$Import\ biblioteki\ \textbf{TensorFlow}\ (\underline{\text{https://www.tensorflow.org/}})\ z\ której\ będziemy\ korzystali\ w\ \textbf{uczeniu\ maszynowym}:$

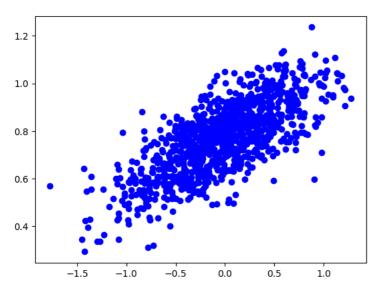
```
import tensorflow as tf
import matplotlib.pyplot as plt
import numpy as np

number_of_points = 1000
x_point = []
y_point = []

a = 0.22
b = 0.78

for i in range(number_of_points):
    x = np.random.normal(0.0,0.5)
    y = (a*x+b)+np.random.normal(0.0,0.1)
    x_point.append(x)
    y_point.append(y)

plt.scatter(x_point,y_point,c='b')
plt.show()
```



```
real_x = np.array(x_point)
real_y = np.array(y_point)
```

Batch Stochastic Gradient Descent - wykorzystujemy cały zbiór danych

Definicja błędu:

```
def loss_fn(real_y, pred_y):
    return tf.reduce_mean((real_y - pred_y)**2)
import random

Loss = []
epochs = 50
learning_rate = 0.5

a = tf.Variable(random.random())
b = tf.Variable(random.random())

for _ in range(epochs):

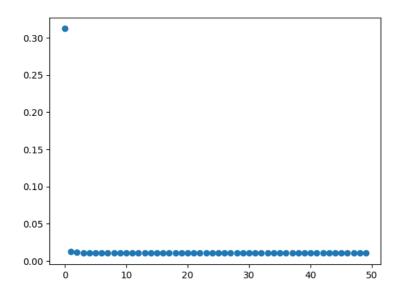
with tf.GradientTape() as tape:
    pred_y = a * real_x + b
    loss = loss_fn(real_y, pred_y)
    Loss.append(loss.numpy())
    grad_a, grad_b = tape.gradient(loss,(a, b))
```

```
a.assign_sub(learning_rate*grad_a)
b.assign_sub(learning_rate*grad_b)

np.max(Loss),np.min(Loss)
    (0.31212983, 0.010254185)

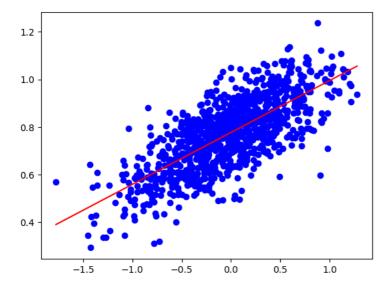
print(a.numpy())
print(b.numpy())
    0.21805292
    0.77771721

plt.scatter(np.arange(epochs),Loss)
plt.show()
```



```
max = np.max(x_point)
min = np.min(x_point)

X = np.linspace(min, max, num=10)
plt.plot(X,a.numpy()*X+b.numpy(),c='r')
plt.scatter(x_point,y_point,c="b")
plt.show()
```



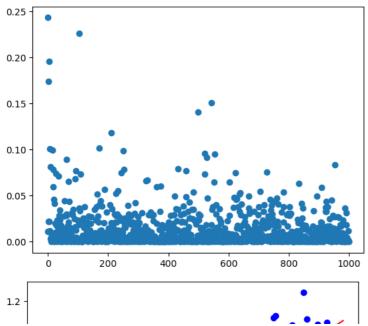
Mini-batch Stochastic Gradient Descent - wykorzystujemy część zbióru danych

```
arr = np.arange(10)
arr
array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9])
```

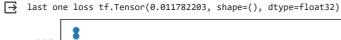
```
6.11.2023, 10:15
```

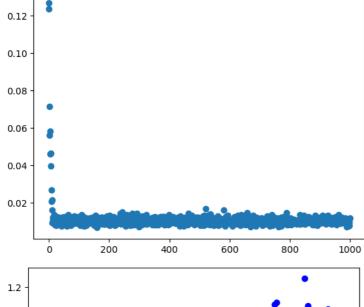
```
np.random.shuffle(arr)
    array([2, 8, 0, 1, 7, 4, 5, 6, 3, 9])
def subset_dataset(x_dataset, y_dataset, subset_size):
   arr = np.arange(len(x_dataset))
   np.random.shuffle(arr)
   x_train = x_dataset[arr[0:subset_size]]
   y_train = y_dataset[arr[0:subset_size]]
   return x_train,y_train
print(subset_dataset(real_x, real_y, 10))
    0.83270043, 0.64120408, 0.92022443, 0.92671379, 0.59391998]))
TODO:
def mini_batch_stochastic_gradient_descent(batch_size):
 Loss = []
 epochs = 1000
 learning_rate = 0.2
                           #wielkość zbioru wykorzystanego do treningu
 batch_size = batch_size
 a = tf.Variable(random.random())
 b = tf.Variable(random.random())
 for i in range(epochs):
   real_x_batch, real_y_batch = subset_dataset(real_x,real_y,batch_size)
   with tf.GradientTape() as tape:
     pred_y = a * real_x_batch + b
     loss = loss_fn(real_y_batch, pred_y)
     Loss.append(loss.numpy())
   dloss_da, dloss_db = tape.gradient(loss,(a, b))
   a.assign_sub(learning_rate*dloss_da) #a = a - alpha*dloss_da
   b.assign_sub(learning_rate*dloss_db) #b = b - alpha*dloss_db
 print("last one loss", str(loss))
 plt.scatter(np.arange(epochs),Loss)
 plt.show()
 max = np.max(x_point)
 min = np.min(x_point)
 X = np.linspace(min, max, num=10)
 plt.plot(X,a.numpy()*X+b.numpy(),c='r')
 plt.scatter(x_point,y_point,c="b")
 plt.show()
Wykres zmian błędu:
mini_batch_stochastic_gradient_descent(1)
```

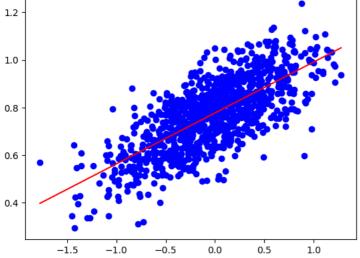
last one loss tf.Tensor(3.4965207e-05, shape=(), dtype=float32)



 $\verb|mini_batch_stochastic_gradient_descent(100)|\\$







mini_batch_stochastic_gradient_descent(1000)

last one loss tf.Tensor(0.010254186, shape=(), dtype=float32)

