```
import numpy as np
                                   + Code
                                               + Text
# np.array is used to create a NumPy matrix
# 1 dimensional array
a1=np.array([1, 4, 5, 7.8, 9, 5, 3])
print(type(a1))
     <class 'numpy.ndarray'>
# 2 dimensional array
a2=np.array([[2, 3], [5, 8], [8, 9]]) # r * c matrix, 3 x 2 matrix
print(a2)
     [[2 3]
      [5 8]
      [8 9]]
# np.zeros() to make a matrix containing only zeros
a3=(2,2)
print(np.zeros(a3))
     [[0. 0.]
      [0. 0.]]
# np.ones() to make a matrix containing only ones
a4=(2,2)
print(np.ones(a3))
     [[1. 1.]
      [1. 1.]]
# to create a array containing all the numbers between a lower and upper bound
a5=np.arange(5, 12) # 5 is included but 12 is not included
print(a5)
     [567891011]
a6=np.random.randint(low=0, high=101, size=(5))
print(a6)
# Note that the highest generated integer np.random.randint is one less than the high argu
     [88 24 51 64 28]
# To create random floating-point values between 0.0 and 1.0, call np.random.random.
a7=np.random.random([5])
print(a7)
     [0.44706797 0.17606624 0.12805897 0.46729196 0.11863898]
```

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# mathematical operations on matrices
# If you want to add or subtract two vectors or matrices, linear algebra requires that the
# Furthermore, if you want to multiply two vectors or matrices, linear algebra imposes str
# Fortunately, NumPy uses a trick called broadcasting to virtually expand the smaller oper
a8 = a7 + 2
print(a8)
a9=a7*3
print(a9)
     [2.44706797 2.17606624 2.12805897 2.46729196 2.11863898]
     [1.34120391 0.52819872 0.38417692 1.40187589 0.35591694]
# Task 1: Create a Linear Dataset
# Your goal is to create a simple dataset consisting of a single feature and a label as fc
# 1. Assign a sequence of integers from 6 to 20 (inclusive) to a NumPy array named feature
# 2. Assign 15 values to a NumPy array named label such that: label = (3)(feature) + 4
feature = np.arange(6, 21)
print(feature)
label = (3*feature)+4
print(label)
     [ 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20]
     [22 25 28 31 34 37 40 43 46 49 52 55 58 61 64]
# Task 2: Add Some Noise to the Dataset
# To make your dataset a little more realistic, insert a little random noise into each \operatorname{\mathsf{el}} \varepsilon
# To be more precise, modify each value assigned to label by adding a different random flc
noise = (np.random.random([15]) * 4) - 2
print(noise)
label = label + noise
print(label)
     [ \ 0.72507294 \ -1.92801494 \ \ 1.41395433 \ -1.54584829 \ \ 1.66641674 \ -0.54999582
      -1.58599986 -0.17834759 1.11460044 -1.33215644 0.36530526 -0.31494639
      -0.35164051 0.03714895 -1.79619833]
     [22.72507294 23.07198506 29.41395433 29.45415171 35.66641674 36.45000418
      38.41400014 42.82165241 47.11460044 47.66784356 52.36530526 54.68505361
      57.64835949 61.03714895 62.20380167
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