

WORKING WITH NANOPARTICLE TOXICITY DATASET IN PYTHON

PROGRAMMING FOR
CHEMISTS

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OUTLINE

GitHub

Process of work


Attempts

Program

Conclusions

GITHUB
WE WON IT.

BUT NOT FOR THE FIRST ATTEMPT

 Belvglin

Создано в Colab

a3b4da9 · 1 minute ago

🕒 18 Commits

📁 Nanoparticle-Toxicity-Dataset	alo	2 weeks ago
📄 Kirill_Nanopart_tox_dataset.ipynb	alo	51 minutes ago
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📖

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PROCESS OF WORK

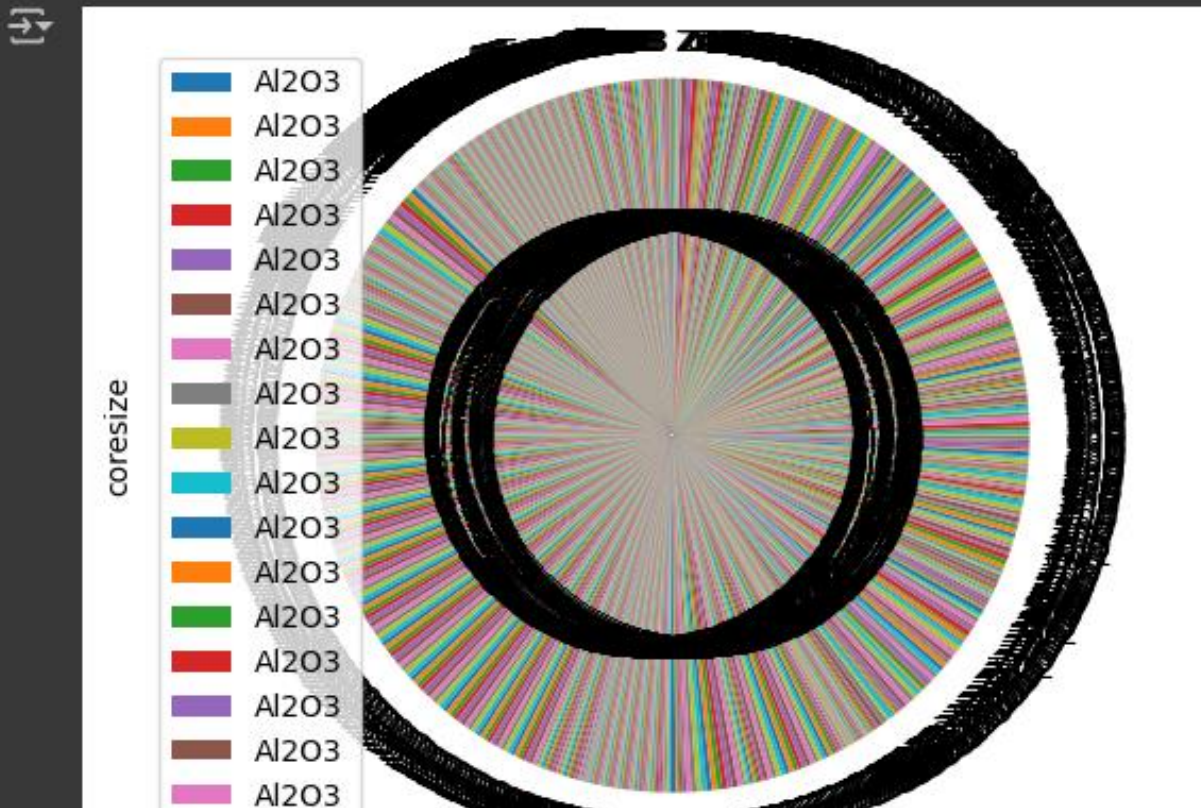
- Choosing a database
- Fighting with GitHub and GitBash: discovered tokens
- Structuring the Dataset
- Spending a lot of hours on finding the best relations of parameters and matching them to the conscious result
- Looking through a lot of courses and papers
- Finally got it.

BUT NOT WITHOUT FAILS

YES, WE ARE CHEMIST, BE TOLERANT

WE GOT THIS...

```
import pandas as pd
cvsData = pd.read_csv("nanotox_dataset.csv")
cvsData.plot.pie(y='coresize', labels=cvsData['NPs'], autopct='%1.1f%%', startangle=90)
plt.axis('equal')
plt.show()
```

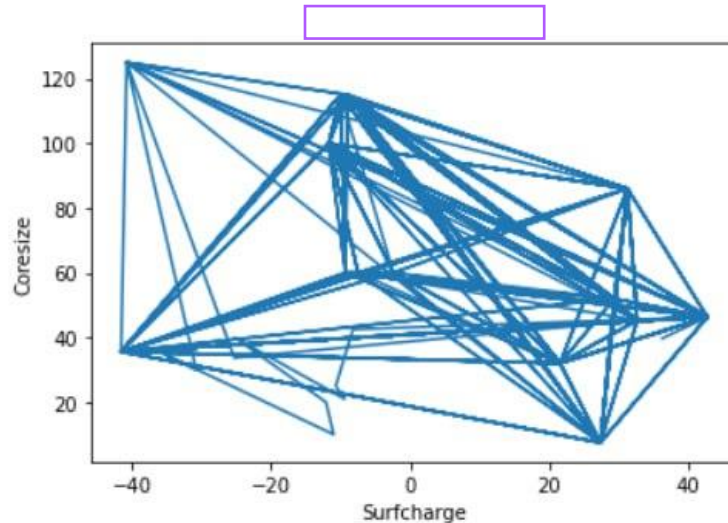



AND THIS...

```
Ввод [3]: import matplotlib.pyplot as plt
import numpy as np
import csv

X = []
Y = []

with open('nanotox_dataset.csv', 'r', encoding='utf-8') as datafile:
    plotting = csv.reader(datafile, delimiter=',')
    for ROWS in plotting:
        X.append(float(ROWS[3]))
        Y.append(float(ROWS[1]))
plt.plot(X, Y)
plt.title( )
plt.xlabel('Surfcharge')
plt.ylabel('Coresize')
plt.show()
```



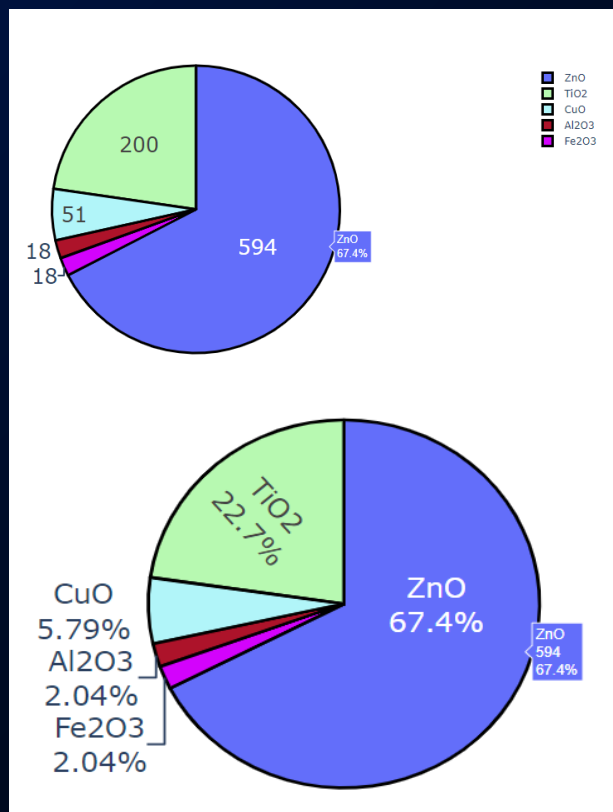


FIRST OF ALL WE FIND OUT ALL RELATIONS WE COULD

It was challenging due to:

- Experimental type of data
- Abundance of source data and only one column with results
- In addition this one parameter was uncountable and has data like 'yes/no'

WE STRUCTURED OUR DATASET TO GET RID OF PARAMETERS WE DON'T NEED



```
# Sorting by column 'Name of column'
cvsData.sort_values(by=['hydrosize','class'],ascending=[False, False])
cvsData.drop("Ec", axis = 1) # Удалить переменную из фрейма данных
#Параметр «axis» в Pandas определяет направление выполнения операции:
#Движение вниз по строкам (по умолчанию) соответствует axis=0, что эквивалентно операциям агрегации
#Движение вбок по столбцам соответствует axis=1, что эквивалентно операциям агрегации данных по
```

	NPs	coresize	hydrosize	surfcharge	surfarea	Expotime	dosage	e	NOxygen	class
0	Al2O3	39.7	267.0	36.3	64.7	24	0.001	1.61	3	nonToxic
1	Al2O3	39.7	267.0	36.3	64.7	24	0.010	1.61	3	nonToxic
2	Al2O3	39.7	267.0	36.3	64.7	24	0.100	1.61	3	nonToxic
3	Al2O3	39.7	267.0	36.3	64.7	24	1.000	1.61	3	nonToxic
4	Al2O3	39.7	267.0	36.3	64.7	24	5.000	1.61	3	nonToxic
...
876	ZnO	45.3	310.0	32.7	21.3	24	20.000	1.65	1	Toxic
877	ZnO	32.0	1093.0	21.6	37.0	24	25.000	1.65	1	Toxic
878	ZnO	46.3	239.0	42.8	24.1	12	100.000	1.90	1	Toxic

```
import plotly as py
import plotly.graph_objs as go
import numpy as np
groups = ['Al2O3', 'Fe2O3', 'TiO2', 'CuO', 'ZnO']
amount = [18, 18, 200, 51, 594]
colors = ['#ad132a', '#d303fc', '#b7f9b1', '#b1f5f9']

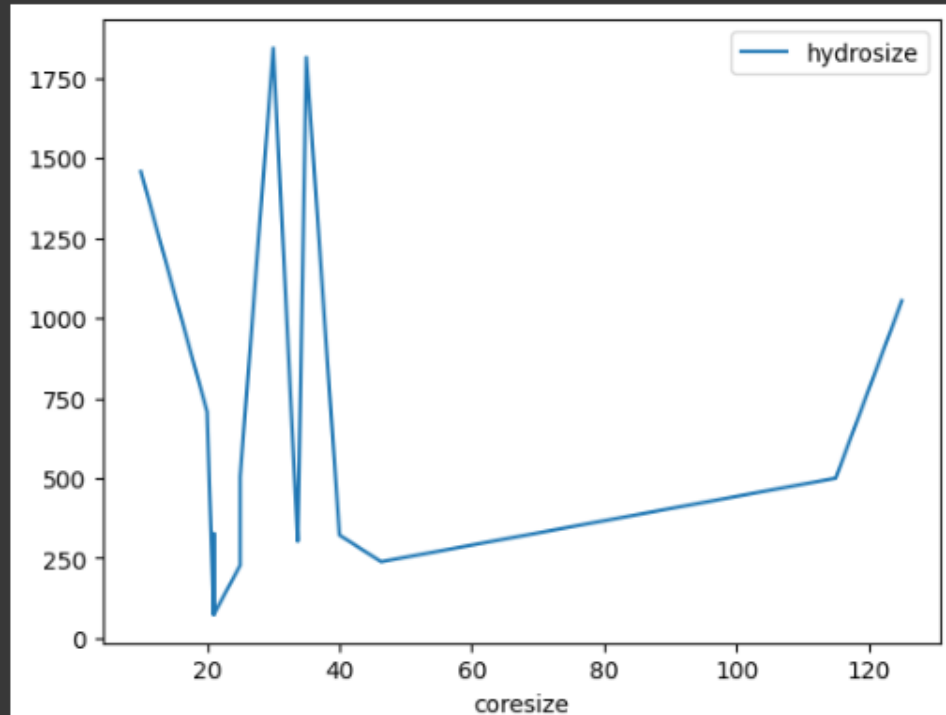
trace = go.Pie(labels=groups, values=amount,
                hoverinfo='label+percent', textinfo='value',
                textfont=dict(size=25),
                marker=dict(colors=colors,
                            line=dict(color='#000000', width=3)))

from plotly.offline import iplot
iplot([trace])
```

SO AT FIRST WE JUST DRAW SOME GRAPHS

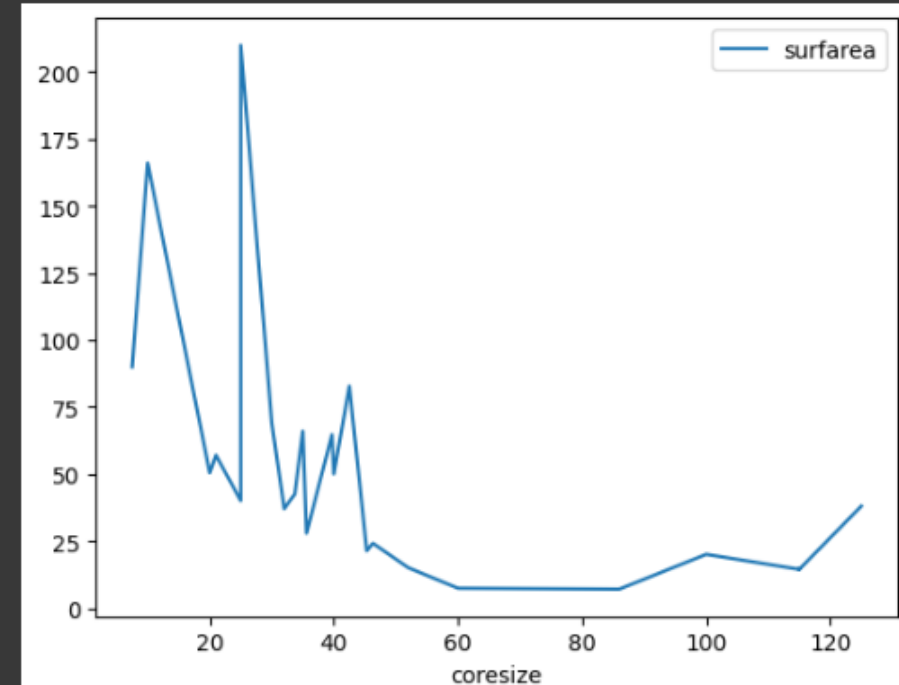
```
data.sort_values(data.columns[1],inplace=True)  
filtr_data = data[data['NPs'] == 'TiO2']  
filtr_data.plot(x='coresize', y='hydrosized')
```

<Axes: xlabel='coresize'>



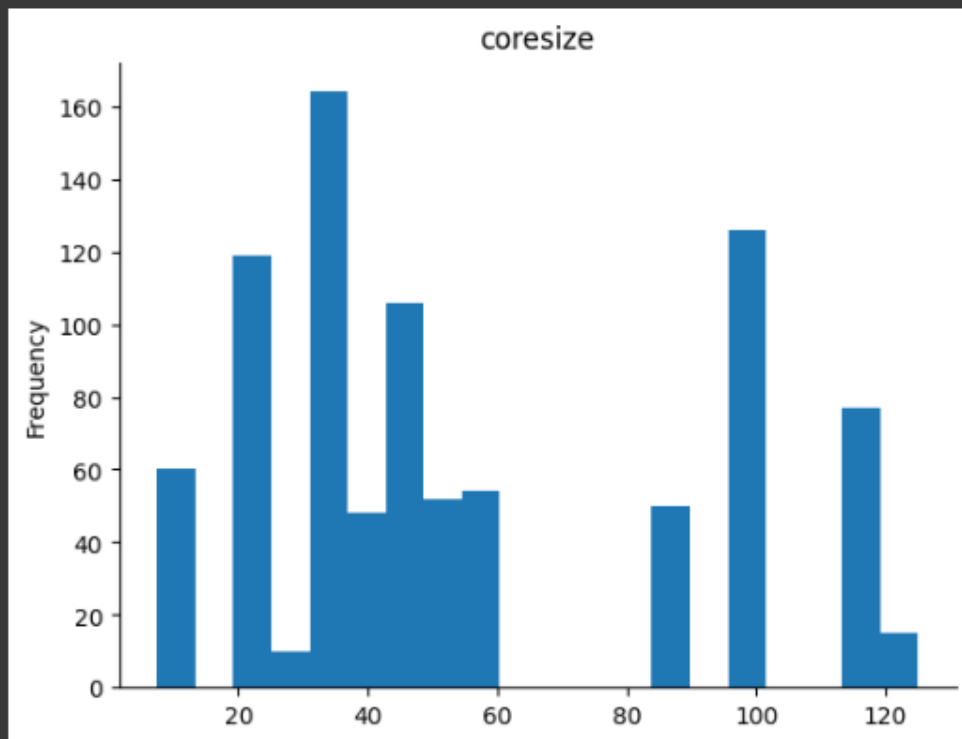
```
import pandas as pd  
cvsData = pd.read_csv("nanotox_dataset.csv")  
cvsData.sort_values(cvsData.columns[1],inplace=True)  
cvsData.plot(x='coresize',y='surfarea')
```

<Axes: xlabel='coresize'>



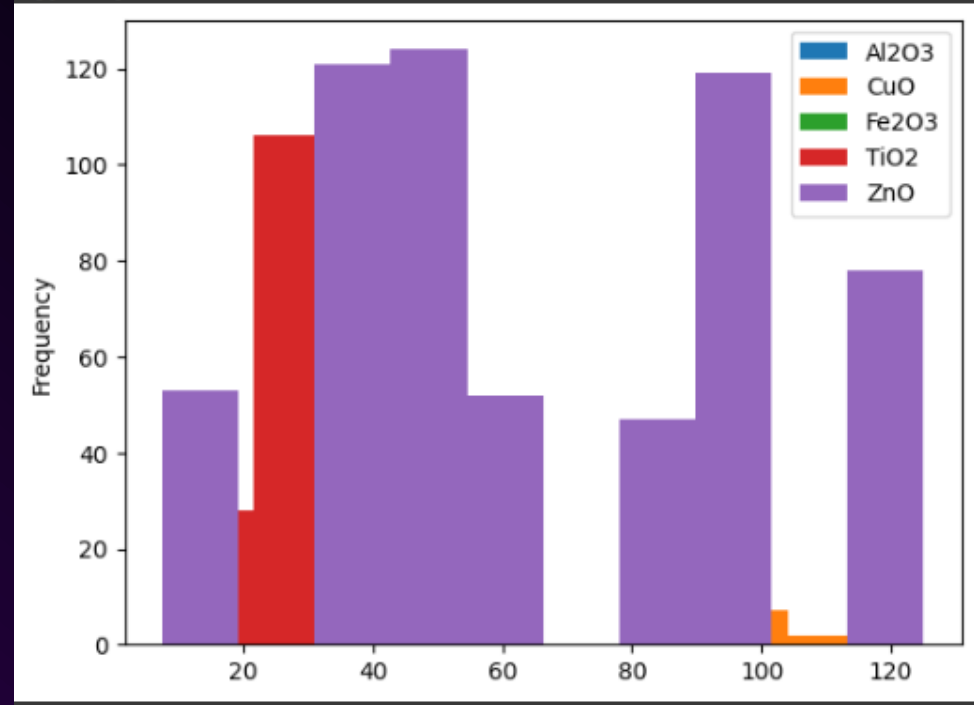
AND MORE

```
from matplotlib import pyplot as plt
df_12['coresize'].plot(kind='hist', bins=20, title='coresize')
plt.gca().spines[['top', 'right',]].set_visible(False)
```

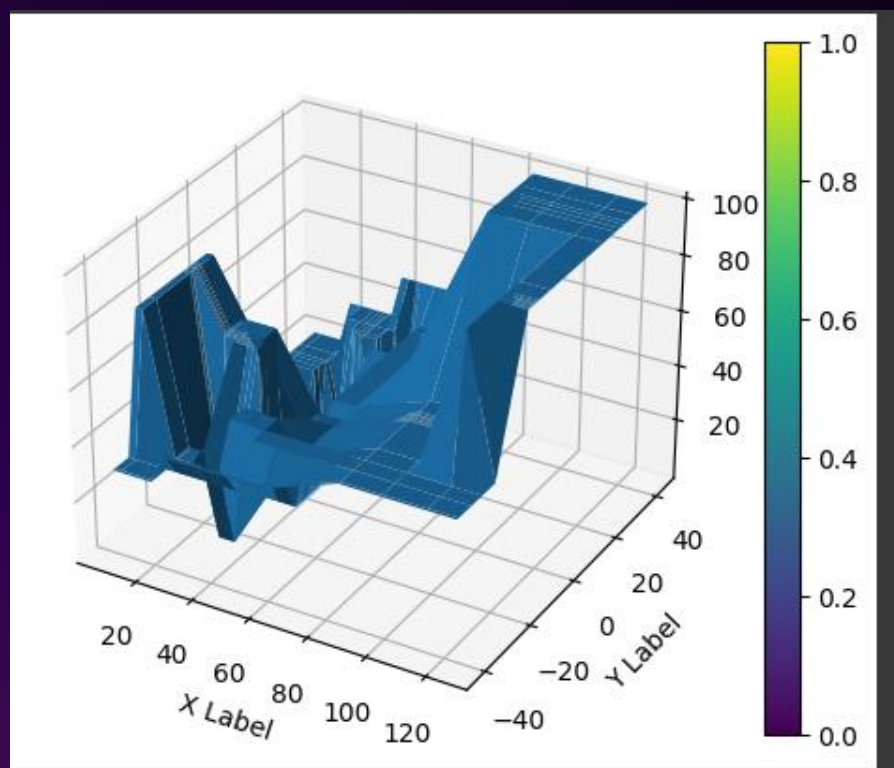


```
df.groupby('NPs')['coresize'].plot(kind='hist', legend = True)
```

dtype: object



BUT WE UPGRADE OUR SKILL AND GOT THIS



```
x = data['coresize'].values
y = data['surfcharge'].values
z = data['dosage'].values

# Создание сетки
X = np.unique(xsurf)
Y = np.unique(ysurf)
X, Y = np.meshgrid(X, Y)

# Интерполяция Z значений
Z = griddata((xsurf, ysurf), z, (X, Y), method='nearest')

# Создание 3D графика
fig = plt.figure()
ax = fig.add_subplot(111, projection='3d')

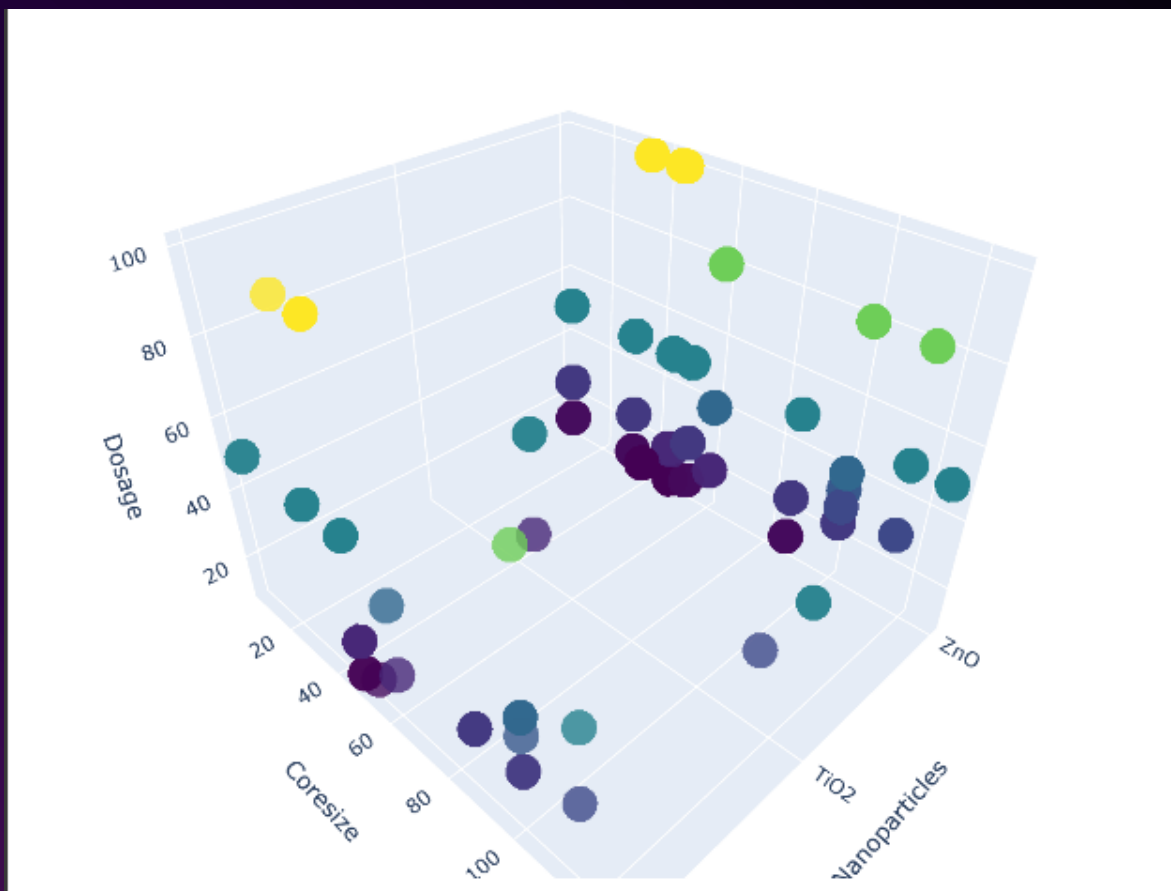
# Построение поверхности
surf = ax.plot_surface(X, Y, Z)

# Добавление цветовой шкалы
fig.colorbar(surf)

# Добавление подписей
ax.set_xlabel('X Label')
ax.set_ylabel('Y Label')
ax.set_zlabel('Z Label')

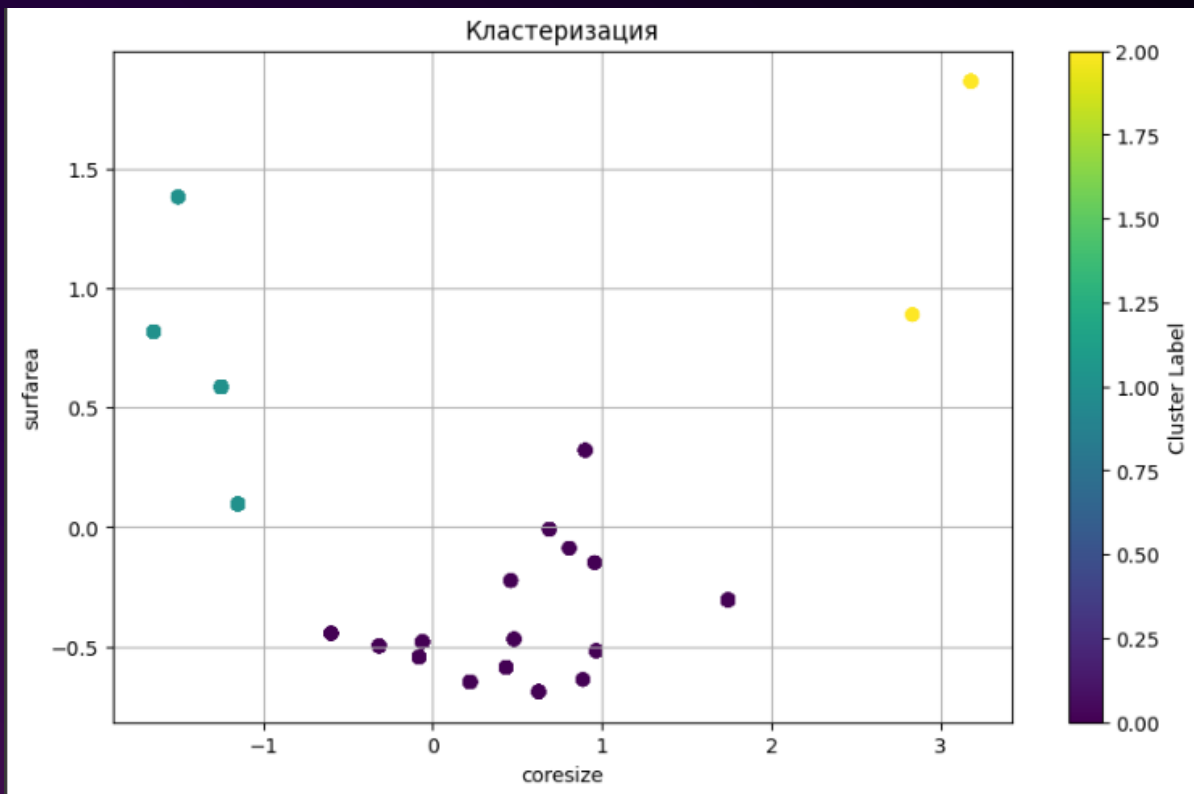
# Отображение графика
plt.show()
```

AND ALSO THIS



```
#Это 3д модель распределения токсичных частиц по размеру и до  
x = df3['NPs']  
y = df3['coresize']  
z = df3['dosage']  
  
trace2 = go.Scatter3d(  
    x=x,  
    y=y,  
    z=z,  
    mode='markers',  
    marker=dict(  
        size=12,  
        color=z,  
        colorscale='Viridis',  
        opacity = 0.8))  
  
data2 = [trace2]  
fig2 = go.Figure(data=data2, layout=layout)  
fig2.update_layout(scene = dict(  
    xaxis_title='Nanoparticles',  
    yaxis_title='Coresize',  
    zaxis_title='Dosage'),  
    width=700,  
    margin=dict(r=20, b=10, l=10, t=10))  
iplot(fig2, filename='3d-plot')
```


WE ALSO MADE A FEW MORE THINGS



```
# Шаг 1: Загрузка данных
df = pd.DataFrame(data, columns=['coresize', 'surfarea'])

# Шаг 2: Предобработка данных
# Удаляем строки с пропущенными значениями и выбираем только числовые колонки
data_cleaned = df.dropna()
X = data_cleaned.select_dtypes(include=[np.number])

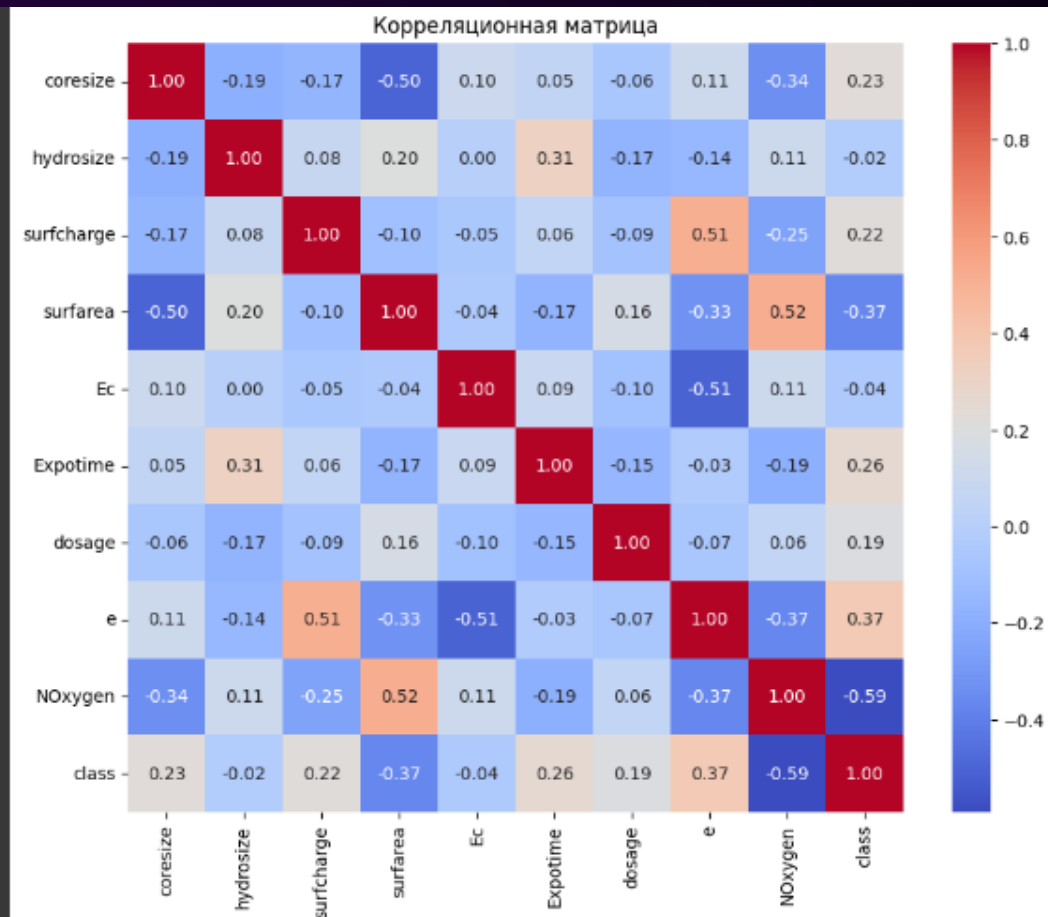
# Нормализация данных
scaler = StandardScaler()
X_scaled = scaler.fit_transform(X)

# Шаг 3: Кластеризация K-means
kmeans = KMeans(n_clusters=3, random_state=10) # Выберите количество кластеров
kmeans.fit(X_scaled)
labels = kmeans.labels_

# Шаг 4: Визуализация результатов
# Снижение размерности до 2D для визуализации
pca = PCA(n_components=2)
X_pca = pca.fit_transform(X_scaled)

plt.figure(figsize=(10, 6))
plt.scatter(X_pca[:, 0], X_pca[:, 1], c=labels, cmap='viridis', marker='o')
plt.title('Кластеризация')
plt.xlabel('coresize')
plt.ylabel('surfarea')
plt.colorbar(label='Cluster Label')
plt.grid()
plt.show()
```

WE ALSO MADE A FEW MORE THINGS

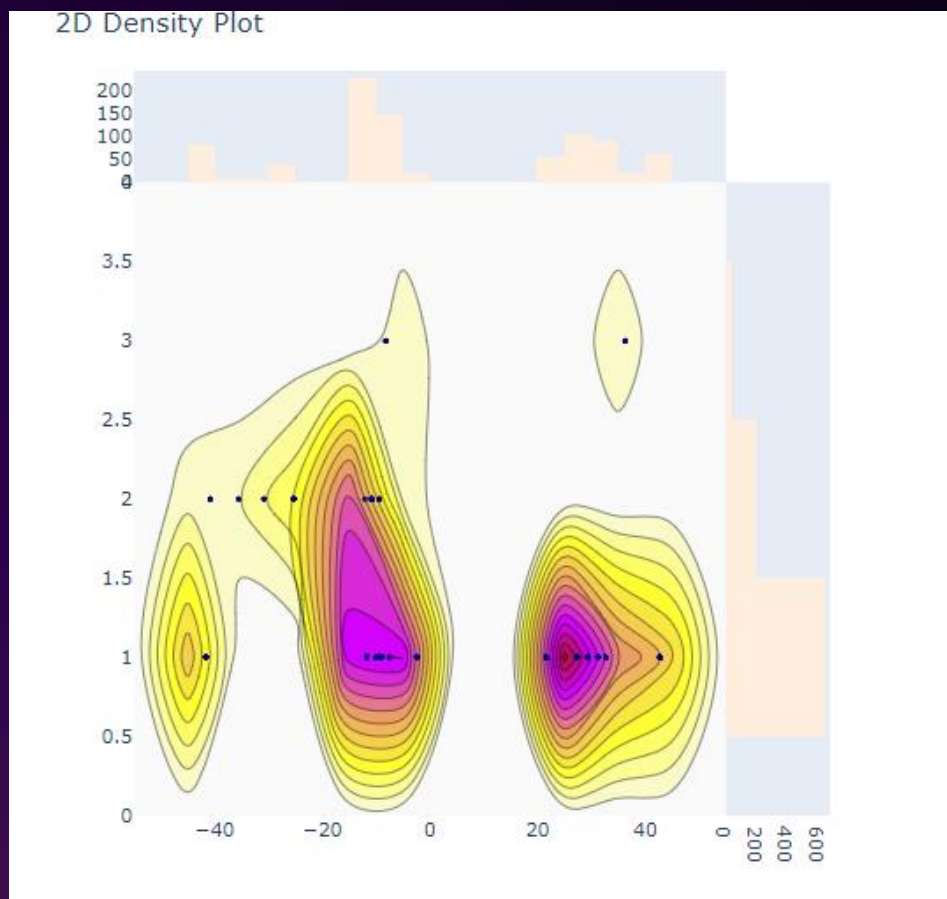


```
# Шаг 2: Извлечение только числовых данных
numeric_data = data.select_dtypes(include=['number'])

# Шаг 3: Построение корреляционной матрицы
correlation_matrix = numeric_data.corr()

# Шаг 4: Визуализация корреляционной матрицы
plt.figure(figsize=(10, 8)) # Задаем размер графика
sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm', fmt='.2f',
            xticklabels=correlation_matrix.columns,
            yticklabels=correlation_matrix.index)
plt.title('Корреляционная матрица')
plt.show()
```

WHAT ABOUT MORE SPECIFIC INFORMATION?

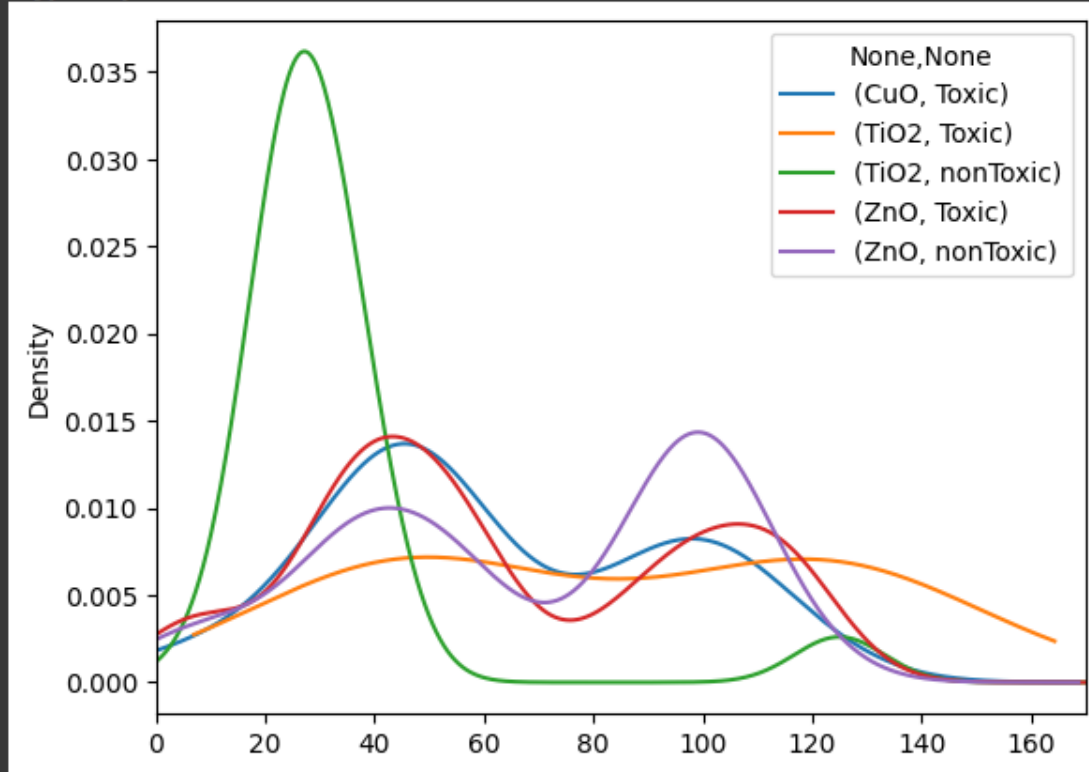


```
#здесь представлена плотность распределения по заряду поверхности и количеству атомов кислорода
import plotly.figure_factory as ff
x = df['surfcharge']
y = df['NOxygen']
colorscale = ['#ad132a', '#d303fc', 'rgb(236,158,105)', (1,1,0.2), (0.98,0.98,0.98)]
fig1 = ff.create_2d_density(
    x,y,colorscale=colorscale,
    hist_color='rgb(255,237,222)', point_size=3)

iplot(fig1, filename='histogram_subplots')
```

TOXICITY

dtype: object



- `df2.groupby(['NPs', 'class'])['coresize'].plot(kind='kde',
xlim=[0,170], legend = True)`

THANKS FOR
YOUR ATTENTION

The background features a gradient from deep purple on the left to dark blue on the right. On the right side, there are several concentric circles of varying thicknesses and colors (white, light blue, dark blue). Some circles are partially visible, creating a sense of depth and motion. There are also small white dots scattered across the background.

(We really did our best)