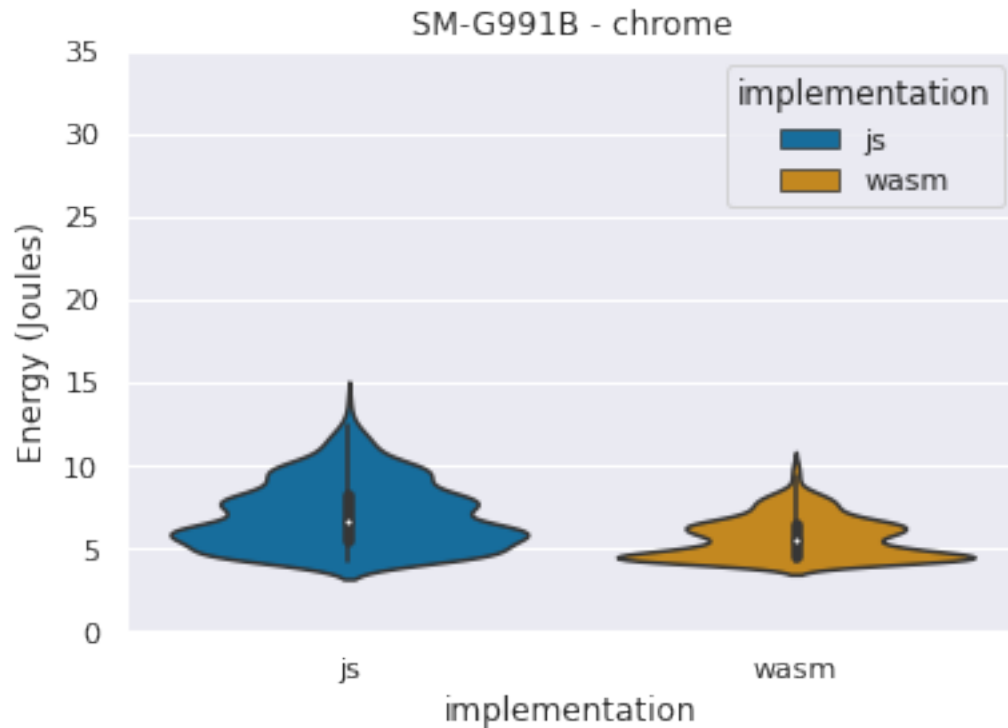
```
if pd.api.types.is_categorical_dtype(vector):
```



```

/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):

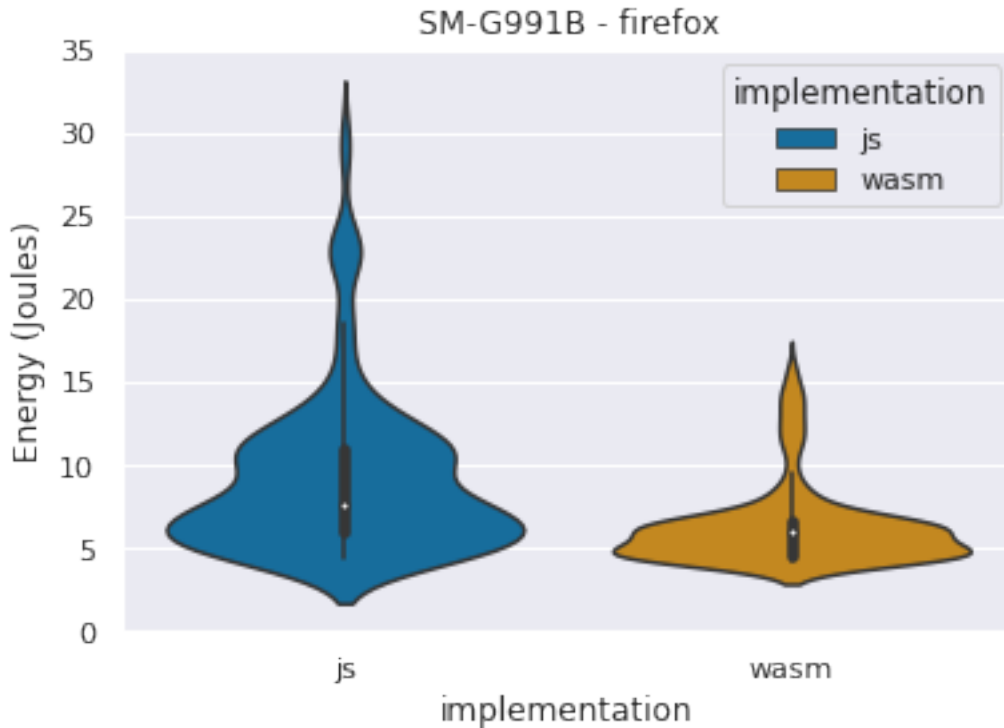
```



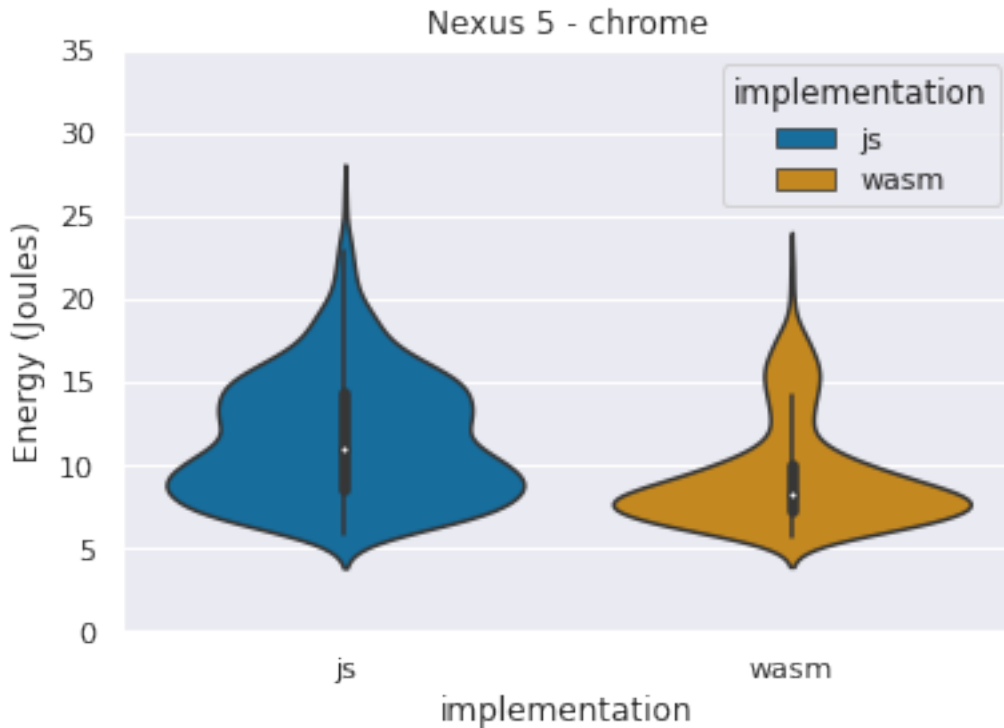
```

/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
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isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):

```



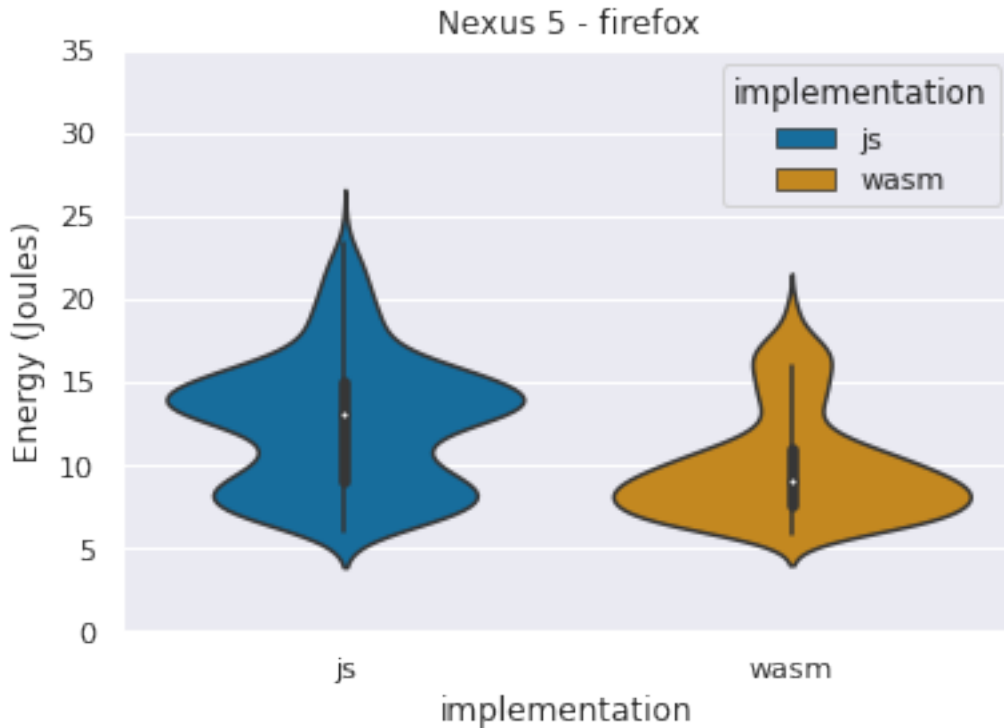
```
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
```



```

/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):

```



```
[13]: data = []
index=0
fig, axes = plt.subplots(1, 4, figsize=(20, 5))

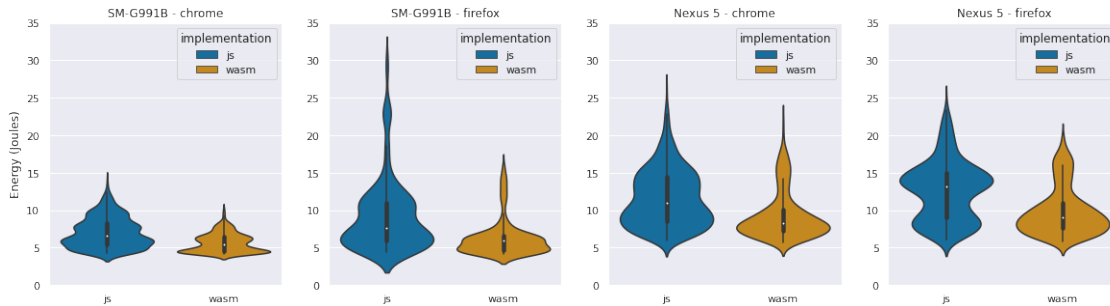
for device in devices:
    for browser in browsers:
        data = df[(df['browser'] == browser) & (df['device'] == device)].
        sort_values(by=['implementation'])
        vp = sns.violinplot(x='implementation', y='energy',
            hue='implementation', hue_order=implementations, data=data,
            palette='colorblind', dodge=False, ax=axes[index])
        vp.set_title(device + " - " + browser)
        axes[index].set_ylim(0, 35)
        axes[index].set_xlabel("")
        if index == 0:
            axes[0].set_ylabel("Energy (Joules)")
        else:
            axes[index].set_ylabel("")
        index+=1
```

/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning: is_categorical_dtype is deprecated and will be removed in a future version. Use isinstance(dtype, CategoricalDtype) instead

```

    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):

```



4.7 Violinplot (By Browser & Low End Device)

```
[14]: data = []

index=0
fig, axes = plt.subplots(1, 2, figsize=(18, 8))
for browser in browsers:
    data = df[(df['browser'] == browser) & (df['device'] == 'Nexus 5')].
    ↪sort_values(by=['implementation'])
    vp = sns.violinplot(x='implementation', y='energy', hue='implementation',
    ↪hue_order=implementations, data=data, palette='colorblind', dodge=False,
    ↪ax=axes[index])
    vp.set_title("Nexus 5 - " + browser)
    axes[index].set_ylim(0, 35)
    axes[index].set_xlabel("")
    if index == 0:
        axes[0].set_ylabel("Energy (Joules)")
    else:
        axes[index].set_ylabel("")
    index+=1
```

/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning: is_categorical_dtype is deprecated and will be removed in a future version. Use isinstance(dtype, CategoricalDtype) instead

```
if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
```

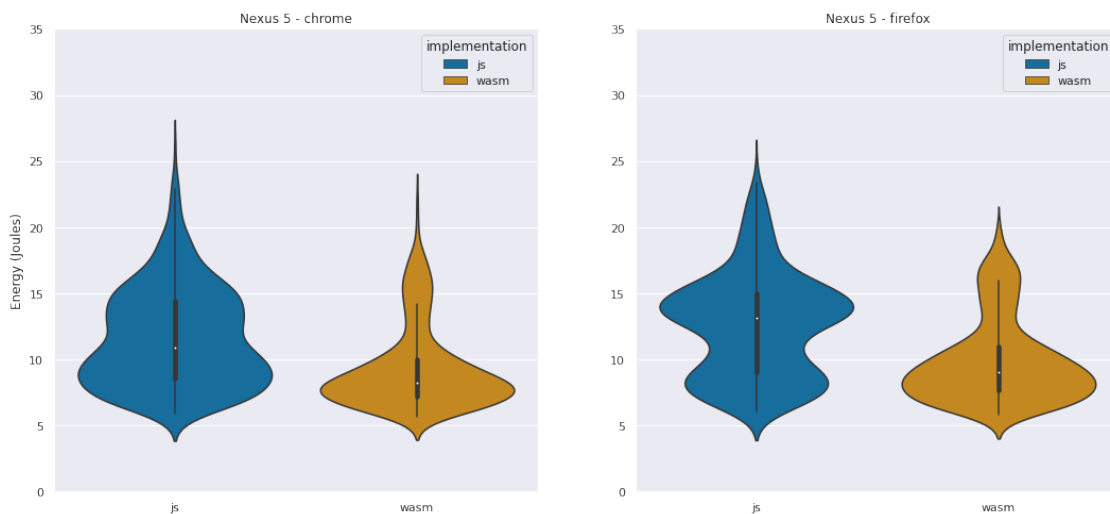
```
if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
```

```
if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
```

```

isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):

```



4.8 Violinplot (By Browser & High End Device)

```

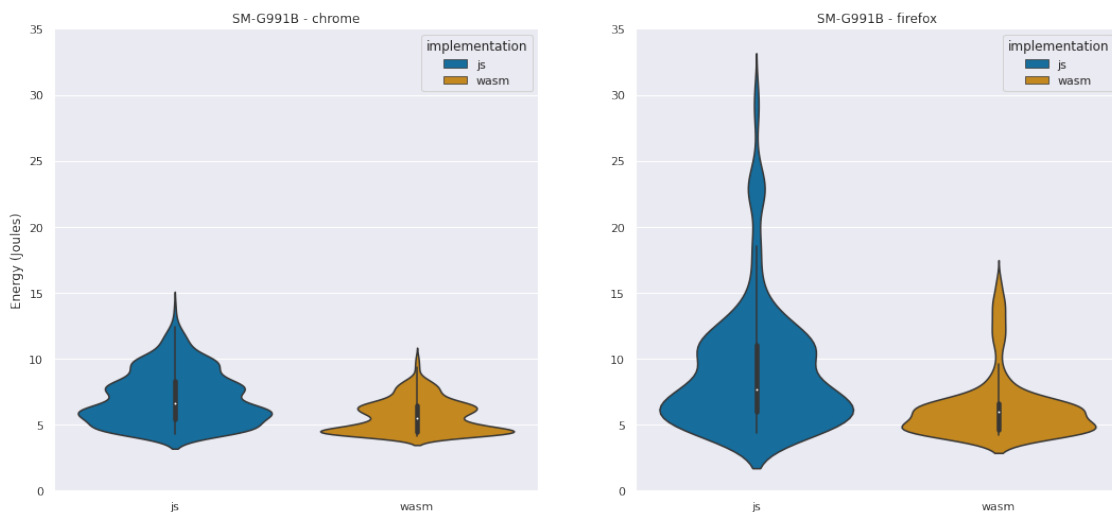
[15]: data = []

index=0
fig, axes = plt.subplots(1, 2, figsize=(18, 8))
for browser in browsers:
    data = df[(df['browser'] == browser) & (df['device'] == 'SM-G991B')].
    ↪sort_values(by=['implementation'])
    vp = sns.violinplot(x='implementation', y='energy', hue='implementation',
    ↪hue_order=implementations, data=data, palette='colorblind', dodge=False,
    ↪ax=axes[index])
    vp.set_title("SM-G991B - " + browser)
    axes[index].set_ylim(0, 35)
    axes[index].set_xlabel("")
    if index == 0:
        axes[0].set_ylabel("Energy (Joules)")
    else:
        axes[index].set_ylabel("")

```

```
index+=1
plt.show()
```

```
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
```



4.9 Violinplot (By Browser & Device Types)

```
[16]: ## Prepare naming of values for final violinplot
df_violin = df.copy()
df_violin['implementation'] = df_violin['implementation'].replace(['wasm'], 'Wasm')
df_violin['implementation'] = df_violin['implementation'].replace(['js'], 'JS')
df_violin['browser'] = df_violin['browser'].replace(['chrome'], 'Chrome')
df_violin['browser'] = df_violin['browser'].replace(['firefox'], 'Firefox')
df_violin['device'] = df_violin['device'].replace(['SM-G991B'], 'Samsung Galaxy S21')
browsers_violin = numpy.sort(df_violin['browser'].unique())
devices_violin = numpy.sort(df_violin['device'].unique())
```

```
[17]: sns.set(font_scale=2)
data = []
index=0
fig, axes = plt.subplots(1, 2, figsize=(18, 8))
for device in devices_violin:
    data = df_violin[(df_violin['device'] == device)].
    sort_values(by=['implementation'])
    vp = sns.violinplot(x='implementation', y='energy', hue='browser',
    hue_order=browsers_violin, data=data, palette='colorblind', dodge=False,
    ax=axes[index], split=True)
    vp.set_title(device)
    axes[index].legend(title="Browsers")
    axes[index].set_ylim(0, 35)
    axes[index].set_xlabel("")
    axes[index].set_ylabel("Energy Consumption (Joules)")
    if index == 0:
        axes[0].set_ylabel("Energy Consumption (Joules)")
    else:
        axes[index].set_ylabel("")
        axes[index].set_yticklabels([])
    index+=1
```

/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning: is_categorical_dtype is deprecated and will be removed in a future version. Use isinstance(dtype, CategoricalDtype) instead

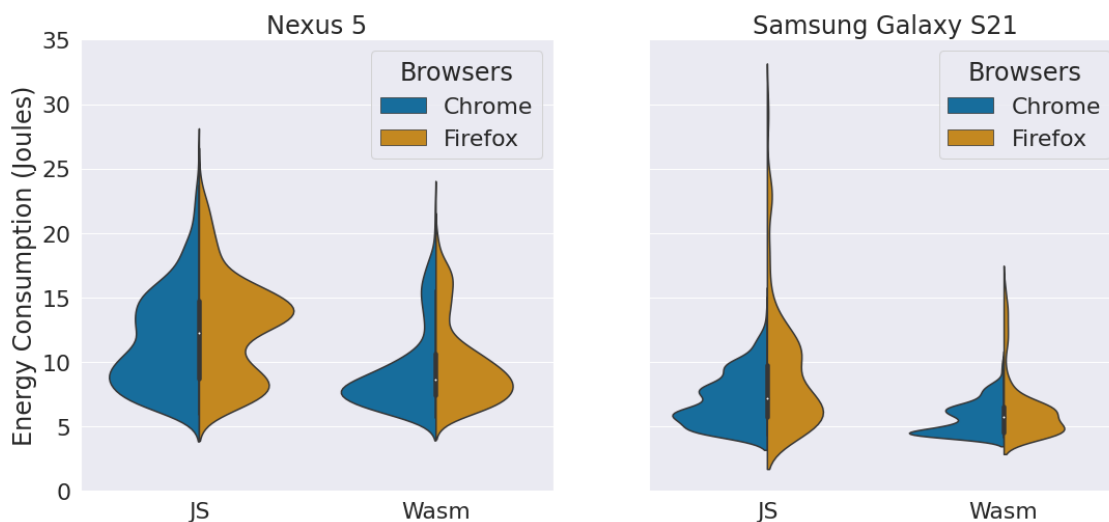
```
if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning: is_categorical_dtype is deprecated and will be removed in a future version. Use isinstance(dtype, CategoricalDtype) instead
```

```
if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning: is_categorical_dtype is deprecated and will be removed in a future version. Use isinstance(dtype, CategoricalDtype) instead
```

```

if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):

```



4.10 Histogram

```

[18]: sns.histplot(data=df, x="energy", hue="implementation",
    ↪hue_order=implementations).set_title("Energy Consumption by Implementations")
plt.xlabel("Energy (Joules)")
plt.ylim(0, 400)
plt.xlim(0, 40)

```

```

/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):

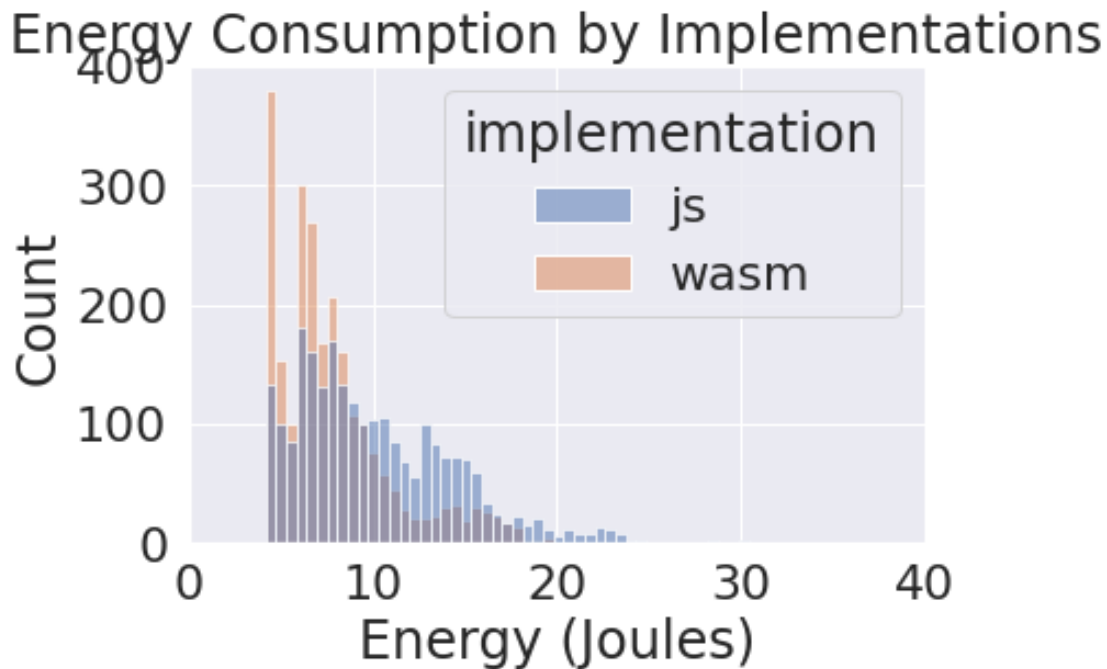
```

```

/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1119: FutureWarning:
use_inf_as_na option is deprecated and will be removed in a future version.
Convert inf values to NaN before operating instead.
    with pd.option_context('mode.use_inf_as_na', True):

```

[18]: (0.0, 40.0)



4.11 Histogram (By Device)

```

[19]: data = []
      for device in devices:
          data = df[(df['device'] == device)]
          plt.xlabel("Energy (Joules)")
          plt.ylim(0, 400)
          plt.xlim(0, 40)
          sns.histplot(data=data, x="energy", hue="implementation",
↪ hue_order=implementations).set_title("Energy Consumption by Implementations_
↪ " + device)
          plt.show()

```

```

/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:

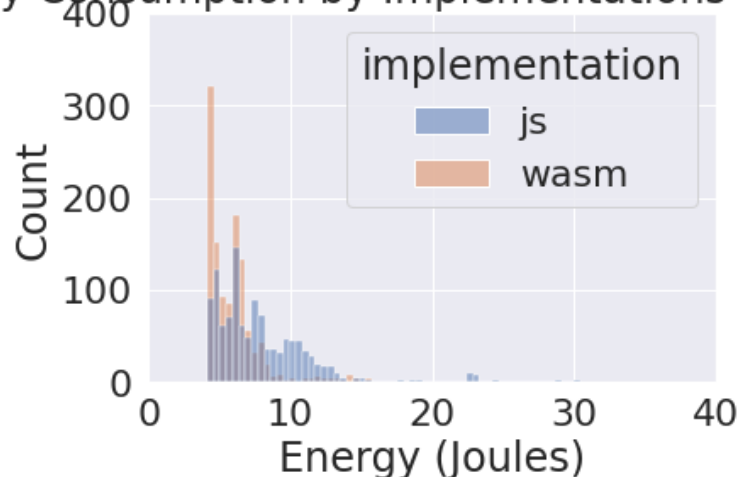
```

```

is_categorical_dtype is deprecated and will be removed in a future version. Use
isininstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isininstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isininstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1119: FutureWarning:
use_inf_as_na option is deprecated and will be removed in a future version.
Convert inf values to NaN before operating instead.
    with pd.option_context('mode.use_inf_as_na', True):

```

Energy Consumption by Implementations - SM-G991B



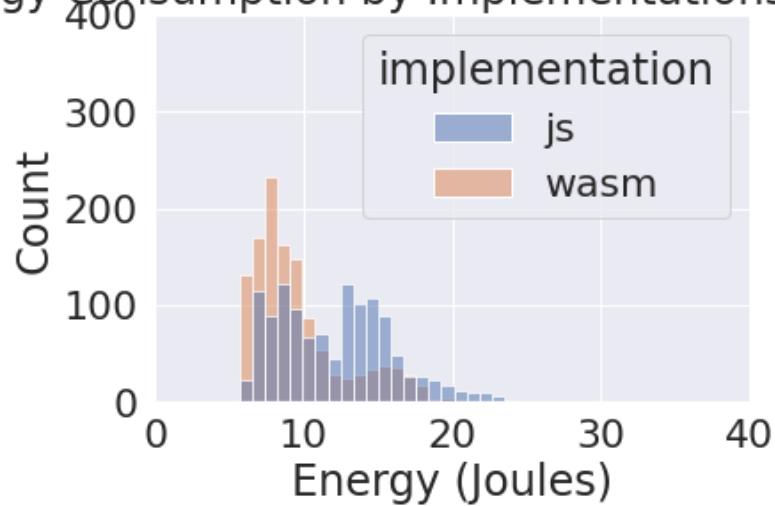
```

/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isininstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isininstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isininstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1119: FutureWarning:
use_inf_as_na option is deprecated and will be removed in a future version.

```

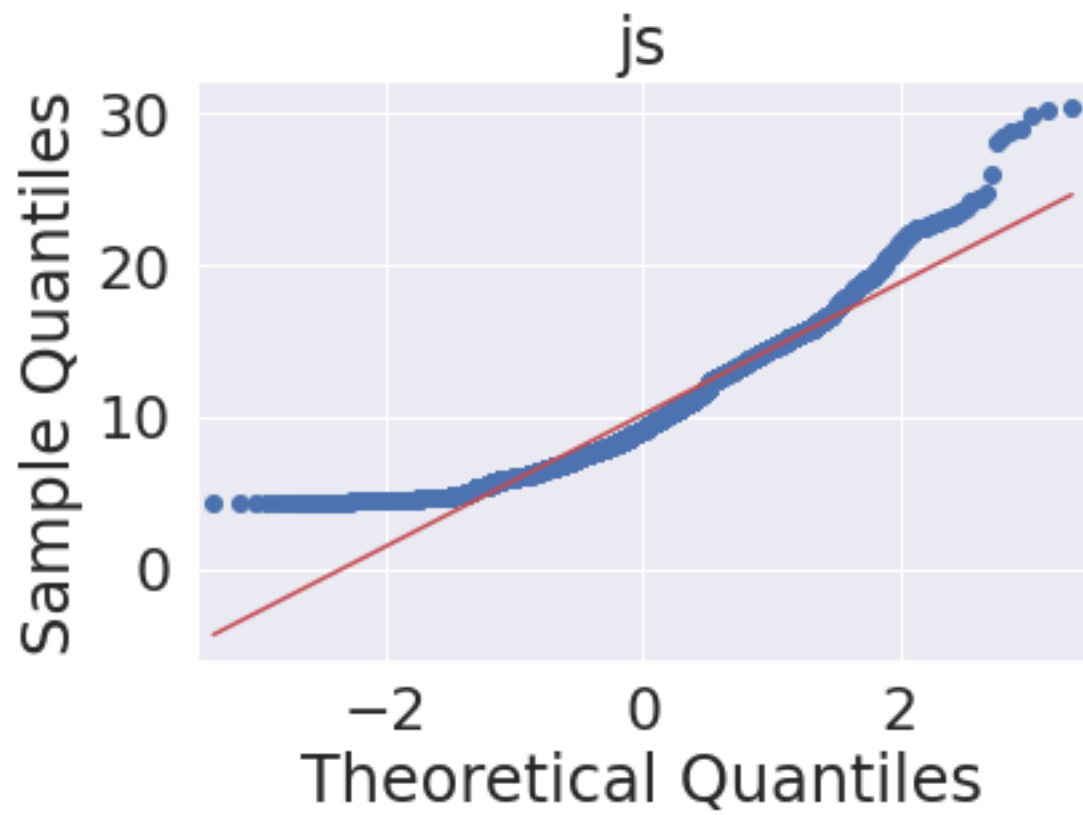
```
Convert inf values to NaN before operating instead.  
with pd.option_context('mode.use_inf_as_na', True):
```

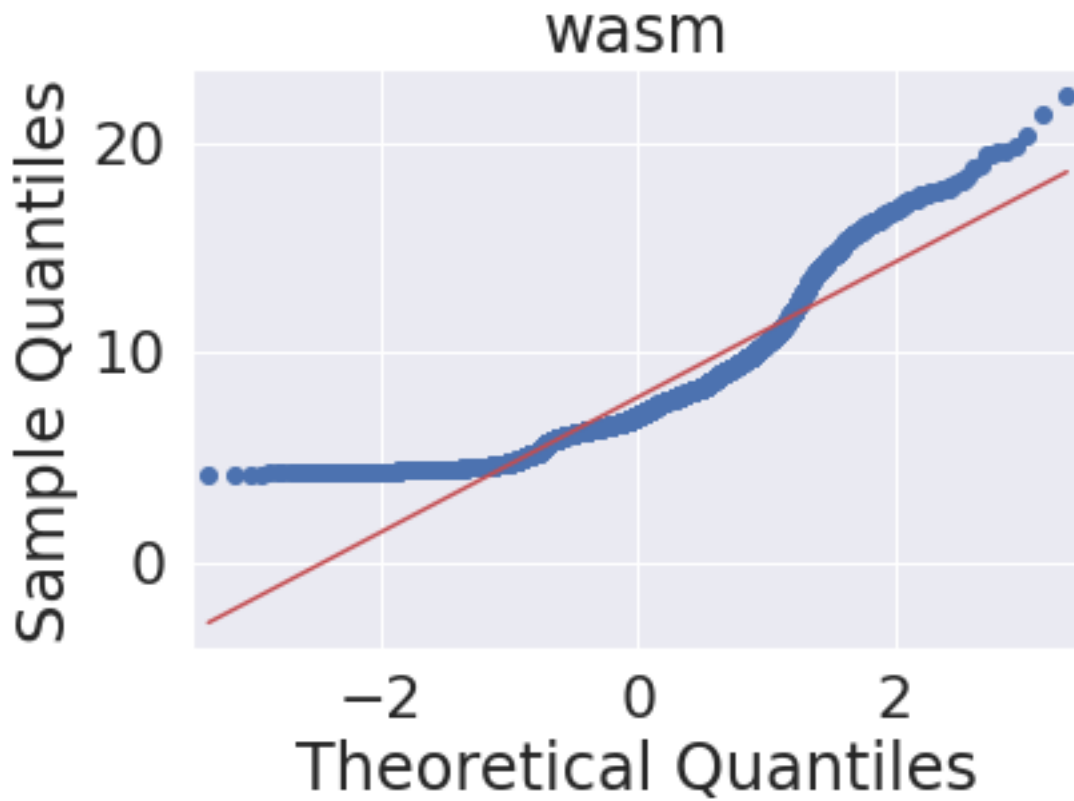
Energy Consumption by Implementations - Nexus 5



4.12 Q-Q-Plot

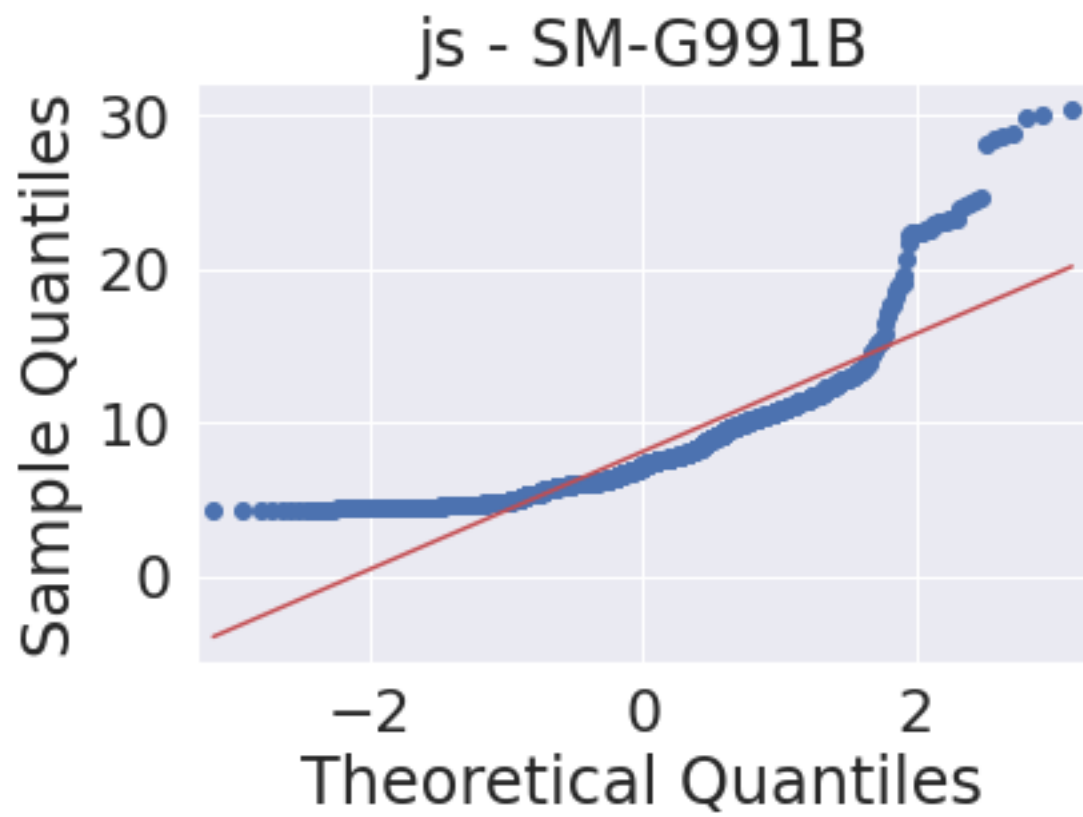
```
[20]: data = []  
for implementation in implementations:  
    data = df[(df['implementation'] == implementation)]  
    qq = sm.qqplot(data.energy, line='s')  
    h = plt.title(implementation)
```

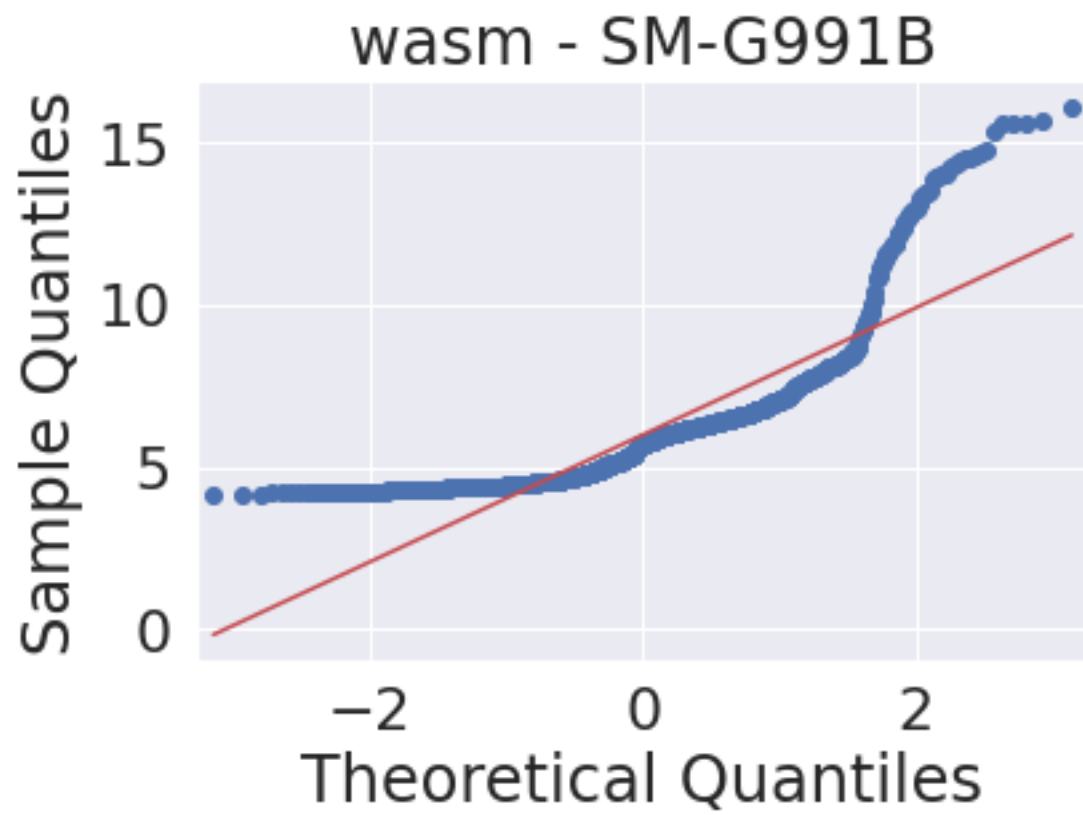


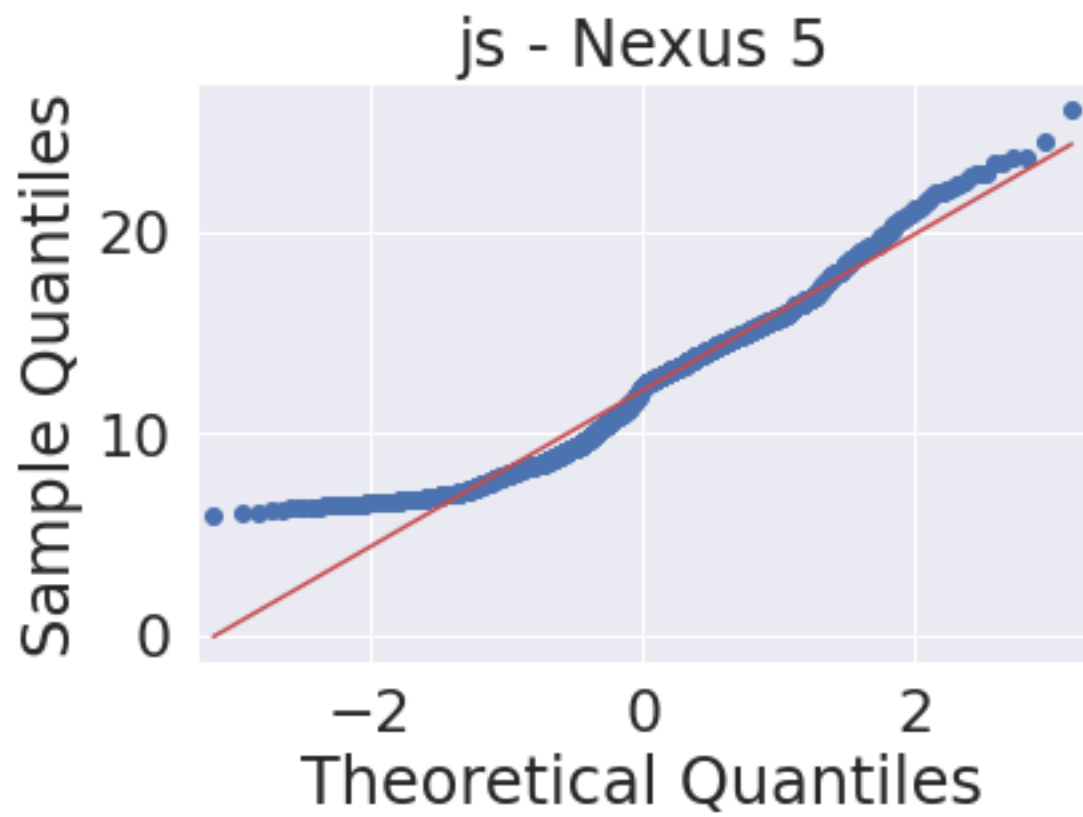


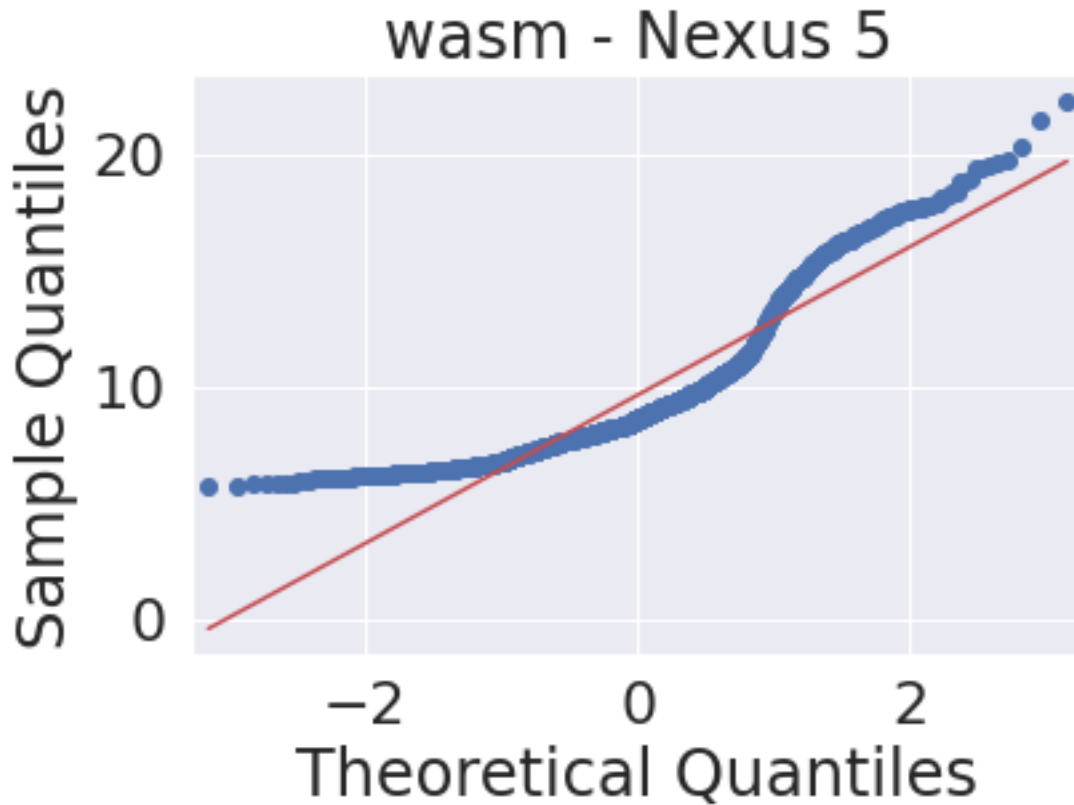
4.13 Q-Q-plot (By Device)

```
[21]: data = []
      for device in devices:
          for implementation in implementations:
              data = df[(df['implementation'] == implementation) & (df['device'] ==
↳ device)]
              qq = sm.qqplot(data.energy, line='s')
              h = plt.title(implementation + " - " + device)
```









4.14 Mann Whitney U Test (same Browsers)

```
[22]: data = []

for browser in browsers:
    for implementationpair in implementationpairs:
        impl1_energy = df[(df['implementation'] == implementationpair[0]) &
        ↪(df['browser'] == browser)]['energy']
        impl2_energy = df[(df['implementation'] == implementationpair[1]) &
        ↪(df['browser'] == browser)]['energy']
        eff = cliff.cliffs_delta(impl1_energy, impl2_energy)

        u = stats.mannwhitneyu(impl1_energy, impl2_energy,
        ↪alternative='two-sided')

        data.append(
            [
                browser,
                implementationpair[0] + ' vs. ' + implementationpair[1],
                u.statistic,
```

```

        u.pvalue,
        eff[0],
        eff[1]
    ]
)

# Create the pandas DataFrame
ut = pd.DataFrame(data, columns = ['browser', 'implementation', 'u', 'p', 'u',
    ↪ 'eff', 'interp'])
display(ut)

interp = ut['interp'].value_counts()
interp = pd.DataFrame(interp, columns = ['interp', 'percent'])
interp['percent'] = (interp['interp'] / interp['interp'].sum()) * 100

display(interp)

```

	browser	implementation	u	p	eff	interp
0	chrome	js vs. wasm	960697.0	1.168028e-41	0.317780	small
1	firefox	js vs. wasm	998597.0	5.082446e-57	0.374314	medium

Empty DataFrame
Columns: [interp, percent]
Index: []

4.15 Mann Whitney U Test (same Browsers - by Device)

```

[23]: data = []

for device in devices:
    for browser in browsers:
        for implementationpair in implementationpairs:
            impl1_energy = df[(df['implementation'] == implementationpair[0]) &
    ↪ (df['browser'] == browser) & (df['device'] == device)]['energy']
            impl2_energy = df[(df['implementation'] == implementationpair[1]) &
    ↪ (df['browser'] == browser) & (df['device'] == device)]['energy']
            eff = cliff.cliffs_delta(impl1_energy, impl2_energy)

            u = stats.mannwhitneyu(impl1_energy, impl2_energy,
    ↪ alternative='two-sided')

            data.append(
                [
                    device,
                    browser,
                    implementationpair[0] + ' vs. ' + implementationpair[1],
                    u.statistic,

```

```

        u.pvalue,
        eff[0],
        eff[1]
    ]
)

# Create the pandas DataFrame
ut = pd.DataFrame(data, columns = ['device', 'browser', 'implementation', 'u', 'p', 'eff', 'interp'])
display(ut)

interp = ut['interp'].value_counts()
interp = pd.DataFrame(interp, columns = ['interp', 'percent'])
interp['percent'] = (interp['interp'] / interp['interp'].sum()) * 100

display(interp)

```

	device	browser	implementation	u	p	eff	interp
0	SM-G991B	chrome	js vs. wasm	250266.0	1.591990e-36	0.423401	medium
1	SM-G991B	firefox	js vs. wasm	256565.0	1.160505e-42	0.459227	medium
2	Nexus 5	chrome	js vs. wasm	263410.0	3.872448e-33	0.395187	medium
3	Nexus 5	firefox	js vs. wasm	261860.0	3.617754e-33	0.396013	medium

```

Empty DataFrame
Columns: [interp, percent]
Index: []

```

4.16 Mann Whitney U Test (Cross Browsers)

```

[24]: data = []

for pairswitch in [[0,1],[1,0]]:
    for implementationpair in implementationpairs:
        for browserpair in browserpairs:
            browser1_energy = df[(df['browser'] == browserpair[0]) &
            ↪ (df['implementation'] == implementationpair[pairswitch[0]])]['energy']
            browser2_energy = df[(df['browser'] == browserpair[1]) &
            ↪ (df['implementation'] == implementationpair[pairswitch[1]])]['energy']
            eff = cliff.cliffs_delta(browser1_energy, browser2_energy)

            u = stats.mannwhitneyu(browser1_energy, browser2_energy,
            ↪ alternative='two-sided')

            data.append(
                [
                    browserpair[0] + ' vs. ' + browserpair[1],

```

```

        implementationpair[pairswitch[0]] + ' vs. ' +
↪implementationpair[pairswitch[1]],
        u.statistic,
        u.pvalue,
        eff[0],
        eff[1]
    ]
)

# Create the pandas DataFrame
ut = pd.DataFrame(data, columns = ['browser', 'implementation', 'u', 'p',
↪'eff', 'interp'])
display(ut)

interp = ut['interp'].value_counts()
interp = pd.DataFrame(interp, columns = ['interp', 'percent'])
interp['percent'] = (interp['interp'] / interp['interp'].sum()) * 100

display(interp)

```

	browser	implementation	u	p	eff	interp	
0	chrome	vs. firefox	js vs. wasm	880131.0	3.841190e-19	0.210271	small
1	chrome	vs. firefox	wasm vs. js	391450.0	3.216940e-86	-0.462606	medium

Empty DataFrame

Columns: [interp, percent]

Index: []

4.17 Mann Whitney U Test (Cross Browsers - By Device)

```

[25]: data = []

for device in devices:
    for pairswitch in [[0,1],[1,0]]:
        for implementationpair in implementationpairs:
            for browserpair in browserpairs:
                browser1_energy = df[(df['browser'] == browserpair[0]) &
↪(df['implementation'] == implementationpair[pairswitch[0]]) & (df['device']
↪== device)]['energy']
                browser2_energy = df[(df['browser'] == browserpair[1]) &
↪(df['implementation'] == implementationpair[pairswitch[1]]) & (df['device']
↪== device)]['energy']
                eff = cliff.cliffs_delta(browser1_energy, browser2_energy)

                u = stats.mannwhitneyu(browser1_energy, browser2_energy,
↪alternative='two-sided')

```

```

        data.append(
            [
                device,
                browserpair[0] + ' vs. ' + browserpair[1],
                implementationpair[pairswitch[0]] + ' vs. ' +
↪implementationpair[pairswitch[1]],
                u.statistic,
                u.pvalue,
                eff[0],
                eff[1]
            ]
        )

# Create the pandas DataFrame
ut = pd.DataFrame(data, columns = ['device', 'browser', 'implementation', 'u',
↪'p', 'eff', 'interp'])
display(ut)

interp = ut['interp'].value_counts()
interp = pd.DataFrame(interp, columns = ['interp', 'percent'])
interp['percent'] = (interp['interp'] / interp['interp'].sum()) * 100

display(interp)

```

	device	browser	implementation	u	p \
0	SM-G991B	chrome	vs. firefox	js vs. wasm	227749.0 1.136373e-17
1	SM-G991B	chrome	vs. firefox	wasm vs. js	74651.0 4.165707e-65
2	Nexus 5	chrome	vs. firefox	js vs. wasm	239728.0 7.751432e-18
3	Nexus 5	chrome	vs. firefox	wasm vs. js	97397.0 1.789286e-49

	eff	interp
0	0.286627	small
1	-0.572545	large
2	0.284300	small
3	-0.486644	large

```

Empty DataFrame
Columns: [interp, percent]
Index: []

```

4.18 Descriptive Statistics

```

[26]: data = []
for implementation in implementations:
    x = df[(df['implementation'] == implementation)]
    mean = numpy.round(numpy.mean(x['energy']), 2)
    median = numpy.round(numpy.median(x['energy']), 2)

```

```

min = numpy.round(numpy.amin(x['energy']), 2)
max = numpy.round(numpy.amax(x['energy']), 2)
std = numpy.round(numpy.std(x['energy']), 2)
sem = numpy.round(stats.sem(x['energy']), 2)
q1 = numpy.round(numpy.quantile(x['energy'], 0.25), 2)
q3 = numpy.round(numpy.quantile(x['energy'], 0.75), 2)

data.append(
    [implementation, mean, std, min, q1, median, q3, max, sem]
)

# Create the pandas DataFrame
stat = pd.DataFrame(data, columns = ['implementation', 'mean', 'std', 'min', 'q1', 'median', 'q3', 'max', 'sem'])
# display(stat)
print(stat.to_string())

# Alternative of pandas: x['energy'].describe()

```

	implementation	mean	std	min	q1	median	q3	max	sem
0	js	10.16	4.34	4.30	6.77	9.19	12.96	30.46	0.09
1	wasm	7.84	3.22	4.16	5.75	6.97	9.04	22.29	0.07

4.19 Descriptive Statistics Difference

```

[27]: data = []

for implementationpair in implementationpairs:
    implementation1 = stat[(stat['implementation'] == implementationpair[1])]
    implementation2 = stat[(stat['implementation'] == implementationpair[0])]

    mean_diff = implementation1.iloc[0]['mean']-implementation2.iloc[0]['mean']
    median_diff = implementation1.iloc[0]['median']-implementation2.
    ↪iloc[0]['median']
    min_diff = implementation1.iloc[0]['min']-implementation2.iloc[0]['min']
    max_diff = implementation1.iloc[0]['max']-implementation2.iloc[0]['max']
    std_diff = implementation1.iloc[0]['std']-implementation2.iloc[0]['std']
    sem_diff = implementation1.iloc[0]['sem']-implementation2.iloc[0]['sem']
    q1_diff = implementation1.iloc[0]['q1']-implementation2.iloc[0]['q1']
    q3_diff = implementation1.iloc[0]['q3']-implementation2.iloc[0]['q3']

    data.append(
        [implementationpair[1] + ' vs. ' + implementationpair[0],
         numpy.round(mean_diff, 2),
         numpy.round(mean_diff/implementation2.iloc[0]['mean']*100, 2),
         numpy.round(median_diff, 2),
         numpy.round(median_diff/implementation2.iloc[0]['median']*100, 2),

```



```

        numpy.round(min_diff, 2),
        numpy.round(min_diff/implementation2.iloc[0]['min']*100, 2),
        numpy.round(max_diff, 2),
        numpy.round(max_diff/implementation2.iloc[0]['max']*100, 2),
        numpy.round(std_diff, 2),
        numpy.round(std_diff/implementation2.iloc[0]['std']*100, 2),
        numpy.round(sem_diff, 2),
        numpy.round(sem_diff/implementation2.iloc[0]['sem']*100, 2),
        numpy.round(q1_diff, 2),
        numpy.round(q1_diff/implementation2.iloc[0]['q1']*100, 2),
        numpy.round(q3_diff, 2),
        numpy.round(q3_diff/implementation2.iloc[0]['q3']*100, 2),
    ]
)

# Create the pandas DataFrame
ut = pd.DataFrame(data, columns = [
    'rq', 'mean_diff', 'mean_diff%', 'median_diff', 'median_diff%', 'min_diff', 'min_diff%', 'max_diff',
    'max_diff%', 'std_diff', 'std_diff%', 'sem_diff', 'sem_diff%', 'q1_diff', 'q1_diff%', 'q3_diff', 'q3_diff%'
])
display(ut)
#print(ut.to_string())

```

	rq	mean_diff	mean_diff%	median_diff	median_diff%	min_diff	\
0	wasm vs. js	-2.32	-22.83	-2.22	-24.16	-0.14	

	min_diff%	max_diff	max_diff%	std_diff	std_diff%	sem_diff	sem_diff%	\
0	-3.26	-8.17	-26.82	-1.12	-25.81	-0.02	-22.22	

	q1_diff	q1_diff%	q3_diff	q3_diff%
0	-1.02	-15.07	-3.92	-30.25

4.20 Descriptive Statistics (By Browser)

```

[28]: data = []
for implementation in implementations:
    for browser in browsers:
        x = df[(df['implementation'] == implementation) & (df['browser'] ==
        browser)]
        mean = numpy.round(numpy.mean(x['energy']), 2)
        median = numpy.round(numpy.median(x['energy']), 2)
        min = numpy.round(numpy.amin(x['energy']), 2)
        max = numpy.round(numpy.amax(x['energy']), 2)
        std = numpy.round(numpy.std(x['energy']), 2)
        sem = numpy.round(stats.sem(x['energy']), 2)
        q1 = numpy.round(numpy.quantile(x['energy'], 0.25), 2)
        q3 = numpy.round(numpy.quantile(x['energy'], 0.75), 2)

        data.append(

```

```

        (implementation, browser, mean, std, min, q1, median, q3, max, sem]

# Create the pandas DataFrame
stat = pd.DataFrame(data, columns = ['implementation', 'browser', 'mean',
    ↪ 'std', 'min', 'q1', 'median', 'q3', 'max', 'sem'])
# display(stat)
print(stat.to_string())

# Alternative of pandas: x['energy'].describe()

```

	implementation	browser	mean	std	min	q1	median	q3	max	sem
0	js	chrome	9.37	3.79	4.30	6.52	8.47	11.28	26.04	0.11
1	js	firefox	10.94	4.70	4.39	7.11	10.43	13.89	30.46	0.14
2	wasm	chrome	7.55	3.04	4.16	5.64	6.85	8.37	22.29	0.09
3	wasm	firefox	8.14	3.36	4.22	5.96	7.09	9.67	19.78	0.10

4.21 Descriptive Statistics Difference (By Browser)

```

[29]: data = []

for implementationpair in implementationpairs:
    for browser in browsers:
        implementation1 = stat[(stat['implementation'] ==
    ↪ implementationpair[1]) & (stat['browser'] == browser)]
        implementation2 = stat[(stat['implementation'] ==
    ↪ implementationpair[0]) & (stat['browser'] == browser)]

        mean_diff = implementation1.iloc[0]['mean']-implementation2.
    ↪ iloc[0]['mean']
        median_diff = implementation1.iloc[0]['median']-implementation2.
    ↪ iloc[0]['median']
        min_diff = implementation1.iloc[0]['min']-implementation2.iloc[0]['min']
        max_diff = implementation1.iloc[0]['max']-implementation2.iloc[0]['max']
        std_diff = implementation1.iloc[0]['std']-implementation2.iloc[0]['std']
        sem_diff = implementation1.iloc[0]['sem']-implementation2.iloc[0]['sem']
        q1_diff = implementation1.iloc[0]['q1']-implementation2.iloc[0]['q1']
        q3_diff = implementation1.iloc[0]['q3']-implementation2.iloc[0]['q3']

        data.append(
            [implementationpair[1] + ' vs. ' + implementationpair[0] + ' ' +
    ↪ browser,
            numpy.round(mean_diff, 2),
            numpy.round(mean_diff/implementation2.iloc[0]['mean']*100, 2),
            numpy.round(median_diff, 2),
            numpy.round(median_diff/implementation2.iloc[0]['median']*100, 2),
            numpy.round(min_diff, 2),

```

```

        numpy.round(min_diff/implementation2.iloc[0]['min']*100, 2),
        numpy.round(max_diff, 2),
        numpy.round(max_diff/implementation2.iloc[0]['max']*100, 2),
        numpy.round(std_diff, 2),
        numpy.round(std_diff/implementation2.iloc[0]['std']*100, 2),
        numpy.round(sem_diff, 2),
        numpy.round(sem_diff/implementation2.iloc[0]['sem']*100, 2),
        numpy.round(q1_diff, 2),
        numpy.round(q1_diff/implementation2.iloc[0]['q1']*100, 2),
        numpy.round(q3_diff, 2),
        numpy.round(q3_diff/implementation2.iloc[0]['q3']*100, 2),
    ]
)

# Create the pandas DataFrame
ut = pd.DataFrame(data, columns = [
    'rq', 'mean_diff', 'mean_diff%', 'median_diff', 'median_diff%', 'min_diff', 'min_diff%', 'max_diff',
    'max_diff%', 'std_diff', 'std_diff%', 'sem_diff', 'sem_diff%', 'q1_diff', 'q1_diff%', 'q3_diff', 'q3_diff%'
])
display(ut)
#print(ut.to_string())

```

		rq	mean_diff	mean_diff%	median_diff	median_diff%	\
0	wasm vs. js chrome		-1.82	-19.42	-1.62	-19.13	
1	wasm vs. js firefox		-2.80	-25.59	-3.34	-32.02	

	min_diff	min_diff%	max_diff	max_diff%	std_diff	std_diff%	sem_diff	\
0	-0.14	-3.26	-3.75	-14.40	-0.75	-19.79	-0.02	
1	-0.17	-3.87	-10.68	-35.06	-1.34	-28.51	-0.04	

	sem_diff%	q1_diff	q1_diff%	q3_diff	q3_diff%
0	-18.18	-0.88	-13.50	-2.91	-25.80
1	-28.57	-1.15	-16.17	-4.22	-30.38

4.22 Descriptive Statistics Difference (Cross Browser)

```

[30]: data = []

for pairswitch in [[0,1],[1,0]]:
    for implementationpair in implementationpairs:
        for browserpair in browserpairs:
            implementation1 = stat[(stat['browser'] ==
            ↪browserpair[pairswitch[1]]) & (stat['implementation'] ==
            ↪implementationpair[1])]
            implementation2 = stat[(stat['browser'] ==
            ↪browserpair[pairswitch[0]]) & (stat['implementation'] ==
            ↪implementationpair[0])]

#for implementationpair in implementationpairs:

```

```

#     for browser in browsers:
#         implementation1 = stat[(stat['implementation'] ==
↳ implementationpair[0]) & (stat['browser'] == browser)]
#         implementation2 = stat[(stat['implementation'] ==
↳ implementationpair[1]) & (stat['browser'] == browser)]

        mean_diff = implementation1.iloc[0]['mean']-implementation2.
↳iloc[0]['mean']
        median_diff = implementation1.iloc[0]['median']-implementation2.
↳iloc[0]['median']
        min_diff = implementation1.iloc[0]['min']-implementation2.iloc[0]['min']
        max_diff = implementation1.iloc[0]['max']-implementation2.iloc[0]['max']
        std_diff = implementation1.iloc[0]['std']-implementation2.iloc[0]['std']
        sem_diff = implementation1.iloc[0]['sem']-implementation2.iloc[0]['sem']
        q1_diff = implementation1.iloc[0]['q1']-implementation2.iloc[0]['q1']
        q3_diff = implementation1.iloc[0]['q3']-implementation2.iloc[0]['q3']

        data.append(
            [
                browserpair[pairswitch[1]] + ' ' + implementationpair[1] + ' vs. '
↳ + browserpair[pairswitch[0]] + ' ' + implementationpair[0],
                numpy.round(mean_diff, 2),
                numpy.round(mean_diff/implementation2.iloc[0]['mean']*100, 2),
                numpy.round(median_diff, 2),
                numpy.round(median_diff/implementation2.iloc[0]['median']*100, 2),
                numpy.round(min_diff, 2),
                numpy.round(min_diff/implementation2.iloc[0]['min']*100, 2),
                numpy.round(max_diff, 2),
                numpy.round(max_diff/implementation2.iloc[0]['max']*100, 2),
                numpy.round(std_diff, 2),
                numpy.round(std_diff/implementation2.iloc[0]['std']*100, 2),
                numpy.round(sem_diff, 2),
                numpy.round(sem_diff/implementation2.iloc[0]['sem']*100, 2),
                numpy.round(q1_diff, 2),
                numpy.round(q1_diff/implementation2.iloc[0]['q1']*100, 2),
                numpy.round(q3_diff, 2),
                numpy.round(q3_diff/implementation2.iloc[0]['q3']*100, 2),
            ]
        )

# Create the pandas DataFrame
ut = pd.DataFrame(data, columns =
↳ ['rq', 'mean_diff', 'mean_diff%', 'median_diff', 'median_diff%', 'min_diff', 'min_diff%', 'max_diff', 'max_diff%'])
display(ut)
#print(ut.to_string())

```

```
rq mean_diff mean_diff% median_diff \
```

0	firefox wasm vs. chrome js	-1.23	-13.13	-1.38
1	chrome wasm vs. firefox js	-3.39	-30.99	-3.58

	median_diff%	min_diff	min_diff%	max_diff	max_diff%	std_diff	\
0	-16.29	-0.08	-1.86	-6.26	-24.04	-0.43	
1	-34.32	-0.23	-5.24	-8.17	-26.82	-1.66	

	std_diff%	sem_diff	sem_diff%	q1_diff	q1_diff%	q3_diff	q3_diff%
0	-11.35	-0.01	-9.09	-0.56	-8.59	-1.61	-14.27
1	-35.32	-0.05	-35.71	-1.47	-20.68	-5.52	-39.74

4.23 Descriptive Statistics (By Browser & By Device)

```
[31]: data = []
for device in devices:
    for implementation in implementations:
        for browser in browsers:
            x = df[(df['implementation'] == implementation) & (df['browser'] ==
↪ browser) & (df['device'] == device)]
            mean = numpy.round(numpy.mean(x['energy']), 2)
            median = numpy.round(numpy.median(x['energy']), 2)
            min = numpy.round(numpy.amin(x['energy']), 2)
            max = numpy.round(numpy.amax(x['energy']), 2)
            std = numpy.round(numpy.std(x['energy']), 2)
            sem = numpy.round(stats.sem(x['energy']), 2)
            q1 = numpy.round(numpy.quantile(x['energy'], 0.25), 2)
            q3 = numpy.round(numpy.quantile(x['energy'], 0.75), 2)

            data.append(
                [implementation, device, browser, mean, std, min, q1, median,
↪ q3, max, sem]
            )

# Create the pandas DataFrame
stat = pd.DataFrame(data, columns = ['implementation', 'device', 'browser',
↪ 'mean', 'std', 'min', 'q1', 'median', 'q3', 'max', 'sem'])
# display(stat)
print(stat.to_string())

# Alternative of pandas: x['energy'].describe()
```

	implementation	device	browser	mean	std	min	q1	median	q3
max									
0	js	SM-G991B	chrome	7.05	2.00	4.30	5.46	6.62	8.29
				13.96	0.08				
1	js	SM-G991B	firefox	9.21	4.82	4.39	5.99	7.67	11.03
				30.46	0.20				

2		wasm	SM-G991B	chrome	5.67	1.25	4.16	4.49	5.47	6.46
10.14	0.05									
3		wasm	SM-G991B	firefox	6.32	2.43	4.22	4.63	5.96	6.62
16.11	0.10									
4		js	Nexus 5	chrome	11.64	3.75	5.92	8.53	10.93	14.44
26.04	0.15									
5		js	Nexus 5	firefox	12.61	3.91	6.06	9.03	13.13	15.01
24.44	0.16									
6		wasm	Nexus 5	chrome	9.34	3.16	5.67	7.23	8.24	10.02
22.29	0.13									
7		wasm	Nexus 5	firefox	9.91	3.19	5.79	7.63	9.04	10.99
19.78	0.13									

4.24 Descriptive Statistics Difference (By Browser & By Device)

```
[32]: data = []

for implementationpair in implementationpairs:
    for device in devices:
        for browser in browsers:
            implementation1 = stat[(stat['implementation'] ==_
↳implementationpair[1]) & (stat['browser'] == browser) & (stat['device'] ==_
↳device)]
            implementation2 = stat[(stat['implementation'] ==_
↳implementationpair[0]) & (stat['browser'] == browser) & (stat['device'] ==_
↳device)]

            mean_diff = implementation1.iloc[0]['mean']-implementation2.
↳iloc[0]['mean']
            median_diff = implementation1.iloc[0]['median']-implementation2.
↳iloc[0]['median']
            min_diff = implementation1.iloc[0]['min']-implementation2.
↳iloc[0]['min']
            max_diff = implementation1.iloc[0]['max']-implementation2.
↳iloc[0]['max']
            std_diff = implementation1.iloc[0]['std']-implementation2.
↳iloc[0]['std']
            sem_diff = implementation1.iloc[0]['sem']-implementation2.
↳iloc[0]['sem']
            q1_diff = implementation1.iloc[0]['q1']-implementation2.
↳iloc[0]['q1']
            q3_diff = implementation1.iloc[0]['q3']-implementation2.
↳iloc[0]['q3']

            data.append(
                [implementationpair[1] + ' vs. ' + implementationpair[0] + ' '_
↳+ browser,
```

```

        device,
        numpy.round(mean_diff, 2),
        numpy.round(mean_diff/implementation2.iloc[0]['mean']*100, 2),
        numpy.round(median_diff, 2),
        numpy.round(median_diff/implementation2.iloc[0]['median']*100, 2),
    ↪2),

    numpy.round(min_diff, 2),
    numpy.round(min_diff/implementation2.iloc[0]['min']*100, 2),
    numpy.round(max_diff, 2),
    numpy.round(max_diff/implementation2.iloc[0]['max']*100, 2),
    numpy.round(std_diff, 2),
    numpy.round(std_diff/implementation2.iloc[0]['std']*100, 2),
    numpy.round(sem_diff, 2),
    numpy.round(sem_diff/implementation2.iloc[0]['sem']*100, 2),
    numpy.round(q1_diff, 2),
    numpy.round(q1_diff/implementation2.iloc[0]['q1']*100, 2),
    numpy.round(q3_diff, 2),
    numpy.round(q3_diff/implementation2.iloc[0]['q3']*100, 2),
    ]
    )

# Create the pandas DataFrame
ut = pd.DataFrame(data, columns = [
    ↪['rq', 'device', 'mean_diff', 'mean_diff%', 'median_diff', 'median_diff%', 'min_diff', 'min_diff%',
display(ut)
#print(ut.to_string())

```

	rq	device	mean_diff	mean_diff%	median_diff	\
0	wasm vs. js chrome	SM-G991B	-1.38	-19.57	-1.15	
1	wasm vs. js firefox	SM-G991B	-2.89	-31.38	-1.71	
2	wasm vs. js chrome	Nexus 5	-2.30	-19.76	-2.69	
3	wasm vs. js firefox	Nexus 5	-2.70	-21.41	-4.09	

	median_diff%	min_diff	min_diff%	max_diff	max_diff%	std_diff	\
0	-17.37	-0.14	-3.26	-3.82	-27.36	-0.75	
1	-22.29	-0.17	-3.87	-14.35	-47.11	-2.39	
2	-24.61	-0.25	-4.22	-3.75	-14.40	-0.59	
3	-31.15	-0.27	-4.46	-4.66	-19.07	-0.72	

	std_diff%	sem_diff	sem_diff%	q1_diff	q1_diff%	q3_diff	q3_diff%
0	-37.50	-0.03	-37.50	-0.97	-17.77	-1.83	-22.07
1	-49.59	-0.10	-50.00	-1.36	-22.70	-4.41	-39.98
2	-15.73	-0.02	-13.33	-1.30	-15.24	-4.42	-30.61
3	-18.41	-0.03	-18.75	-1.40	-15.50	-4.02	-26.78

4.25 Descriptive Statistics Difference (Cross Browser & By Device)

```
[33]: data = []

for pairswitch in [[0,1],[1,0]]:
    for device in devices:
        for implementationpair in implementationpairs:
            for browserpair in browserpairs:
                implementation1 = stat[(stat['browser'] ==_
↪browserpair[pairswitch[1]]) & (stat['implementation'] ==_
↪implementationpair[1]) & (stat['device'] == device)]
                implementation2 = stat[(stat['browser'] ==_
↪browserpair[pairswitch[0]]) & (stat['implementation'] ==_
↪implementationpair[0]) & (stat['device'] == device)]

                #for implementationpair in implementationpairs:
                #    for browser in browsers:
                #        implementation1 = stat[(stat['implementation'] ==_
↪implementationpair[0]) & (stat['browser'] == browser)]
                #        implementation2 = stat[(stat['implementation'] ==_
↪implementationpair[1]) & (stat['browser'] == browser)]

                mean_diff = implementation1.iloc[0]['mean']-implementation2.
↪iloc[0]['mean']
                median_diff = implementation1.iloc[0]['median']-implementation2.
↪iloc[0]['median']
                min_diff = implementation1.iloc[0]['min']-implementation2.
↪iloc[0]['min']
                max_diff = implementation1.iloc[0]['max']-implementation2.
↪iloc[0]['max']
                std_diff = implementation1.iloc[0]['std']-implementation2.
↪iloc[0]['std']
                sem_diff = implementation1.iloc[0]['sem']-implementation2.
↪iloc[0]['sem']
                q1_diff = implementation1.iloc[0]['q1']-implementation2.
↪iloc[0]['q1']
                q3_diff = implementation1.iloc[0]['q3']-implementation2.
↪iloc[0]['q3']

                data.append(
                    [
                        browserpair[pairswitch[1]] + ' ' + implementationpair[1] + '_
↪vs. ' + browserpair[pairswitch[0]] + ' ' + implementationpair[0],
                        device,
                        numpy.round(mean_diff, 2),
                        numpy.round(mean_diff/implementation2.iloc[0]['mean']*100, 2),
```



```

        numpy.round(median_diff, 2),
        numpy.round(median_diff/implementation2.iloc[0]['median']*100, 2),
    ↪2),

        numpy.round(min_diff, 2),
        numpy.round(min_diff/implementation2.iloc[0]['min']*100, 2),
        numpy.round(max_diff, 2),
        numpy.round(max_diff/implementation2.iloc[0]['max']*100, 2),
        numpy.round(std_diff, 2),
        numpy.round(std_diff/implementation2.iloc[0]['std']*100, 2),
        numpy.round(sem_diff, 2),
        numpy.round(sem_diff/implementation2.iloc[0]['sem']*100, 2),
        numpy.round(q1_diff, 2),
        numpy.round(q1_diff/implementation2.iloc[0]['q1']*100, 2),
        numpy.round(q3_diff, 2),
        numpy.round(q3_diff/implementation2.iloc[0]['q3']*100, 2),
    ]
)

# Create the pandas DataFrame
ut = pd.DataFrame(data, columns = [
    ↪['rq', 'device', 'mean_diff', 'mean_diff%', 'median_diff', 'median_diff%', 'min_diff', 'min_diff%',
display(ut)
#print(ut.to_string())

```

		rq	device	mean_diff	mean_diff%	median_diff	\
0	firefox wasm vs. chrome	js	SM-G991B	-0.73	-10.35	-0.66	
1	firefox wasm vs. chrome	js	Nexus 5	-1.73	-14.86	-1.89	
2	chrome wasm vs. firefox	js	SM-G991B	-3.54	-38.44	-2.20	
3	chrome wasm vs. firefox	js	Nexus 5	-3.27	-25.93	-4.89	

	median_diff%	min_diff	min_diff%	max_diff	max_diff%	std_diff	\
0	-9.97	-0.08	-1.86	2.15	15.40	0.43	
1	-17.29	-0.13	-2.20	-6.26	-24.04	-0.56	
2	-28.68	-0.23	-5.24	-20.32	-66.71	-3.57	
3	-37.24	-0.39	-6.44	-2.15	-8.80	-0.75	

	std_diff%	sem_diff	sem_diff%	q1_diff	q1_diff%	q3_diff	q3_diff%
0	21.50	0.02	25.00	-0.83	-15.20	-1.67	-20.14
1	-14.93	-0.02	-13.33	-0.90	-10.55	-3.45	-23.89
2	-74.07	-0.15	-75.00	-1.50	-25.04	-4.57	-41.43
3	-19.18	-0.03	-18.75	-1.80	-19.93	-4.99	-33.24

5 RQ2: JS Energy Browser

5.1 Shapiro Wilk Test

```
[34]: data = []
non_normal = 0

for browser in browsers:
    energy = df[(df['browser'] == browser) & (df['implementation'] == 'js')]['energy']

    if len(energy) >= 3:
        shapiro_test = stats.shapiro(energy)

        non_normal += (1 if shapiro_test.pvalue <= 0.05 else 0)

        data.append(
            [browser, 'js',
             shapiro_test.statistic,
             shapiro_test.pvalue
            ]
        )

# Create the pandas DataFrame
swt = pd.DataFrame(data, columns = ['browser', 'implementation', 'w', 'p'])
#print(swt.to_string())
display(swt)

print("\n{n} non-normally distributed samples".format(non_normal))
print("{n} normally distributed samples".format(len(swt) - non_normal))
print("{:.2f}% non-normally distributed samples".format(non_normal/len(swt)*100))
```

	browser	implementation	w	p
0	chrome	js	0.923120	2.485691e-24
1	firefox	js	0.939191	8.412692e-22

```
2 non-normally distributed samples
0 normally distributed samples
100.00% non-normally distributed samples
```

5.2 Shapiro Wilk Test (By Device)

```
[35]: data = []
non_normal = 0

for device in devices:
```

```

for browser in browsers:
    energy = df[(df['browser'] == browser) & (df['implementation'] == 'js')]
    ↪ (df['device'] == device)]['energy']

    if len(energy) >= 3:
        shapiro_test = stats.shapiro(energy)

        non_normal += (1 if shapiro_test.pvalue <= 0.05 else 0)

        data.append(
            [
                device,
                browser, 'js',
                shapiro_test.statistic,
                shapiro_test.pvalue
            ]
        )

# Create the pandas DataFrame
swt = pd.DataFrame(data, columns = ['device', 'browser', 'implementation', 'w',
    ↪ 'p'])
#print(swt.to_string())
display(swt)

print("\n{} non-normally distributed samples".format(non_normal))
print("{} normally distributed samples".format(len(swt) - non_normal))
print("{:.2f}% non-normally distributed samples".format(non_normal/
    ↪ len(swt)*100))

```

	device	browser	implementation	w	p
0	SM-G991B	chrome	js	0.943418	2.788160e-14
1	SM-G991B	firefox	js	0.808114	6.720060e-26
2	Nexus 5	chrome	js	0.949549	1.371225e-13
3	Nexus 5	firefox	js	0.963830	3.718362e-11

```

4 non-normally distributed samples
0 normally distributed samples
100.00% non-normally distributed samples

```

5.3 Mann Whitney U Test

```

[36]: data = []

for browserpair in browserpairs:
    browser1_energy = df[(df['browser'] == browserpair[0]) &
    ↪ (df['implementation'] == 'js')]['energy']

```

```

    browser2_energy = df[(df['browser'] == browserpair[1]) &
↳(df['implementation'] == 'js')]['energy']
    eff = cliff.cliffs_delta(browser1_energy, browser2_energy)

    u = stats.mannwhitneyu(browser1_energy, browser2_energy,
↳alternative='two-sided')

    data.append(
        [
            browserpair[0] + ' vs. ' + browserpair[1],
            u.statistic,
            u.pvalue,
            eff[0],
            eff[1]
        ]
    )

# Create the pandas DataFrame
ut = pd.DataFrame(data, columns = ['rq', 'u', 'p', 'eff', 'interp'])
display(ut)

interp = ut['interp'].value_counts()
interp = pd.DataFrame(interp, columns = ['interp', 'percent'])
interp['percent'] = (interp['interp'] / interp['interp'].sum()) * 100

display(interp)

```

```

           rq           u           p      eff interp
0  chrome vs. firefox  586526.0  2.471443e-16 -0.192797  small

```

```

Empty DataFrame
Columns: [interp, percent]
Index: []

```

5.4 Mann Whitney U Test (By Device)

```

[37]: data = []

for device in devices:
    for browserpair in browserpairs:
        browser1_energy = df[(df['browser'] == browserpair[0]) &
↳(df['implementation'] == 'js') & (df['device'] == device)]['energy']
        browser2_energy = df[(df['browser'] == browserpair[1]) &
↳(df['implementation'] == 'js') & (df['device'] == device)]['energy']
        eff = cliff.cliffs_delta(browser1_energy, browser2_energy)

```

```

u = stats.mannwhitneyu(browser1_energy, browser2_energy,
↪alternative='two-sided')

data.append(
    [
        device,
        browserpair[0] + ' vs. ' + browserpair[1],
        u.statistic,
        u.pvalue,
        eff[0],
        eff[1]
    ]
)

# Create the pandas DataFrame
ut = pd.DataFrame(data, columns = ['device', 'rq', 'u', 'p', 'eff', 'interp'])
display(ut)

interp = ut['interp'].value_counts()
interp = pd.DataFrame(interp, columns = ['interp', 'percent'])
interp['percent'] = (interp['interp'] / interp['interp'].sum()) * 100

display(interp)

```

	device		rq	u	p	eff	interp
0	SM-G991B	chrome vs. firefox	130114.0	9.184663e-15	-0.259970		small
1	Nexus 5	chrome vs. firefox	160175.0	9.598055e-06	-0.146084		negligible

```

Empty DataFrame
Columns: [interp, percent]
Index: []

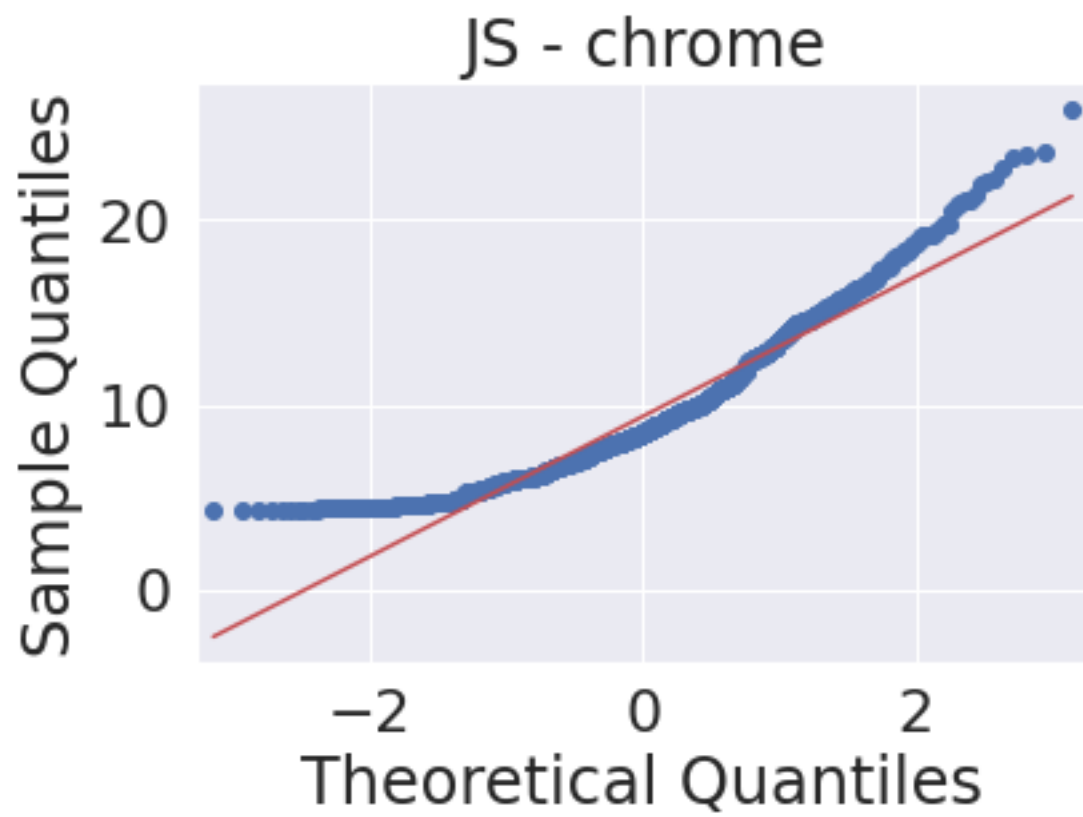
```

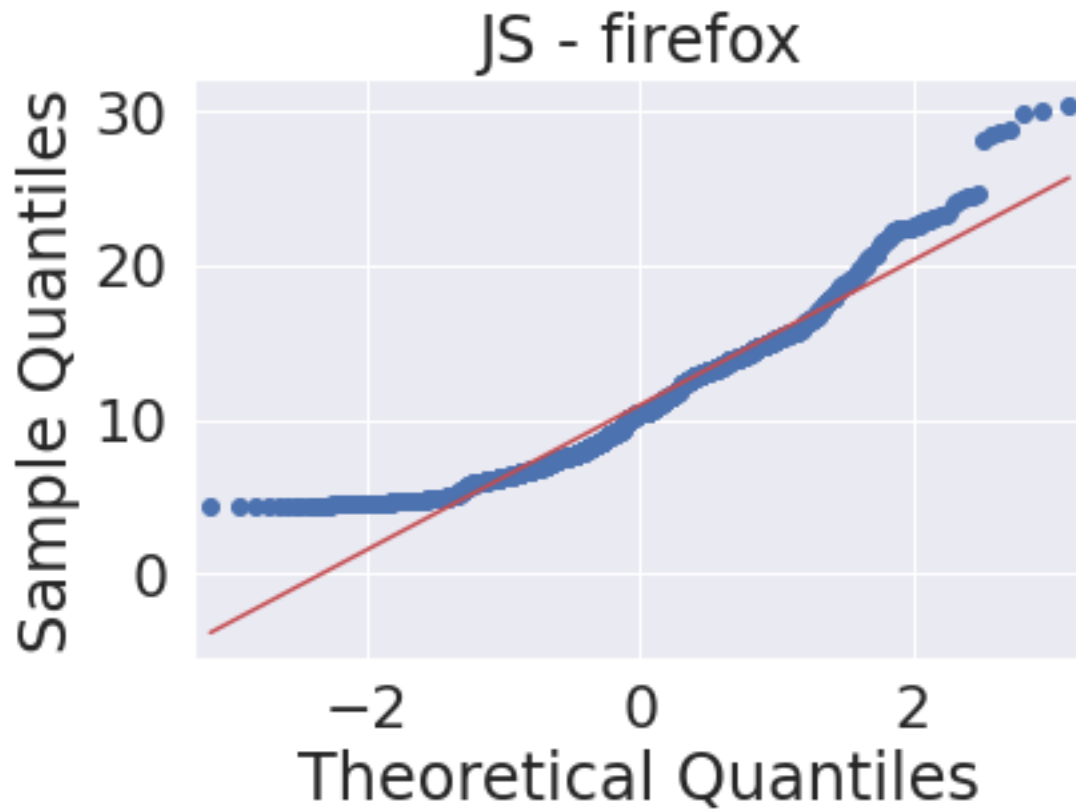
5.5 Q-Q-Plot

```

[38]: for browser in browsers:
        data = df[(df['implementation'] == 'js') & (df['browser'] == browser)]
        qq = sm.qqplot(data.energy, line='s')
        h = plt.title('JS - ' + browser)

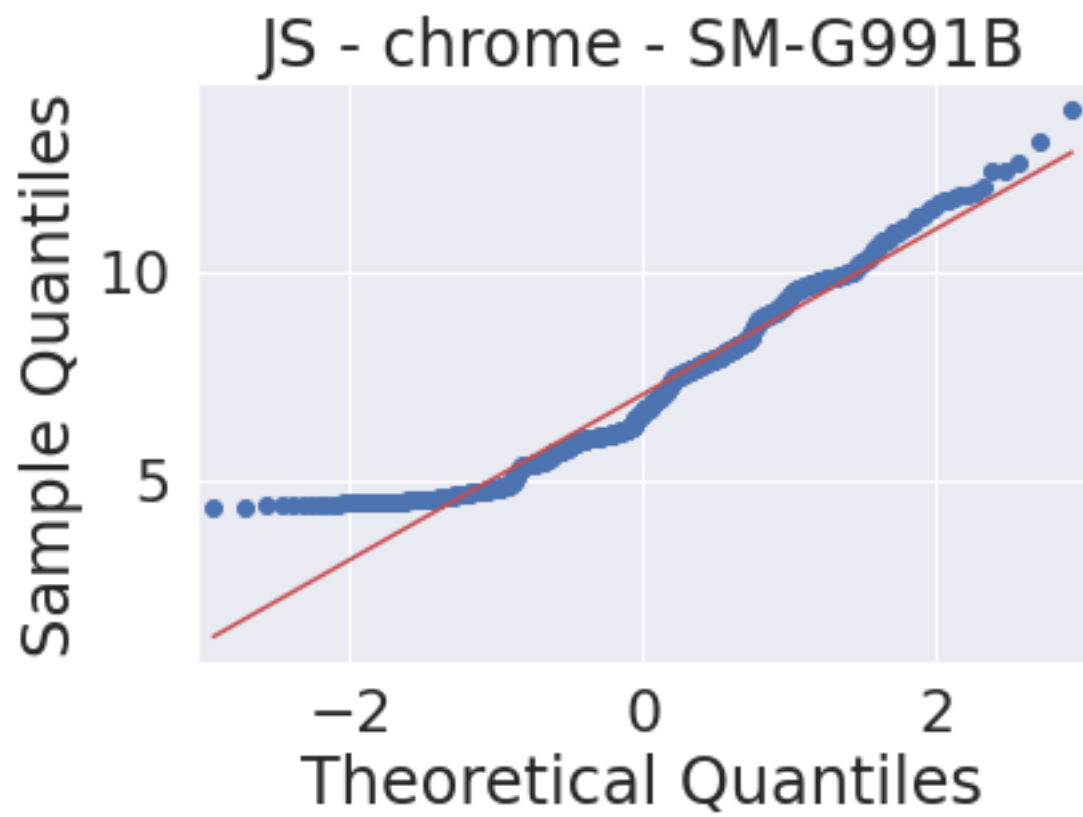
```

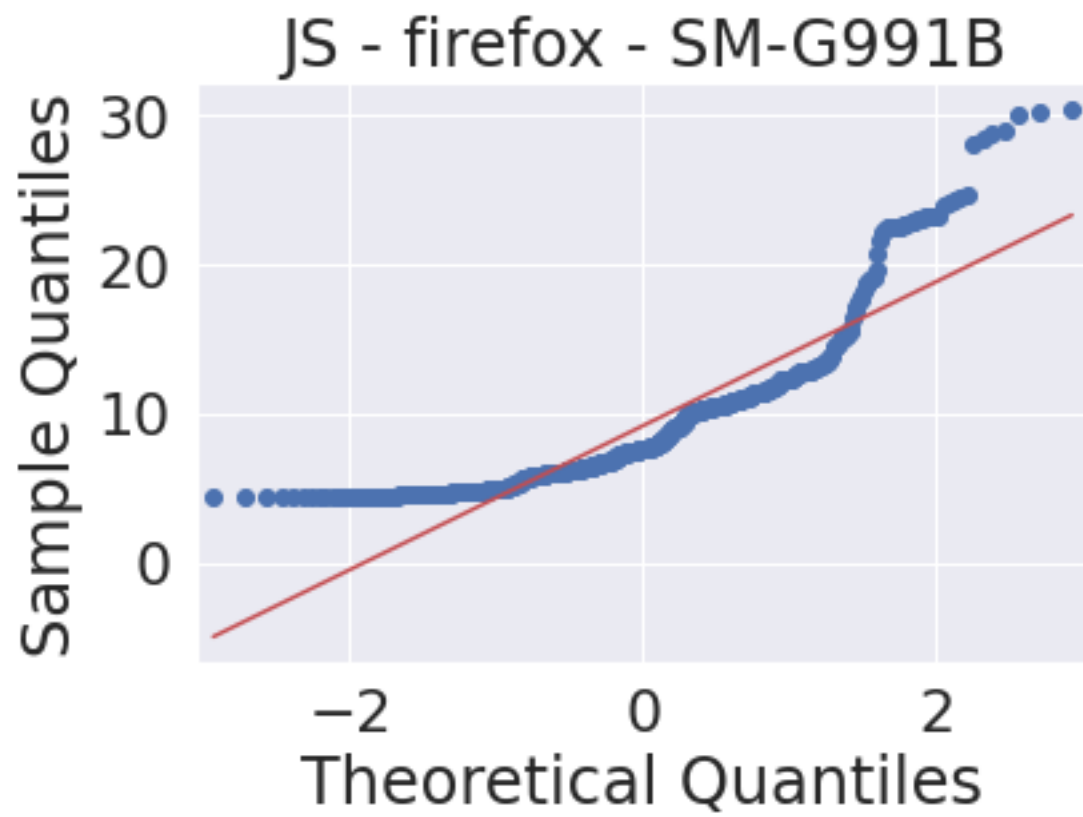


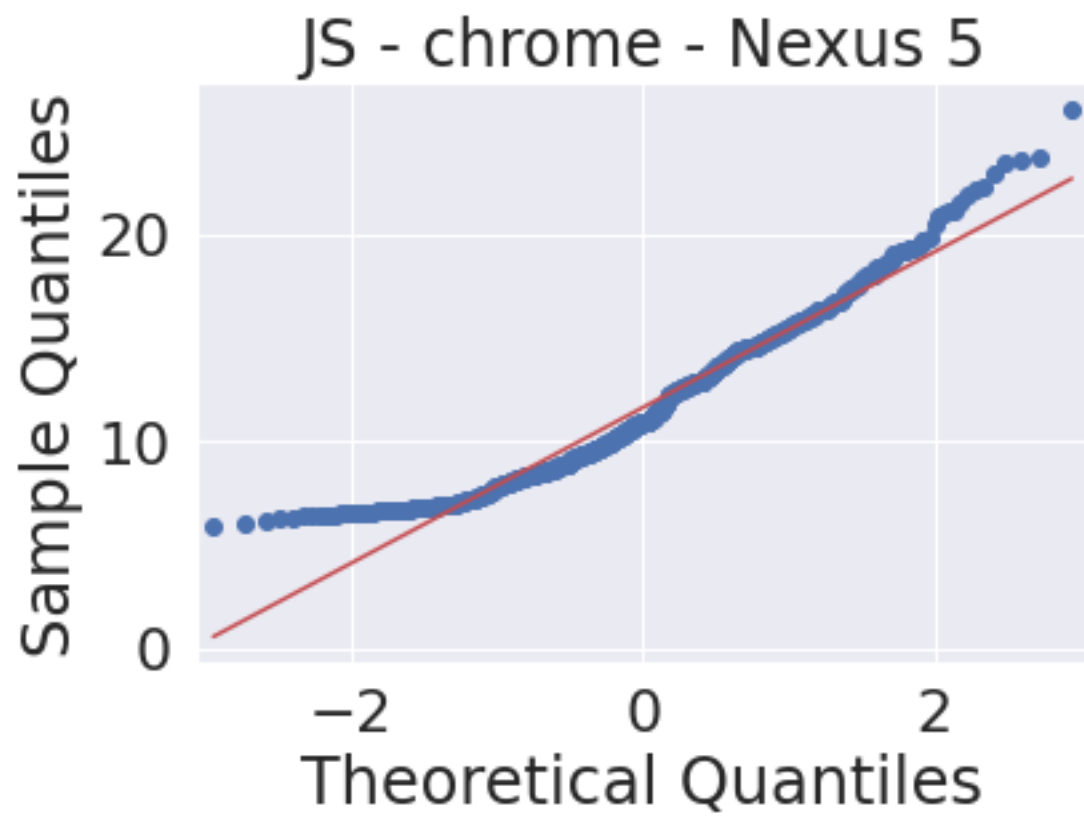


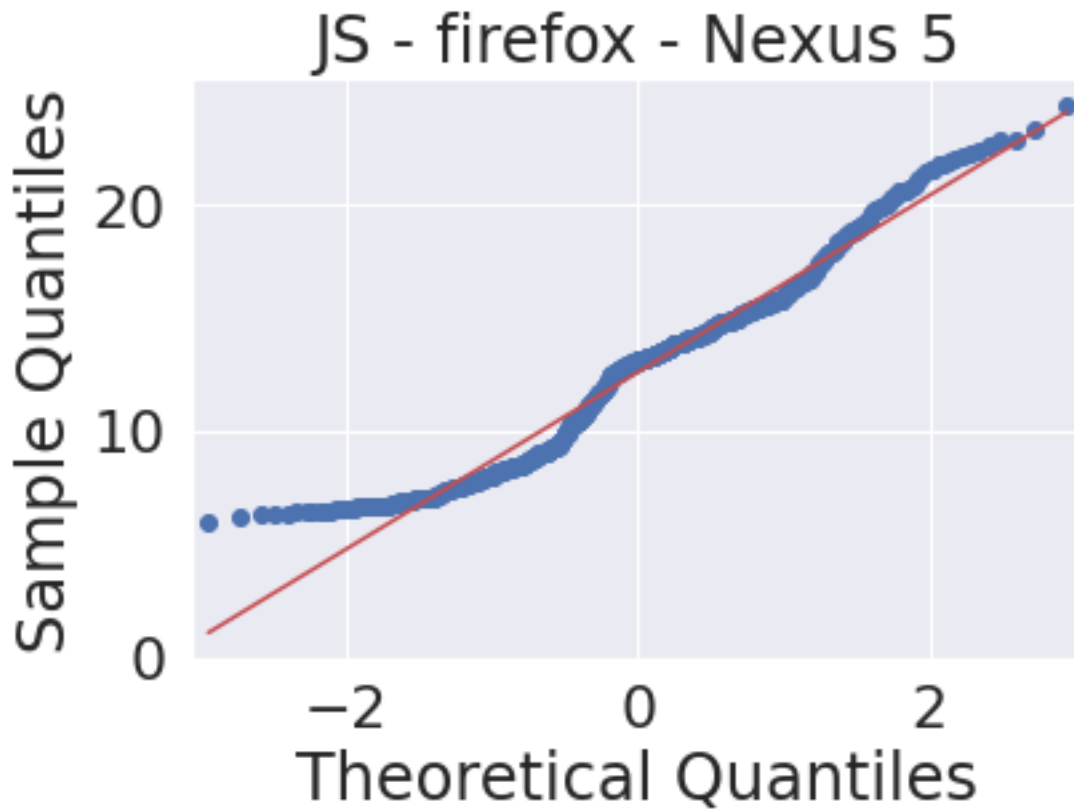
5.6 Q-Q-Plot (By Device)

```
[39]: for device in devices:
        for browser in browsers:
            data = df[(df['implementation'] == 'js') & (df['browser'] == browser) &
                ↪(df['device'] == device)]
            qq = sm.qqplot(data.energy, line='s')
            h = plt.title('JS - ' + browser + ' - ' + device)
```









5.7 Histogram

```
[40]: data = df[(df['implementation'] == 'js')]
sns.histplot(data=data, x="energy", hue="browser", hue_order=browsers).
    ↪set_title("JS Energy Consumption by Browsers")
plt.xlabel("Energy (Joules)")
plt.ylim(0, 200)
plt.xlim(0, 35)
```

```
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
```

```
    if pd.api.types.is_categorical_dtype(vector):
```

```
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
```

```
    if pd.api.types.is_categorical_dtype(vector):
```

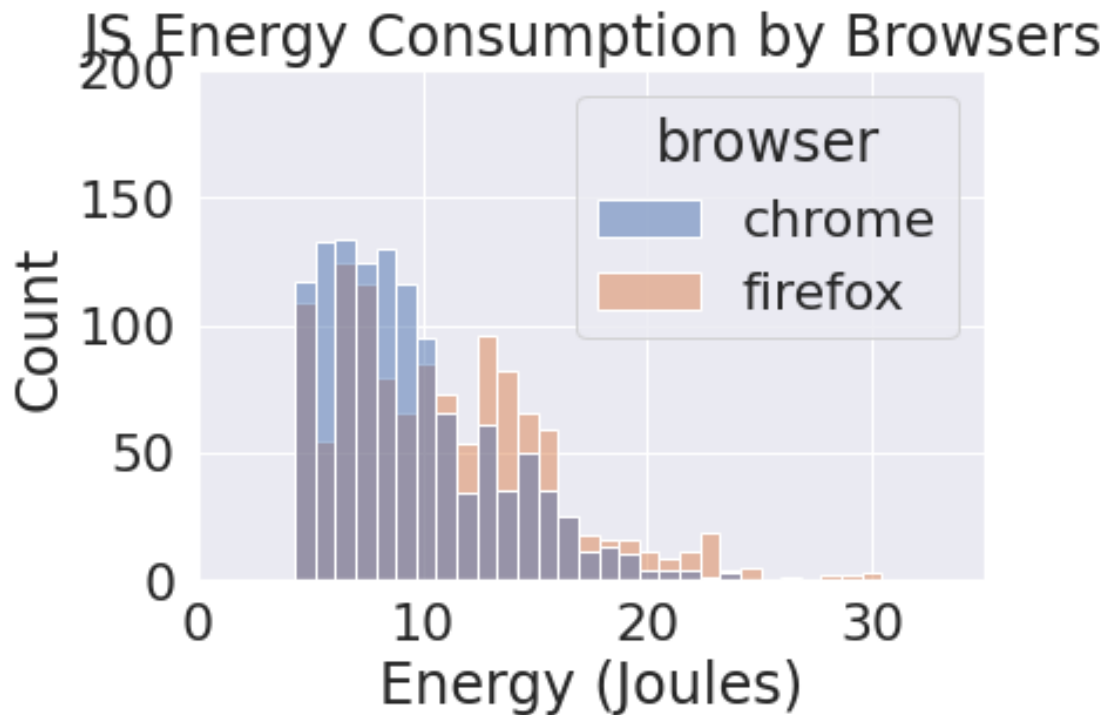
```
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
```

```
    if pd.api.types.is_categorical_dtype(vector):
```

```
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1119: FutureWarning:
use_inf_as_na option is deprecated and will be removed in a future version.
Convert inf values to NaN before operating instead.
```

```
with pd.option_context('mode.use_inf_as_na', True):
```

```
[40]: (0.0, 35.0)
```



5.8 Histogramm (By Device)

```
[41]: data = []
for device in devices:
    data = df[(df['implementation'] == 'js') & (df['device'] == device)]
    sns.histplot(data=data, x="energy", hue="browser", hue_order=browsers).
    ↪set_title("JS Energy Consumption by Browsers" + " - " + device)
    plt.xlabel("Energy (Joules)")
    plt.ylim(0, 200)
    plt.xlim(0, 35)
    plt.show()
```

```
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
```

```
if pd.api.types.is_categorical_dtype(vector):
```

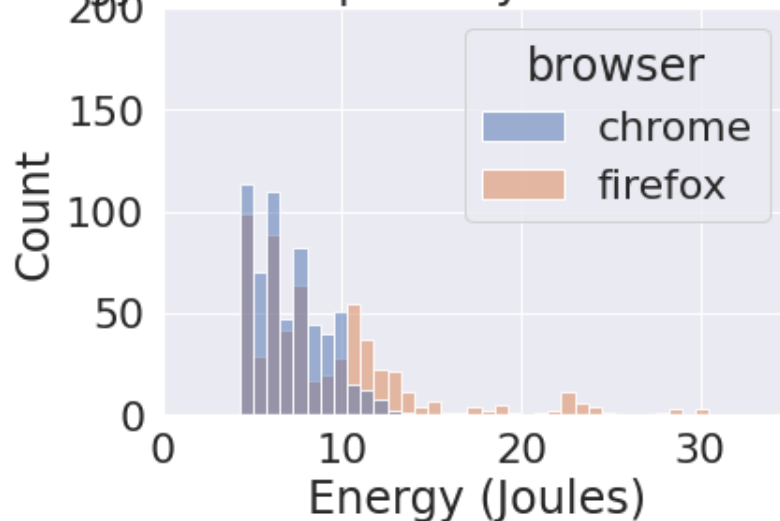
```
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
```

```

is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1119: FutureWarning:
use_inf_as_na option is deprecated and will be removed in a future version.
Convert inf values to NaN before operating instead.
    with pd.option_context('mode.use_inf_as_na', True):

```

JS Energy Consumption by Browsers - SM-G991B

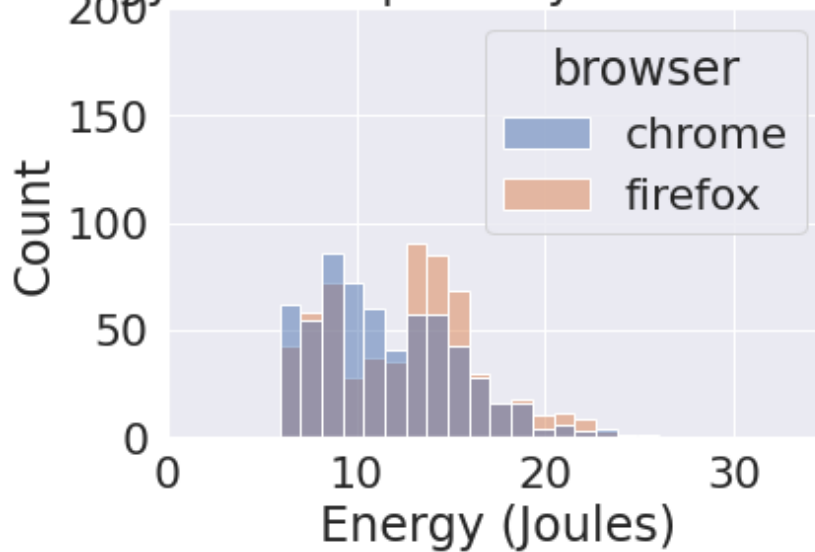


```

/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1119: FutureWarning:
use_inf_as_na option is deprecated and will be removed in a future version.
Convert inf values to NaN before operating instead.
    with pd.option_context('mode.use_inf_as_na', True):

```

JS Energy Consumption by Browsers - Nexus 5



5.9 Descriptive Statistics

```
[42]: data = []
for browser in browsers:
    x = df[(df['browser'] == browser) & (df['implementation'] == 'js')]
    mean = numpy.round(numpy.mean(x['energy']), 2)
    median = numpy.round(numpy.median(x['energy']), 2)
    min = numpy.round(numpy.amin(x['energy']), 2)
    max = numpy.round(numpy.amax(x['energy']), 2)
    std = numpy.round(numpy.std(x['energy']), 2)
    sem = numpy.round(stats.sem(x['energy']), 2)
    q1 = numpy.round(numpy.quantile(x['energy'], 0.25), 2)
    q3 = numpy.round(numpy.quantile(x['energy'], 0.75), 2)

    data.append(
        [browser, mean, std, min, q1, median, q3, max, sem]
    )

# Create the pandas DataFrame
stat = pd.DataFrame(data, columns = ['browser', 'mean', 'std', 'min', 'q1', 'q2', 'q3', 'median', 'max', 'sem'])
display(stat)
#print(stat.to_string())

# Alternative of pandas: x['energy'].describe()
```

browser	mean	std	min	q1	median	q3	max	sem
---------	------	-----	-----	----	--------	----	-----	-----

0	chrome	9.37	3.79	4.30	6.52	8.47	11.28	26.04	0.11
1	firefox	10.94	4.70	4.39	7.11	10.43	13.89	30.46	0.14

5.10 Descriptive Statistics Difference

```
[43]: data = []

for browserpair in browserpairs:
    browser1 = stat[(stat['browser'] == browserpair[0])]
    browser2 = stat[(stat['browser'] == browserpair[1])]

    mean_diff = browser1.iloc[0]['mean']-browser2.iloc[0]['mean']
    median_diff = browser1.iloc[0]['median']-browser2.iloc[0]['median']
    min_diff = browser1.iloc[0]['min']-browser2.iloc[0]['min']
    max_diff = browser1.iloc[0]['max']-browser2.iloc[0]['max']
    std_diff = browser1.iloc[0]['std']-browser2.iloc[0]['std']
    sem_diff = browser1.iloc[0]['sem']-browser2.iloc[0]['sem']
    q1_diff = browser1.iloc[0]['q1']-browser2.iloc[0]['q1']
    q3_diff = browser1.iloc[0]['q3']-browser2.iloc[0]['q3']

    data.append(
        [browserpair[0] + ' vs. ' + browserpair[1],
         numpy.round(mean_diff, 2),
         numpy.round(mean_diff/browser2.iloc[0]['mean']*100, 2),
         numpy.round(median_diff, 2),
         numpy.round(median_diff/browser2.iloc[0]['median']*100, 2),
         numpy.round(min_diff, 2),
         numpy.round(min_diff/browser2.iloc[0]['min']*100, 2),
         numpy.round(max_diff, 2),
         numpy.round(max_diff/browser2.iloc[0]['max']*100, 2),
         numpy.round(std_diff, 2),
         numpy.round(std_diff/browser2.iloc[0]['std']*100, 2),
         numpy.round(sem_diff, 2),
         numpy.round(sem_diff/browser2.iloc[0]['sem']*100, 2),
         numpy.round(q1_diff, 2),
         numpy.round(q1_diff/browser2.iloc[0]['q1']*100, 2),
         numpy.round(q3_diff, 2),
         numpy.round(q3_diff/browser2.iloc[0]['q3']*100, 2),
        ]
    )

# Create the pandas DataFrame
ut = pd.DataFrame(data, columns =_
    ↳ ['rq', 'mean_diff', 'mean_diff%', 'median_diff', 'median_diff%', 'min_diff', 'min_diff%', 'max_diff', 'max_diff%', 'q1_diff', 'q1_diff%', 'q3_diff', 'q3_diff%'])
display(ut)
#print(ut.to_string())
```

```
rq mean_diff mean_diff% median_diff median_diff% \
```

0	chrome vs. firefox	-1.57	-14.35	-1.96	-18.79
---	--------------------	-------	--------	-------	--------

	min_diff	min_diff%	max_diff	max_diff%	std_diff	std_diff%	sem_diff	\
0	-0.09	-2.05	-4.42	-14.51	-0.91	-19.36	-0.03	

	sem_diff%	q1_diff	q1_diff%	q3_diff	q3_diff%
0	-21.43	-0.59	-8.3	-2.61	-18.79

5.11 Descriptive Statistics (By Device)

```
[44]: data = []
      for device in devices:
          for browser in browsers:
              x = df[(df['browser'] == browser) & (df['implementation'] == 'js') &
↳ (df['device'] == device)]
              mean = numpy.round(numpy.mean(x['energy']), 2)
              median = numpy.round(numpy.median(x['energy']), 2)
              min = numpy.round(numpy.amin(x['energy']), 2)
              max = numpy.round(numpy.amax(x['energy']), 2)
              std = numpy.round(numpy.std(x['energy']), 2)
              sem = numpy.round(stats.sem(x['energy']), 2)
              q1 = numpy.round(numpy.quantile(x['energy'], 0.25), 2)
              q3 = numpy.round(numpy.quantile(x['energy'], 0.75), 2)

              data.append(
                  [device, browser, mean, std, min, q1, median, q3, max, sem]
              )

      # Create the pandas DataFrame
      stat = pd.DataFrame(data, columns = ['device', 'browser', 'mean', 'std', 'min',
↳ 'q1', 'median', 'q3', 'max', 'sem'])
      display(stat)
      #print(stat.to_string())

      # Alternative of pandas: x['energy'].describe()
```

	device	browser	mean	std	min	q1	median	q3	max	sem
0	SM-G991B	chrome	7.05	2.00	4.30	5.46	6.62	8.29	13.96	0.08
1	SM-G991B	firefox	9.21	4.82	4.39	5.99	7.67	11.03	30.46	0.20
2	Nexus 5	chrome	11.64	3.75	5.92	8.53	10.93	14.44	26.04	0.15
3	Nexus 5	firefox	12.61	3.91	6.06	9.03	13.13	15.01	24.44	0.16

5.12 Descriptive Statistics Difference (By Device)

```
[45]: data = []

for device in devices:
    for browserpair in browserpairs:
        browser1 = stat[(stat['browser'] == browserpair[0]) & (stat['device']_
↪ == device)]
        browser2 = stat[(stat['browser'] == browserpair[1]) & (stat['device']_
↪ == device)]

        mean_diff = browser1.iloc[0]['mean']-browser2.iloc[0]['mean']
        median_diff = browser1.iloc[0]['median']-browser2.iloc[0]['median']
        min_diff = browser1.iloc[0]['min']-browser2.iloc[0]['min']
        max_diff = browser1.iloc[0]['max']-browser2.iloc[0]['max']
        std_diff = browser1.iloc[0]['std']-browser2.iloc[0]['std']
        sem_diff = browser1.iloc[0]['sem']-browser2.iloc[0]['sem']
        q1_diff = browser1.iloc[0]['q1']-browser2.iloc[0]['q1']
        q3_diff = browser1.iloc[0]['q3']-browser2.iloc[0]['q3']

        data.append(
            [
                device,
                browserpair[0] + ' vs. ' + browserpair[1],
                numpy.round(mean_diff, 2),
                numpy.round(mean_diff/browser2.iloc[0]['mean']*100, 2),
                numpy.round(median_diff, 2),
                numpy.round(median_diff/browser2.iloc[0]['median']*100, 2),
                numpy.round(min_diff, 2),
                numpy.round(min_diff/browser2.iloc[0]['min']*100, 2),
                numpy.round(max_diff, 2),
                numpy.round(max_diff/browser2.iloc[0]['max']*100, 2),
                numpy.round(std_diff, 2),
                numpy.round(std_diff/browser2.iloc[0]['std']*100, 2),
                numpy.round(sem_diff, 2),
                numpy.round(sem_diff/browser2.iloc[0]['sem']*100, 2),
                numpy.round(q1_diff, 2),
                numpy.round(q1_diff/browser2.iloc[0]['q1']*100, 2),
                numpy.round(q3_diff, 2),
                numpy.round(q3_diff/browser2.iloc[0]['q3']*100, 2),
            ]
        )

# Create the pandas DataFrame
ut = pd.DataFrame(data, columns = ['device',_
↪ 'rq','mean_diff','mean_diff%','median_diff','median_diff%','min_diff','min_diff%','max_diff'])
display(ut)
```

```
#print(ut.to_string())
```

	device		rq	mean_diff	mean_diff%	median_diff	\
0	SM-G991B	chrome vs. firefox		-2.16	-23.45	-1.05	
1	Nexus 5	chrome vs. firefox		-0.97	-7.69	-2.20	

	median_diff%	min_diff	min_diff%	max_diff	max_diff%	std_diff	\
0	-13.69	-0.09	-2.05	-16.5	-54.17	-2.82	
1	-16.76	-0.14	-2.31	1.6	6.55	-0.16	

	std_diff%	sem_diff	sem_diff%	q1_diff	q1_diff%	q3_diff	q3_diff%
0	-58.51	-0.12	-60.00	-0.53	-8.85	-2.74	-24.84
1	-4.09	-0.01	-6.25	-0.50	-5.54	-0.57	-3.80

6 RQ2: WASM Energy Browser

6.1 Shapiro Wilk Test

```
[46]: data = []
non_normal = 0

for browser in browsers:
    energy = df[(df['browser'] == browser) & (df['implementation'] == 'wasm')]['energy']

    if len(energy) >= 3:
        shapiro_test = stats.shapiro(energy)

        non_normal += (1 if shapiro_test.pvalue <= 0.05 else 0)

    data.append(
        [browser, 'wasm',
         shapiro_test.statistic,
         shapiro_test.pvalue
        ]
    )

# Create the pandas DataFrame
swt = pd.DataFrame(data, columns = ['browser', 'implementation', 'w', 'p'])
#print(swt.to_string())
display(swt)

print("\n{} non-normally distributed samples".format(non_normal))
print("{} normally distributed samples".format(len(swt) - non_normal))
print("{:.2f}% non-normally distributed samples".format(non_normal/len(swt)*100))
```

	browser	implementation	w	p
0	chrome	wasm	0.837385	2.093366e-33
1	firefox	wasm	0.882936	3.217827e-29

2 non-normally distributed samples
0 normally distributed samples
100.00% non-normally distributed samples

6.2 Shapiro Wilk Test (By Device)

```
[47]: data = []
non_normal = 0

for device in devices:
    for browser in browsers:
        energy = df[(df['browser'] == browser) & (df['implementation'] == 'wasm') & (df['device'] == device)]['energy']

        if len(energy) >= 3:
            shapiro_test = stats.shapiro(energy)

            non_normal += (1 if shapiro_test.pvalue <= 0.05 else 0)

            data.append([
                device,
                browser, 'wasm',
                shapiro_test.statistic,
                shapiro_test.pvalue
            ])

# Create the pandas DataFrame
swt = pd.DataFrame(data, columns = ['device', 'browser', 'implementation', 'w', 'p'])
#print(swt.to_string())
display(swt)

print("\n{} non-normally distributed samples".format(non_normal))
print("{} normally distributed samples".format(len(swt) - non_normal))
print("{:.2f}% non-normally distributed samples".format(non_normal/len(swt)*100))
```

	device	browser	implementation	w	p
0	SM-G991B	chrome	wasm	0.911131	4.198296e-18
1	SM-G991B	firefox	wasm	0.723657	3.320376e-30
2	Nexus 5	chrome	wasm	0.828244	3.247344e-25

```
3 Nexus 5 firefox wasm 0.879442 2.271143e-21
```

```
4 non-normally distributed samples
0 normally distributed samples
100.00% non-normally distributed samples
```

6.3 Mann Whitney U Test

```
[48]: data = []

for browserpair in browserpairs:
    browser1_energy = df[(df['browser'] == browserpair[0]) &
↳(df['implementation'] == 'wasm')]['energy']
    browser2_energy = df[(df['browser'] == browserpair[1]) &
↳(df['implementation'] == 'wasm')]['energy']
    eff = cliff.cliffs_delta(browser1_energy, browser2_energy)

    u = stats.mannwhitneyu(browser1_energy, browser2_energy,
↳alternative='two-sided')

    data.append(
        [browserpair[0] + ' vs. ' + browserpair[1],
        u.statistic,
        u.pvalue,
        eff[0],
        eff[1]
        ]
    )

# Create the pandas DataFrame
ut = pd.DataFrame(data, columns = ['rq', 'u', 'p', 'eff', 'interp'])
display(ut)

interp = ut['interp'].value_counts()
interp = pd.DataFrame(interp, columns = ['interp', 'percent'])
interp['percent'] = (interp['interp'] / interp['interp'].sum()) * 100

display(interp)
```

```
           rq           u           p           eff           interp
0 chrome vs. firefox 655827.0 0.000019 -0.100408 negligible
```

```
Empty DataFrame
Columns: [interp, percent]
Index: []
```

6.4 Mann Whitney U Test (By Device)

```
[49]: data = []

for device in devices:
    for browserpair in browserpairs:
        browser1_energy = df[(df['browser'] == browserpair[0]) &
↪(df['implementation'] == 'wasm') & (df['device'] == device)]['energy']
        browser2_energy = df[(df['browser'] == browserpair[1]) &
↪(df['implementation'] == 'wasm') & (df['device'] == device)]['energy']
        eff = cliff.cliffs_delta(browser1_energy, browser2_energy)

        u = stats.mannwhitneyu(browser1_energy, browser2_energy,
↪alternative='two-sided')

        data.append(
            [
                device,
                browserpair[0] + ' vs. ' + browserpair[1],
                u.statistic,
                u.pvalue,
                eff[0],
                eff[1]
            ]
        )

# Create the pandas DataFrame
ut = pd.DataFrame(data, columns = ['device', 'rq', 'u', 'p', 'eff', 'interp'])
display(ut)

interp = ut['interp'].value_counts()
interp = pd.DataFrame(interp, columns = ['interp', 'percent'])
interp['percent'] = (interp['interp'] / interp['interp'].sum()) * 100

display(interp)
```

	device		rq	u	p	eff	interp
0	SM-G991B	chrome vs. firefox		154952.0	0.000402	-0.118702	negligible
1	Nexus 5	chrome vs. firefox		162482.0	0.000023	-0.139392	negligible

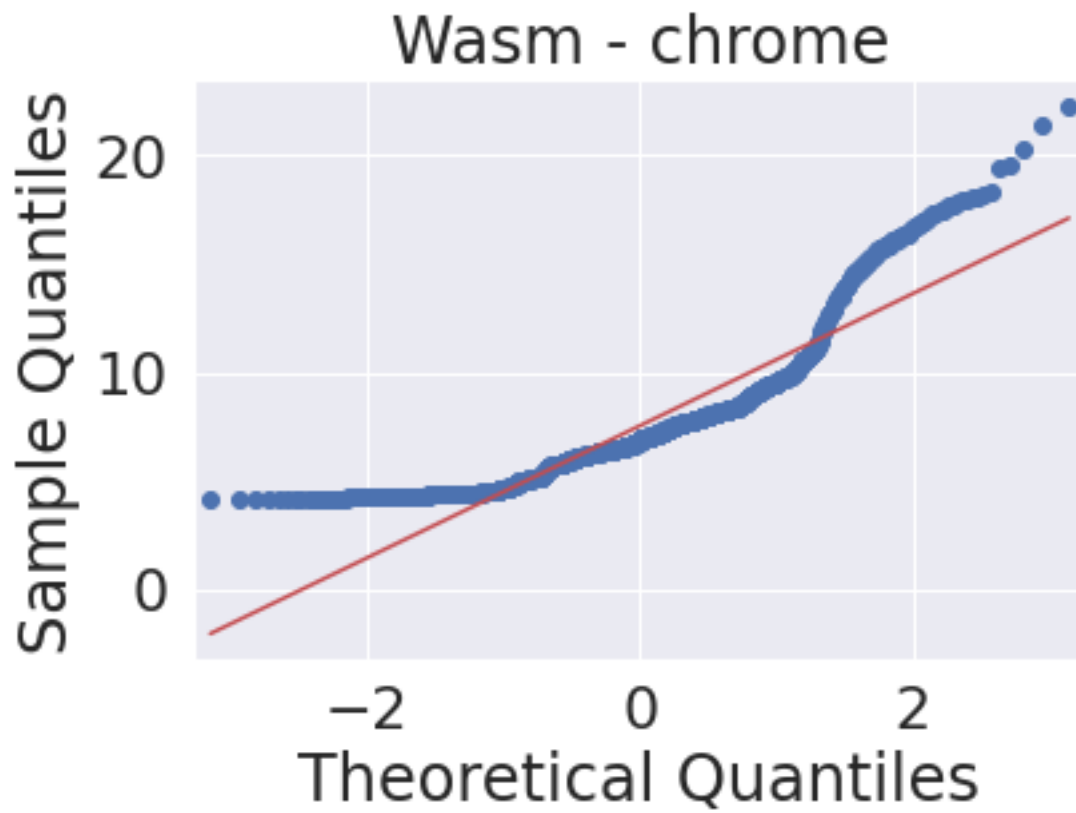
Empty DataFrame

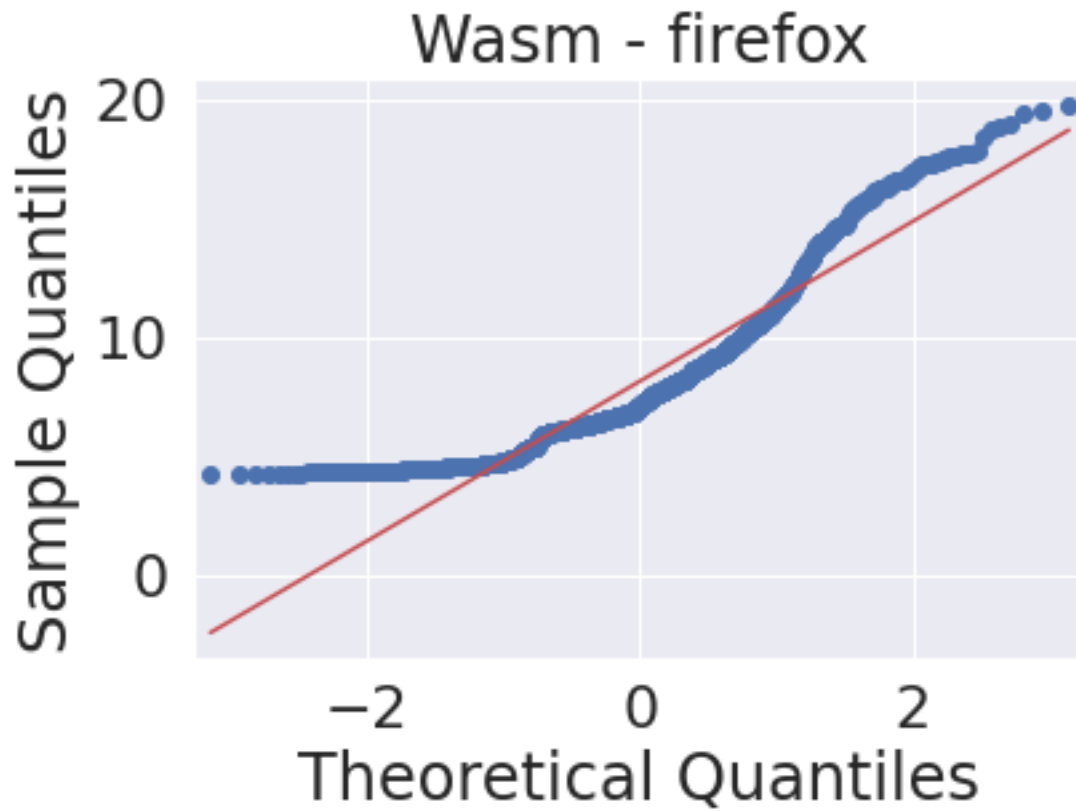
Columns: [interp, percent]

Index: []

6.5 Q-Q-Plot

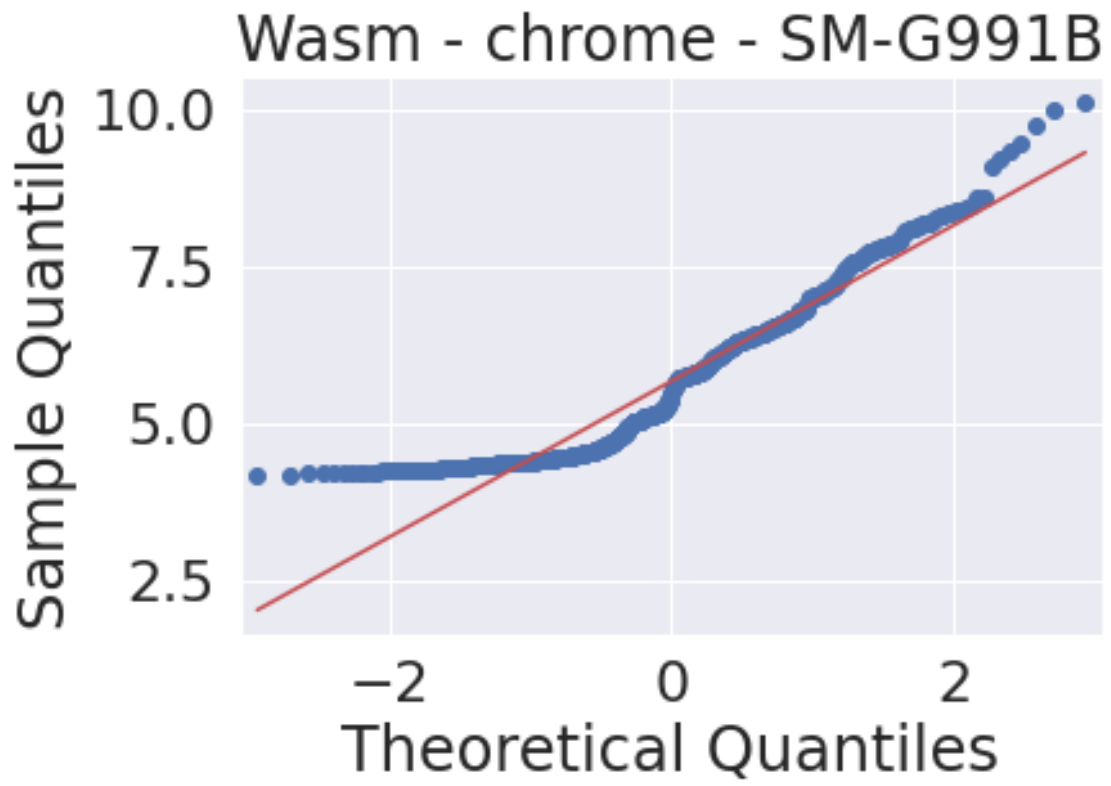
```
[50]: for browser in browsers:  
      data = df[(df['implementation'] == 'wasm') & (df['browser'] == browser)]  
      qq = sm.qqplot(data.energy, line='s')  
      h = plt.title('Wasm - ' + browser)
```

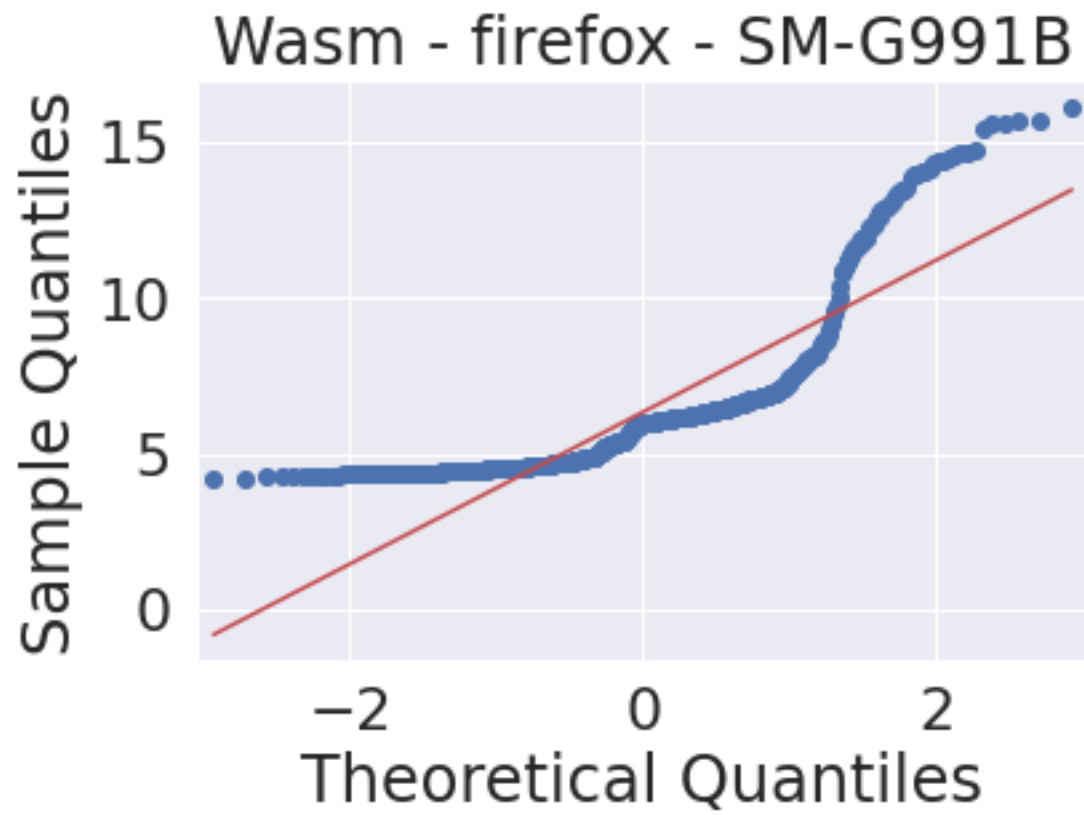


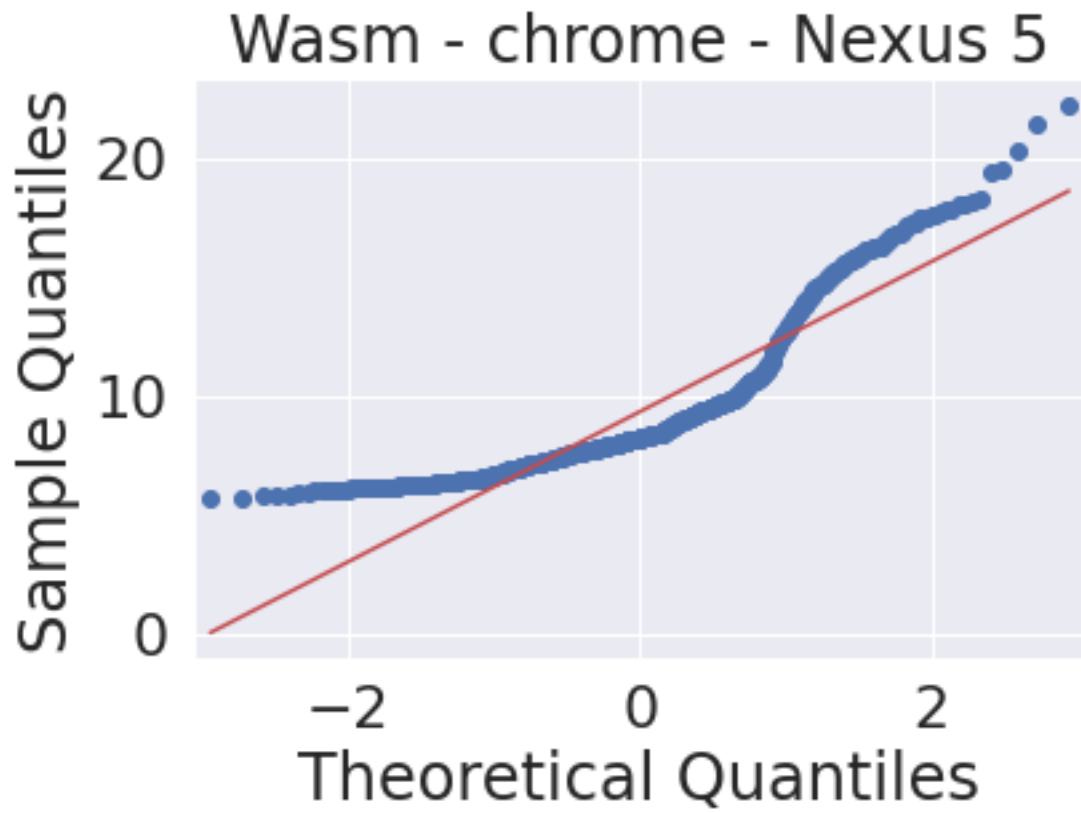


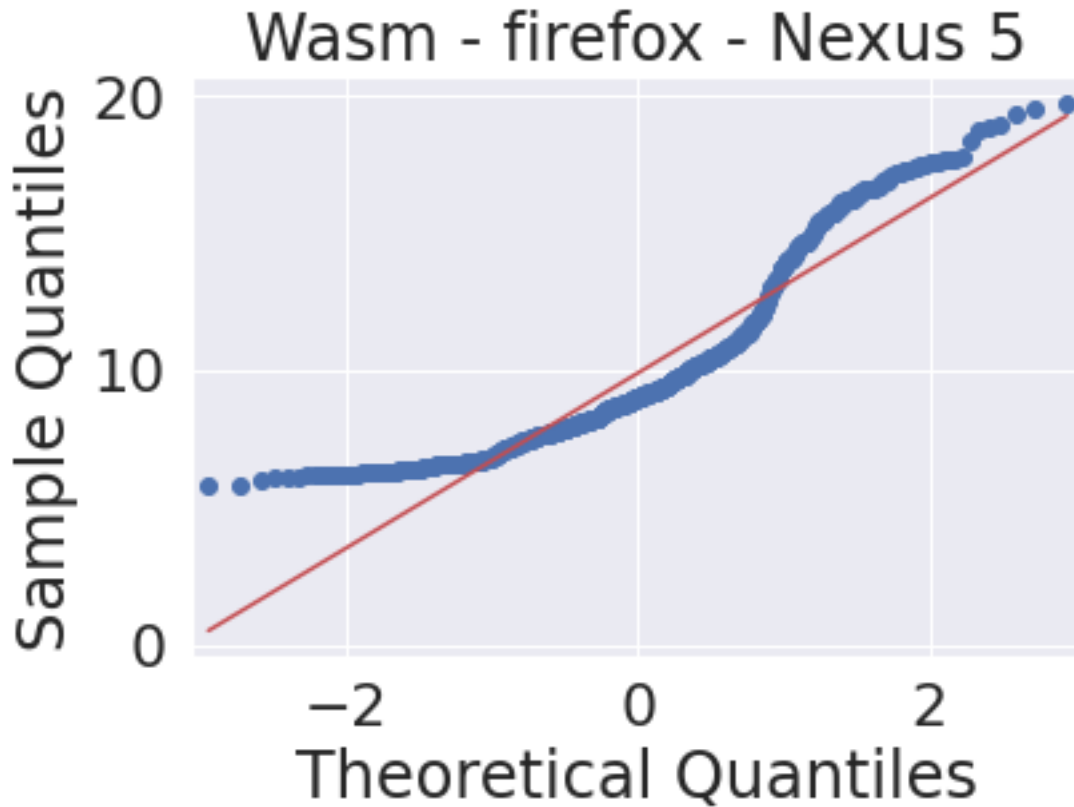
6.6 Q-Q-Plot (By Device)

```
[51]: for device in devices:
        for browser in browsers:
            data = df[(df['implementation'] == 'wasm') & (df['browser'] == browser)
↪ & (df['device'] == device)]
            qq = sm.qqplot(data.energy, line='s')
            h = plt.title('Wasm - ' + browser + ' - ' + device)
```









6.7 Histogram

```
[52]: data = df[(df['implementation'] == 'wasm')]
sns.histplot(data=data, x="energy", hue="browser", hue_order=browsers).
    ↪set_title("Wasm Energy Consumption by Browsers")
plt.xlabel("Energy (Joules)")
plt.ylim(0, 200)
plt.xlim(0, 30)
```

```
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
```

```
    if pd.api.types.is_categorical_dtype(vector):
```

```
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
```

```
    if pd.api.types.is_categorical_dtype(vector):
```

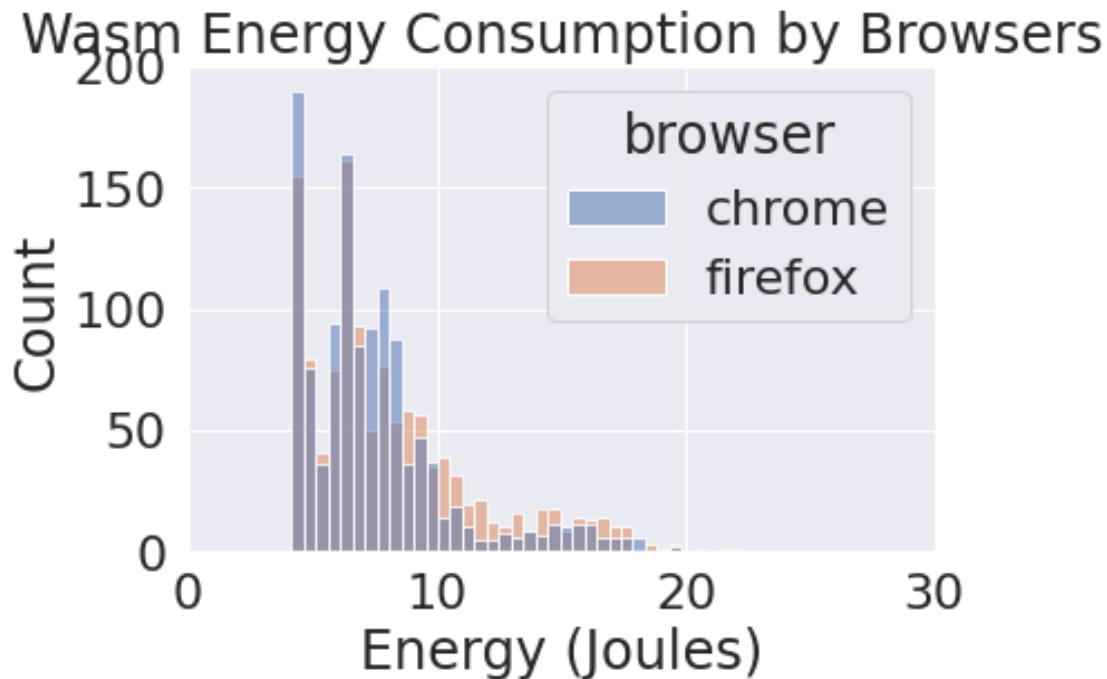
```
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
```

```
    if pd.api.types.is_categorical_dtype(vector):
```

```
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1119: FutureWarning:
use_inf_as_na option is deprecated and will be removed in a future version.
Convert inf values to NaN before operating instead.
```

```
with pd.option_context('mode.use_inf_as_na', True):
```

```
[52]: (0.0, 30.0)
```



6.8 Histogram (By Device)

```
[53]: data = []
for device in devices:
    data = df[(df['implementation'] == 'wasm') & (df['device'] == device)]
    sns.histplot(data=data, x="energy", hue="browser", hue_order=browsers).
    ↪set_title("Wasm Energy Consumption by Browsers")
    plt.xlabel("Energy (Joules)")
    plt.ylim(0, 175)
    plt.xlim(0, 30)
    plt.show()
```

```
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
```

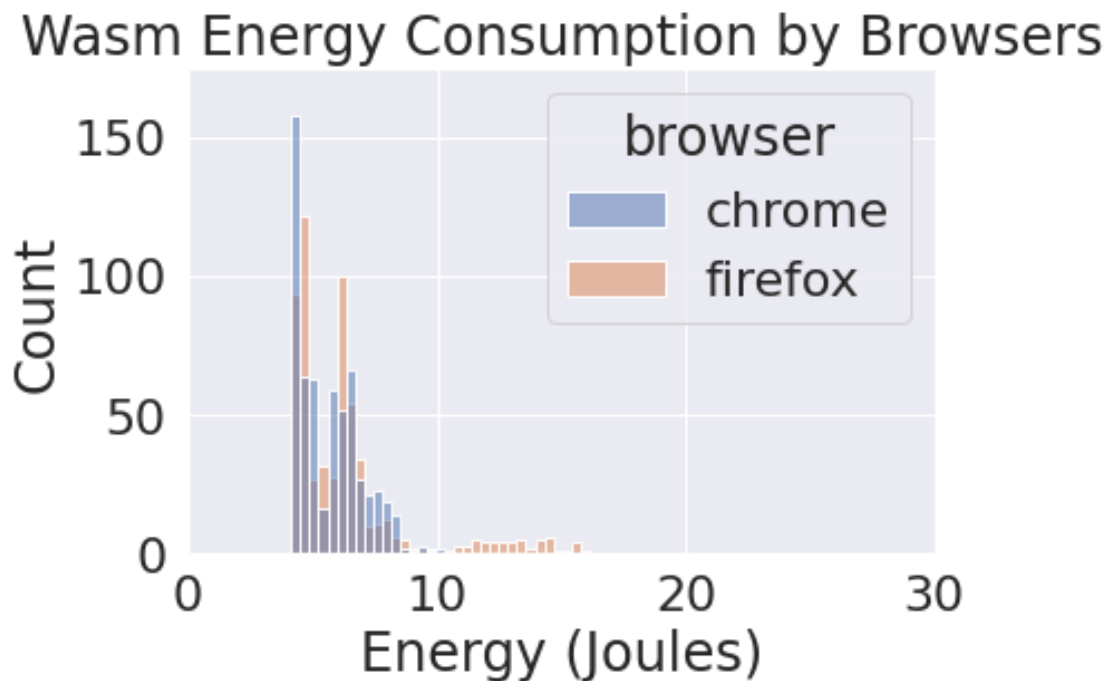
```
if pd.api.types.is_categorical_dtype(vector):
```

```
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
```

```

isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1119: FutureWarning:
use_inf_as_na option is deprecated and will be removed in a future version.
Convert inf values to NaN before operating instead.
    with pd.option_context('mode.use_inf_as_na', True):

```

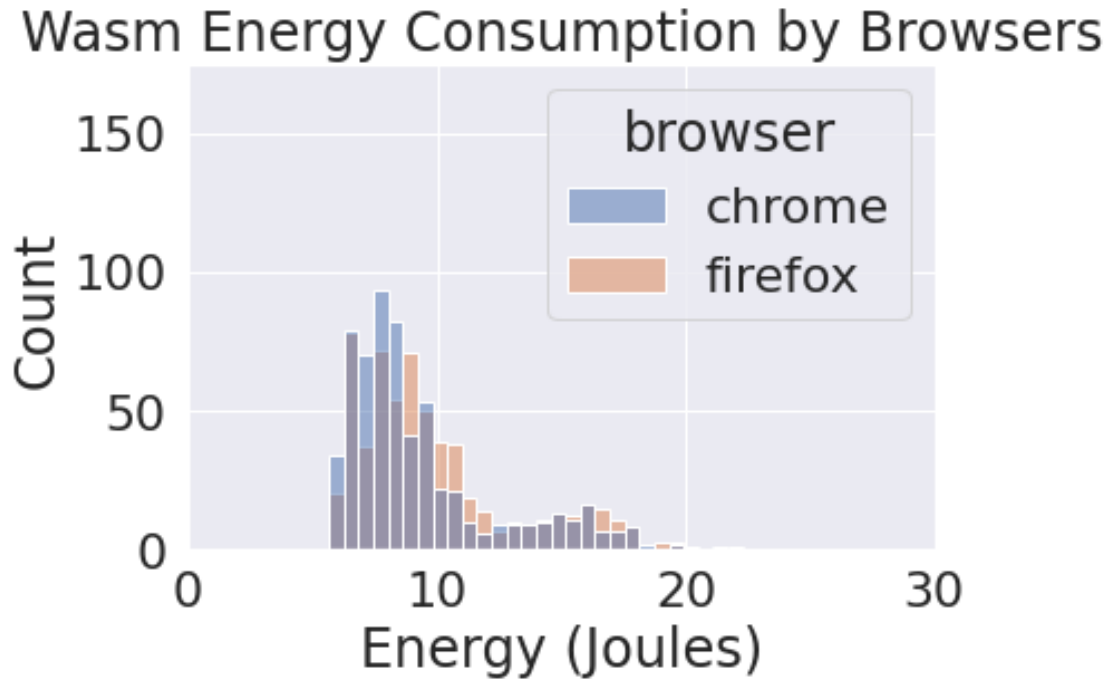


```

/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1498: FutureWarning:
is_categorical_dtype is deprecated and will be removed in a future version. Use
isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
/opt/conda/lib/python3.9/site-packages/seaborn/_oldcore.py:1119: FutureWarning:
use_inf_as_na option is deprecated and will be removed in a future version.

```

Convert inf values to NaN before operating instead.
 with pd.option_context('mode.use_inf_as_na', True):



6.9 Descriptive Statistics

```
[54]: data = []
for browser in browsers:
    x = df[(df['browser'] == browser) & (df['implementation'] == 'wasm')]
    mean = numpy.round(numpy.mean(x['energy']), 2)
    median = numpy.round(numpy.median(x['energy']), 2)
    min = numpy.round(numpy.amin(x['energy']), 2)
    max = numpy.round(numpy.amax(x['energy']), 2)
    std = numpy.round(numpy.std(x['energy']), 2)
    sem = numpy.round(stats.sem(x['energy']), 2)
    q1 = numpy.round(numpy.quantile(x['energy'], 0.25), 2)
    q3 = numpy.round(numpy.quantile(x['energy'], 0.75), 2)

    data.append(
        [browser, mean, std, min, q1, median, q3, max, sem]
    )

# Create the pandas DataFrame
stat = pd.DataFrame(data, columns = ['browser', 'mean', 'std', 'min', 'q1', 'q2', 'q3', 'max', 'sem'])
```

```
# display(stat)
print(stat.to_string())

# Alternative of pandas: x['energy'].describe()
```

	browser	mean	std	min	q1	median	q3	max	sem
0	chrome	7.55	3.04	4.16	5.64	6.85	8.37	22.29	0.09
1	firefox	8.14	3.36	4.22	5.96	7.09	9.67	19.78	0.10

6.10 Descriptive Statistics Difference

```
[55]: data = []

for browserpair in browserpairs:
    browser1 = stat[(stat['browser'] == browserpair[0])]
    browser2 = stat[(stat['browser'] == browserpair[1])]

    mean_diff = browser1.iloc[0]['mean']-browser2.iloc[0]['mean']
    median_diff = browser1.iloc[0]['median']-browser2.iloc[0]['median']
    min_diff = browser1.iloc[0]['min']-browser2.iloc[0]['min']
    max_diff = browser1.iloc[0]['max']-browser2.iloc[0]['max']
    std_diff = browser1.iloc[0]['std']-browser2.iloc[0]['std']
    sem_diff = browser1.iloc[0]['sem']-browser2.iloc[0]['sem']
    q1_diff = browser1.iloc[0]['q1']-browser2.iloc[0]['q1']
    q3_diff = browser1.iloc[0]['q3']-browser2.iloc[0]['q3']

    data.append(
        [browserpair[0] + ' vs. ' + browserpair[1],
         numpy.round(mean_diff, 2),
         numpy.round(mean_diff/browser2.iloc[0]['mean']*100, 2),
         numpy.round(median_diff, 2),
         numpy.round(median_diff/browser2.iloc[0]['median']*100, 2),
         numpy.round(min_diff, 2),
         numpy.round(min_diff/browser2.iloc[0]['min']*100, 2),
         numpy.round(max_diff, 2),
         numpy.round(max_diff/browser2.iloc[0]['max']*100, 2),
         numpy.round(std_diff, 2),
         numpy.round(std_diff/browser2.iloc[0]['std']*100, 2),
         numpy.round(sem_diff, 2),
         numpy.round(sem_diff/browser2.iloc[0]['sem']*100, 2),
         numpy.round(q1_diff, 2),
         numpy.round(q1_diff/browser2.iloc[0]['q1']*100, 2),
         numpy.round(q3_diff, 2),
         numpy.round(q3_diff/browser2.iloc[0]['q3']*100, 2),
        ]
    )
```

```

# Create the pandas DataFrame
ut = pd.DataFrame(data, columns = [
    'rq', 'mean_diff', 'mean_diff%', 'median_diff', 'median_diff%', 'min_diff', 'min_diff%', 'max_diff',
    'max_diff%', 'std_diff', 'std_diff%', 'sem_diff', 'sem_diff%'
])
display(ut)
#print(ut.to_string())

```

	rq	mean_diff	mean_diff%	median_diff	median_diff%				
0	chrome vs. firefox	-0.59	-7.25	-0.24	-3.39				
	min_diff	min_diff%	max_diff	max_diff%	std_diff	std_diff%	sem_diff		
0	-0.06	-1.42	2.51	12.69	-0.32	-9.52	-0.01		
	sem_diff%	q1_diff	q1_diff%	q3_diff	q3_diff%				
0	-10.0	-0.32	-5.37	-1.3	-13.44				

6.11 Descriptive Statistics (By Device)

```

[56]: data = []

for device in devices:
    for browser in browsers:
        x = df[(df['browser'] == browser) & (df['implementation'] == 'wasm')] &
        (df['device'] == device)
        mean = numpy.round(numpy.mean(x['energy']), 2)
        median = numpy.round(numpy.median(x['energy']), 2)
        min = numpy.round(numpy.amin(x['energy']), 2)
        max = numpy.round(numpy.amax(x['energy']), 2)
        std = numpy.round(numpy.std(x['energy']), 2)
        sem = numpy.round(stats.sem(x['energy']), 2)
        q1 = numpy.round(numpy.quantile(x['energy'], 0.25), 2)
        q3 = numpy.round(numpy.quantile(x['energy'], 0.75), 2)

        data.append(
            [device, browser, mean, std, min, q1, median, q3, max, sem]
        )

# Create the pandas DataFrame
stat = pd.DataFrame(data, columns = ['device', 'browser', 'mean', 'std', 'min',
    'q1', 'median', 'q3', 'max', 'sem'])
# display(stat)
print(stat.to_string())

# Alternative of pandas: x['energy'].describe()

```

	device	browser	mean	std	min	q1	median	q3	max	sem
0	SM-G991B	chrome	5.67	1.25	4.16	4.49	5.47	6.46	10.14	0.05
1	SM-G991B	firefox	6.32	2.43	4.22	4.63	5.96	6.62	16.11	0.10

2	Nexus 5	chrome	9.34	3.16	5.67	7.23	8.24	10.02	22.29	0.13
3	Nexus 5	firefox	9.91	3.19	5.79	7.63	9.04	10.99	19.78	0.13

6.12 Descriptive Statistics Difference (By Device)

```
[57]: data = []

for device in devices:
    for browserpair in browserpairs:
        browser1 = stat[(stat['browser'] == browserpair[0]) & (stat['device'] == device)]
        browser2 = stat[(stat['browser'] == browserpair[1]) & (stat['device'] == device)]

        mean_diff = browser1.iloc[0]['mean']-browser2.iloc[0]['mean']
        median_diff = browser1.iloc[0]['median']-browser2.iloc[0]['median']
        min_diff = browser1.iloc[0]['min']-browser2.iloc[0]['min']
        max_diff = browser1.iloc[0]['max']-browser2.iloc[0]['max']
        std_diff = browser1.iloc[0]['std']-browser2.iloc[0]['std']
        sem_diff = browser1.iloc[0]['sem']-browser2.iloc[0]['sem']
        q1_diff = browser1.iloc[0]['q1']-browser2.iloc[0]['q1']
        q3_diff = browser1.iloc[0]['q3']-browser2.iloc[0]['q3']

        data.append([
            device,
            browserpair[0] + ' vs. ' + browserpair[1],
            numpy.round(mean_diff, 2),
            numpy.round(mean_diff/browser2.iloc[0]['mean']*100, 2),
            numpy.round(median_diff, 2),
            numpy.round(median_diff/browser2.iloc[0]['median']*100, 2),
            numpy.round(min_diff, 2),
            numpy.round(min_diff/browser2.iloc[0]['min']*100, 2),
            numpy.round(max_diff, 2),
            numpy.round(max_diff/browser2.iloc[0]['max']*100, 2),
            numpy.round(std_diff, 2),
            numpy.round(std_diff/browser2.iloc[0]['std']*100, 2),
            numpy.round(sem_diff, 2),
            numpy.round(sem_diff/browser2.iloc[0]['sem']*100, 2),
            numpy.round(q1_diff, 2),
            numpy.round(q1_diff/browser2.iloc[0]['q1']*100, 2),
            numpy.round(q3_diff, 2),
            numpy.round(q3_diff/browser2.iloc[0]['q3']*100, 2),
        ])

# Create the pandas DataFrame
```

```

ut = pd.DataFrame(data, columns = ['device',
    ↪ 'rq', 'mean_diff', 'mean_diff%', 'median_diff', 'median_diff%', 'min_diff', 'min_diff%', 'max_diff', 'max_diff%'])
display(ut)
#print(ut.to_string())

```

	device		rq	mean_diff	mean_diff%	median_diff	\
0	SM-G991B	chrome vs. firefox		-0.65	-10.28	-0.49	
1	Nexus 5	chrome vs. firefox		-0.57	-5.75	-0.80	

	median_diff%	min_diff	min_diff%	max_diff	max_diff%	std_diff	\
0	-8.22	-0.06	-1.42	-5.97	-37.06	-1.18	
1	-8.85	-0.12	-2.07	2.51	12.69	-0.03	

	std_diff%	sem_diff	sem_diff%	q1_diff	q1_diff%	q3_diff	q3_diff%
0	-48.56	-0.05	-50.0	-0.14	-3.02	-0.16	-2.42
1	-0.94	0.00	0.0	-0.40	-5.24	-0.97	-8.83

[]: