

```
In [1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import os
```

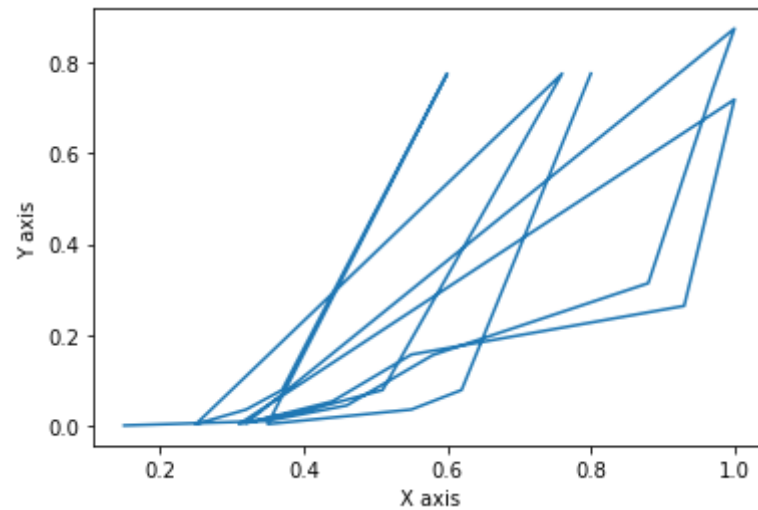
```
In [2]: data1 = pd.read_csv('D:/class/Project/ML/sensitivity analysis/dataset-1.csv')
data2 = pd.read_csv('D:/class/Project/ML/sensitivity analysis/ref dataset.csv')
```

```
In [3]: data1.columns
```

```
Out[3]: Index(['coat', 'time', 'in-vitro', 'in-vivo'], dtype='object')
```

```
In [4]: #graph for dataset 1
```

```
In [5]: import matplotlib.pyplot as plt
x = data1['in-vitro']
y = data1['in-vivo']
plt.plot(x,y)
plt.xlabel('X axis')
plt.ylabel('Y axis')
plt.show()
```



```
In [6]: import pandas as pd
import matplotlib.pyplot as plt

# create a dataframe with x, y, and coat columns
x = data1['in-vitro']
y = data1['in-vivo']
coat1 = data1['coat']
data_dict = {'x': x, 'y': y, 'coat': coat1}
data = pd.DataFrame(data_dict)
print(data)
# group the data by coat type
grouped_data = data.groupby('coat')

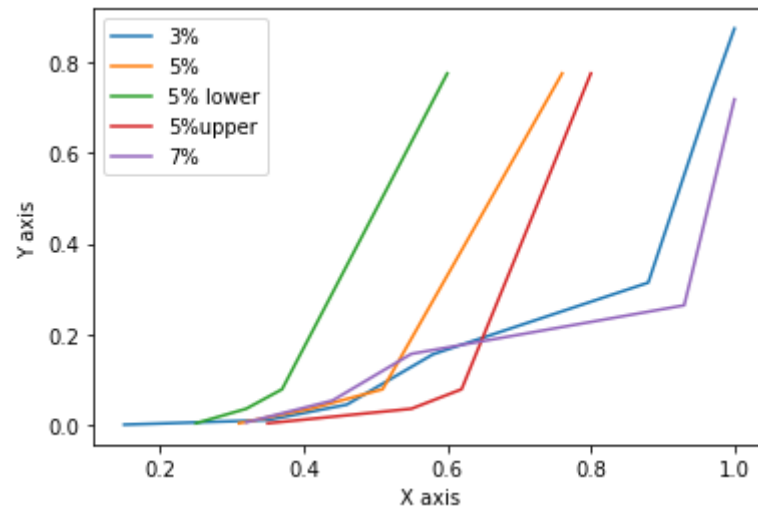
# plot each group separately
for coat_type, group in grouped_data:
    plt.plot(group['x'], group['y'], label=coat_type)

#set x and y labels
plt.xlabel('X axis')
plt.ylabel('Y axis')

# add a Legend
plt.legend()

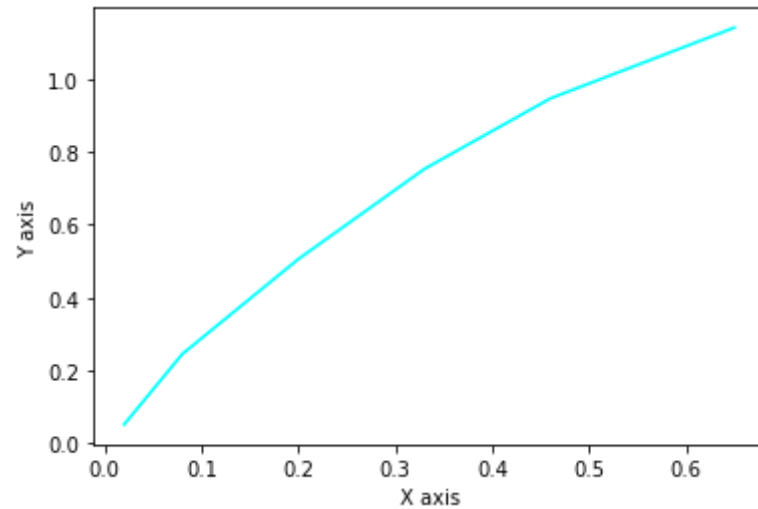
# show the graph
plt.show()
```

	x	y	coat
0	0.15	0.001	3%
1	0.35	0.011	3%
2	0.46	0.045	3%
3	0.58	0.156	3%
4	0.88	0.314	3%
5	0.97	0.741	3%
6	1.00	0.874	3%
7	0.32	0.006	7%
8	0.44	0.054	7%
9	0.55	0.157	7%
10	0.93	0.264	7%
11	1.00	0.718	7%
12	0.31	0.004	5%
13	0.41	0.036	5%
14	0.51	0.079	5%
15	0.76	0.775	5%
16	0.25	0.004	5% lower
17	0.32	0.036	5% lower
18	0.37	0.079	5% lower
19	0.60	0.775	5% lower
20	0.35	0.004	5%upper
21	0.55	0.036	5%upper
22	0.62	0.079	5%upper
23	0.80	0.775	5%upper



In [7]: *#graph for dataset-2*

```
In [8]: import matplotlib.pyplot as plt
x = data2['in-vitro']
y = data2['in-vivo']
plt.plot(x,y, color='cyan')
plt.xlabel('X axis')
plt.ylabel('Y axis')
plt.show()
```



```
In [9]: #sigmoid graph for dataset-1 using in-vitro
```

```
In [10]: import numpy as np
import matplotlib.pyplot as plt

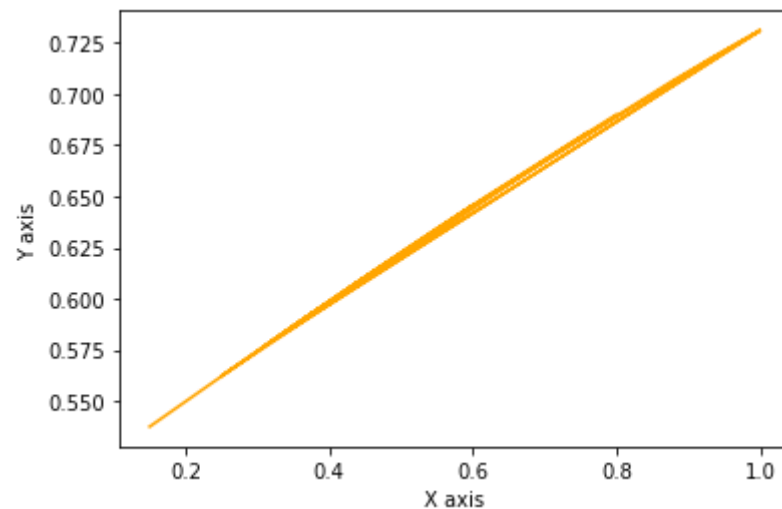
# define the sigmoid function
x = data1['in-vitro']
def sigmoid_func(x, a, b):
    return 1 / (1 + np.exp(-(x-a)/b))

# calculate y values using the sigmoid function
y = sigmoid_func(x, 0, 1)

# plot the graph
plt.plot(x, y, color = "orange")

# set x and y labels
plt.xlabel('X axis')
plt.ylabel('Y axis')

# show the graph
plt.show()
```



```
In [11]: #sigmoid graph for dataset-1 using in-vitro
```

```
In [12]: import numpy as np
import matplotlib.pyplot as plt

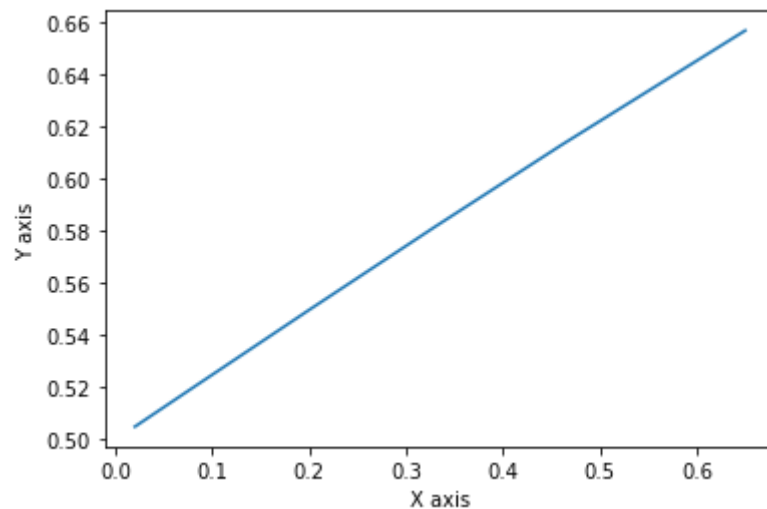
# define the sigmoid function
x = data2['in-vitro']
def sigmoid_func(x, a, b):
    return 1 / (1 + np.exp(-(x-a)/b))

# calculate y values using the sigmoid function
y = sigmoid_func(x, 0, 1)

# plot the graph
plt.plot(x, y)

# set x and y labels
plt.xlabel('X axis')
plt.ylabel('Y axis')

# show the graph
plt.show()
```





```
In [13]: #exponential graph for dataset-1 using x
```

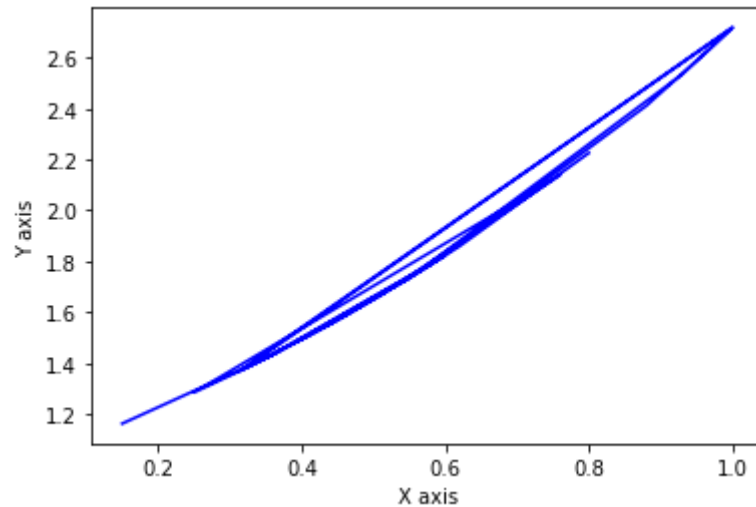
```
In [14]: x = data1['in-vitro']
def exp_func(x):
    return np.exp(x)

# calculate y values using the exponential function
y = exp_func(x)

# plot the graph
plt.plot(x, y, color = "blue")

# set x and y labels
plt.xlabel('X axis')
plt.ylabel('Y axis')

# show the graph
plt.show()
```



```
In [15]: #exponential graph for dataset-2 using x
```

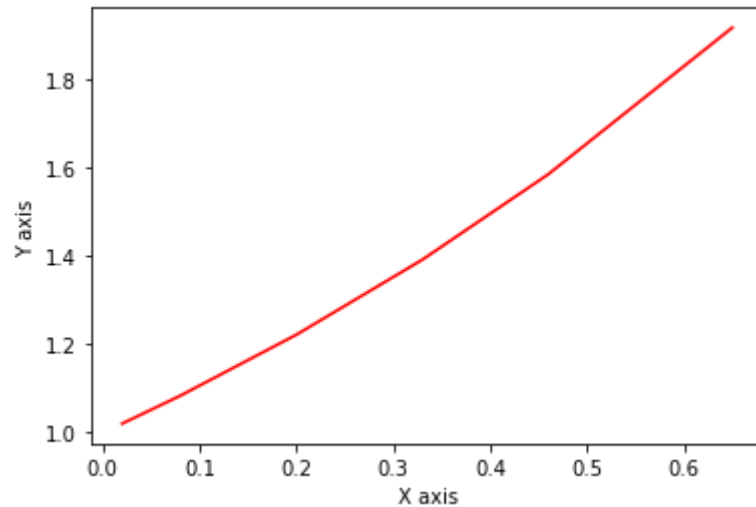
```
In [16]: x = data2['in-vitro']
def exp_func(x):
    return np.exp(x)

# calculate y values using the exponential function
y = exp_func(x)

# plot the graph
plt.plot(x, y, color = "red")

# set x and y labels
plt.xlabel('X axis')
plt.ylabel('Y axis')

# show the graph
plt.show()
```



```
In [17]: #for Data-1 training and splitting
```

```
In [18]: from sklearn.model_selection import train_test_split
X = data1[['in-vitro']]
y = data1[['in-vivo']]
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
print(X_train.shape, X_test.shape, y_train.shape, y_test.shape)
```

C:\Users\HP\anaconda3\lib\site-packages\scipy\\_\_init\_\_.py:146: UserWarning: A NumPy version  $\geq 1.16.5$  and  $< 1.23.0$  is required for this version of SciPy (detected version 1.24.2

warnings.warn(f"A NumPy version  $\geq \{np\_minversion\}$  and  $< \{np\_maxversion\}$ ")

(19, 1) (5, 1) (19, 1) (5, 1)

```
In [19]: #testing
```

```
In [20]: from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_squared_error, r2_score
lin_reg = LinearRegression()
lin_reg.fit(X_train, y_train)
y_pred = lin_reg.predict(X_test)
mse = mean_squared_error(y_test, y_pred)
print("Mean Squared Error:", mse)
r2 = r2_score(y_test, y_pred)
print("R-squared:", r2)
```

Mean Squared Error: 0.011853761559354989

R-squared: 0.8432667396266914

```
In [ ]: #random forest for dataset-1
```

```
In [21]: from sklearn.ensemble import RandomForestRegressor
from sklearn.model_selection import GridSearchCV
rf_params = {"max_depth":[100,200,300],
             "max_features":[1,2,3],
             "n_estimators":[100,150,200],
             "min_samples_split":[3,4,5]}
rf_model = RandomForestRegressor(random_state = 24)
gs = GridSearchCV(rf_model,rf_params,cv=5,n_jobs=-1,verbose=2,scoring='roc_auc')
#with np.errstate(divide='ignore', invalid='ignore'):
gs.fit(X_train,y_train.values.ravel())
print(str(gs.best_params_))
```

Fitting 5 folds for each of 81 candidates, totalling 405 fits

```
{'max_depth': 100, 'max_features': 1, 'min_samples_split': 3, 'n_estimators': 100}
```

C:\Users\HP\anaconda3\lib\site-packages\sklearn\model\_selection\\_validation.py:372: FitFailedWarning:  
 270 fits failed out of a total of 405.  
 The score on these train-test partitions for these parameters will be set to nan.  
 If these failures are not expected, you can try to debug them by setting error\_score='raise'.

Below are more details about the failures:

-----  
 270 fits failed with the following error:

Traceback (most recent call last):

```
File "C:\Users\HP\anaconda3\lib\site-packages\sklearn\model_selection\_validation.py", line 680, in _fit_and_score
    estimator.fit(X_train, y_train, **fit_params)
File "C:\Users\HP\anaconda3\lib\site-packages\sklearn\ensemble\_forest.py", line 450, in fit
    trees = Parallel(
File "C:\Users\HP\anaconda3\lib\site-packages\joblib\parallel.py", line 1043, in __call__
    if self.dispatch_one_batch(iterator):
File "C:\Users\HP\anaconda3\lib\site-packages\joblib\parallel.py", line 861, in dispatch_one_batch
    self._dispatch(tasks)
File "C:\Users\HP\anaconda3\lib\site-packages\joblib\parallel.py", line 779, in _dispatch
    job = self._backend.apply_async(batch, callback=cb)
File "C:\Users\HP\anaconda3\lib\site-packages\joblib\_parallel_backends.py", line 208, in apply_async
    result = ImmediateResult(func)
File "C:\Users\HP\anaconda3\lib\site-packages\joblib\_parallel_backends.py", line 572, in __init__
    self.results = batch()
File "C:\Users\HP\anaconda3\lib\site-packages\joblib\parallel.py", line 262, in __call__
    return [func(*args, **kwargs)
File "C:\Users\HP\anaconda3\lib\site-packages\joblib\parallel.py", line 262, in <listcomp>
    return [func(*args, **kwargs)
File "C:\Users\HP\anaconda3\lib\site-packages\sklearn\utils\fixes.py", line 216, in __call__
    return self.function(*args, **kwargs)
File "C:\Users\HP\anaconda3\lib\site-packages\sklearn\ensemble\_forest.py", line 185, in _parallel_build_trees
    tree.fit(X, y, sample_weight=curr_sample_weight, check_input=False)
File "C:\Users\HP\anaconda3\lib\site-packages\sklearn\tree\_classes.py", line 1315, in fit
    super().fit(
File "C:\Users\HP\anaconda3\lib\site-packages\sklearn\tree\_classes.py", line 308, in fit
    raise ValueError("max_features must be in (0, n_features]")
```

ValueError: max\_features must be in (0, n\_features]

warnings.warn(some\_fits\_failed\_message, FitFailedWarning)

C:\Users\HP\anaconda3\lib\site-packages\sklearn\model\_selection\\_search.py:969: UserWarning: One or more of the test  
 scores are non-finite: [nan nan nan nan nan nan nan nan nan nan nan nan nan nan nan nan nan  
 nan nan nan nan nan nan nan nan nan nan nan nan nan nan nan nan nan

```
nan nan nan nan nan nan nan nan nan nan nan nan nan nan nan nan nan
nan nan nan nan nan nan nan nan nan nan nan nan nan nan nan nan nan
nan nan nan nan nan nan nan nan nan]
warnings.warn(
```

```
In [22]: rf_tune = RandomForestRegressor(max_depth= 100, max_features= 1, min_samples_split= 3, n_estimators= 100,random_state
rf_tune.fit(X_train,y_train.values.ravel())
y_pred = rf_tune.predict(X_test)
mse = mean_squared_error(y_test, y_pred)
print("Mean Squared Error:", mse)
r2 = r2_score(y_test, y_pred)
print("R-squared:", r2)
```

```
Mean Squared Error: 0.0009580374568277777
R-squared: 0.9873326004225328
```

```
In [24]: #for data 2 training and splitting
```

```
In [50]: from sklearn.model_selection import train_test_split
s = data2[['in-vitro']]
t = data2[['in-vivo']]
print(s,t)
s_train, s_test, t_train, t_test = train_test_split(s, t, test_size=0.2, random_state=0)
print(s_train.shape, s_test.shape, t_train.shape, t_test.shape)
```

```
in-vitro
0      0.02
1      0.08
2      0.20
3      0.33
4      0.46
5      0.65      in-vivo
0 0.052223
1 0.245313
2 0.506252
3 0.753220
4 0.946752
5 1.141000
(4, 1) (2, 1) (4, 1) (2, 1)
```

```
In [51]: #testing for refrence dataset
```

```
In [52]: from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_squared_error, r2_score
lin_reg = LinearRegression()
lin_reg.fit(s_train, t_train)
t_pred = lin_reg.predict(s_test)
mse = mean_squared_error(t_test, t_pred)
print("Mean Squared Error:", mse)
r2 = r2_score(t_test, t_pred)
print("R-squared:", r2)
```

Mean Squared Error: 0.02537232308258872

R-squared: 0.748106162966848

```
In [30]: #random forest for Ref dataset
```

```
In [40]: from sklearn.ensemble import RandomForestRegressor
from sklearn.model_selection import GridSearchCV
rf_params = {"max_depth":[100,200,300],
             "max_features":[1,2,3],
             "n_estimators":[100,150,200],
             "min_samples_split":[2,3,4]}
rf_model = RandomForestRegressor(random_state = 24)
gs = GridSearchCV(rf_model,rf_params,cv=3,n_jobs=-1,verbose=2,scoring='roc_auc')
#with np.errstate(divide='ignore', invalid='ignore'):
gs.fit(s_train,t_train)
print(str(gs.best_params_))
```

Fitting 3 folds for each of 81 candidates, totalling 243 fits

```
{'max_depth': 100, 'max_features': 1, 'min_samples_split': 2, 'n_estimators': 100}
```





```
nan nan nan nan nan nan nan nan nan nan nan nan nan nan nan nan nan
nan nan nan nan nan nan nan nan nan nan nan nan nan nan nan nan nan
nan nan nan nan nan nan nan nan]
```

```
warnings.warn(
```

```
C:\Users\HP\anaconda3\lib\site-packages\sklearn\model_selection\_search.py:926: DataConversionWarning: A column-vect
or y was passed when a 1d array was expected. Please change the shape of y to (n_samples,), for example using ravel
().
```

```
self.best_estimator_.fit(X, y, **fit_params)
```

```
In [41]: rf_tune = RandomForestRegressor(max_depth= 100, max_features= 1, min_samples_split= 2, n_estimators= 100,random_state
rf_tune.fit(X_train,y_train.values.ravel())
y_pred = rf_tune.predict(X_test)
mse = mean_squared_error(y_test, y_pred)
print("Mean Squared Error:", mse)
r2 = r2_score(y_test, y_pred)
print("R-squared:", r2)
```

```
Mean Squared Error: 0.0016449261580888871
```

```
R-squared: 0.978250394312416
```

```
In [28]: #slope dataset-1
```

```
In [62]: x = data1['in-vitro']
y = data1['in-vivo']
slope, intercept = np.polyfit(x, y, 1)

# Print the slope
print("Slope:", slope)

# Calculate the residuals
residuals = y - (slope * x + intercept)
std_residuals = np.std(residuals)
mean_slope = np.mean(slope)
CV = (std_residuals / mean_slope) * 100

print("CV%:", CV)
```

```
Slope: 1.0036922433158615
```

```
CV%: 18.62990282217695
```

```
In [30]: slope = 1.0036922433158615
y = [(slope * xi) for xi in x]

# Print predicted y values
print(y)
```

```
[0.15055383649737922, 0.3512922851605515, 0.4616984319252963, 0.5821415011231996, 0.8832491741179581, 0.973581476016
3856, 1.0036922433158615, 0.3211815178610757, 0.44162458705897906, 0.5520307338237239, 0.9334337862837513, 1.0036922
433158615, 0.31114459542791706, 0.4115138197595032, 0.5118830440910894, 0.7628061049200547, 0.25092306082896537, 0.3
211815178610757, 0.37136613002686875, 0.6022153459895169, 0.3512922851605515, 0.5520307338237239, 0.622289190855834
1, 0.8029537946526892]
```

```
In [ ]: #slope for Dataset-2
```

```
In [34]: x = data2['in-vitro']
y = data2['in-vivo']
slope, intercept = np.polyfit(x, y, 1)

# Print the slope
print("Slope:", slope)
# Calculate the residuals
residuals = y - (slope * x + intercept)
std_residuals = np.std(residuals)
mean_slope = np.mean(slope)
CV = (std_residuals / mean_slope) * 100
# Print the CV%
print("CV%:", CV)
```

```
Slope: 1.720400073632538
CV%: 3.8690352042126133
```

```
In [32]: slope = 1.720400073632538  
y = [(slope * xi) for xi in x]  
  
# Print predicted y values  
print(y)
```

```
[0.03440800147265076, 0.13763200589060304, 0.3440800147265076, 0.5677320242987376, 0.7913840338709676, 1.1182600478611497]
```

```
In [ ]: #update the dataset and run program again
```

```
In [60]: data_update = pd.read_csv('D:/class/Project/ML/sensitivity analysis/updatedataset.csv')
```

```
In [61]: import pandas as pd
import matplotlib.pyplot as plt

# create a dataframe with x, y, and coat columns
x = data_update[['in-vitro']]
y = data_update[['in-vivo']]
time1 = data_update[['time']]
data = pd.DataFrame({'x': x, 'y': y, 'time': time1}, index=[0])
#data = pd.DataFrame(data_dict)
print(data)
# group the data by coat type
grouped_data = data.groupby('time')

# plot each group separately
for time_type, group in grouped_data:
    plt.plot(group['x'], group['y'], label=time_type)

#set x and y labels
plt.xlabel('X axis')
plt.ylabel('Y axis')

# add a Legend
plt.legend()

# show the graph
plt.show()
```

-----  
**ValueError**

Traceback (most recent call last)

Input **In [61]**, in <cell line: 8>()  
6 y = data\_update[['in-vivo']]

7 time1 = data\_update[['time']]

----&gt; 8 data = pd.DataFrame({'x': x, 'y': y, 'time': time1}, index=[0])

9 #data = pd.DataFrame(data\_dict)

10 print(data)

File ~\anaconda3\lib\site-packages\pandas\core\frame.py:636, in DataFrame.\_\_init\_\_(self, data, index, columns, dtype, copy)

630 mgr = self.\_init\_mgr(  
631 data, axes={"index": index, "columns": columns}, dtype=dtype, copy=copy  
632 )

634 elif isinstance(data, dict):

635 # GH#38939 de facto copy defaults to False only in non-dict cases

--&gt; 636 mgr = dict\_to\_mgr(data, index, columns, dtype=dtype, copy=copy, typ=manager)

637 elif isinstance(data, ma.MaskedArray):

638 import numpy.ma.mrecords as mrecords

File ~\anaconda3\lib\site-packages\pandas\core\internals\construction.py:502, in dict\_to\_mgr(data, index, columns, dtype, typ, copy)

494 arrays = [  
495 x496 if not hasattr(x, "dtype") or not isinstance(x.dtype, ExtensionDtype)  
497 else x.copy()  
498 for x in arrays

499 ]

500 # TODO: can we get rid of the dt64tz special case above?

--&gt; 502 return arrays\_to\_mgr(arrays, columns, index, dtype=dtype, typ=typ, consolidate=copy)

File ~\anaconda3\lib\site-packages\pandas\core\internals\construction.py:125, in arrays\_to\_mgr(arrays, columns, index, dtype, verify\_integrity, typ, consolidate)

122 index = ensure\_index(index)

124 # don't force copy because getting jammed in an ndarray anyway

--&gt; 125 arrays = \_homogenize(arrays, index, dtype)

126 # \_homogenize ensures

127 # - all(len(x) == len(index) for x in arrays)

128 # - all(x.ndim == 1 for x in arrays)

(...)

131

```

132 else:
133     index = ensure_index(index)

```

File ~\anaconda3\lib\site-packages\pandas\core\internals\construction.py:625, in \_homogenize(data, index, dtype)

```

622         val = dict(val)
623         val = lib.fast_multiget(val, oindex._values, default=np.nan)
--> 625     val = sanitize_array(
626         val, index, dtype=dtype, copy=False, raise_cast_failure=False
627     )
628     com.require_length_match(val, index)
630 homogenized.append(val)

```

File ~\anaconda3\lib\site-packages\pandas\core\construction.py:598, in sanitize\_array(data, index, dtype, copy, raise\_cast\_failure, allow\_2d)

```

595         subarr = cast(np.ndarray, subarr)
596         subarr = maybe_infer_to_datetimelike(subarr)
--> 598 subarr = _sanitize_ndim(subarr, data, dtype, index, allow_2d=allow_2d)
600 if isinstance(subarr, np.ndarray):
601     # at this point we should have dtype be None or subarr.dtype == dtype
602     dtype = cast(np.dtype, dtype)

```

File ~\anaconda3\lib\site-packages\pandas\core\construction.py:649, in \_sanitize\_ndim(result, data, dtype, index, allow\_2d)

```

647     if allow_2d:
648         return result
--> 649     raise ValueError("Data must be 1-dimensional")
650 if is_object_dtype(dtype) and isinstance(dtype, ExtensionDtype):
651     # i.e. PandasDtype("O")
653     result = com.asarray_tuplesafe(data, dtype=np.dtype("object"))

```

**ValueError:** Data must be 1-dimensional

In [ ]:

