

Gradient Descent Practical Demonstration

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In [1]:

```
1 import numpy as np
2 import matplotlib.pyplot as plt
```

In [2]:

```
1 x = np.arange(10)
2 x
```

Out[2]:

```
array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9])
```

In [3]:

```
1 # creating y which is a function of x
2
3 # y = f(x - 5)^2
4
5 y = (x-5)**2
6 y
```

Out[3]:

```
array([25, 16, 9, 4, 1, 0, 1, 4, 9, 16])
```

Goal of Gradient Descent

Given a function $f(x)$, we want to find the value of x that Minimizes f

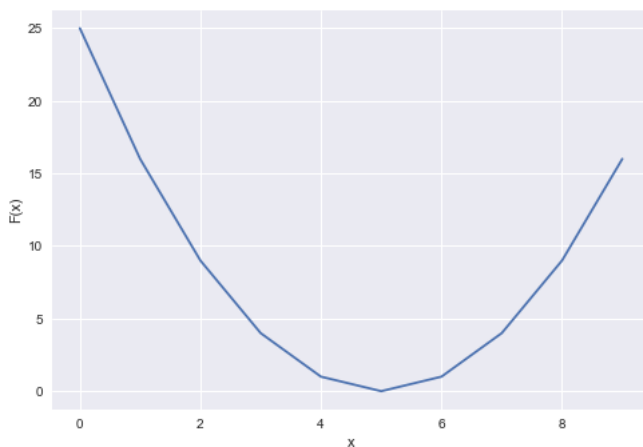
Visualizing X vs Y

In [4]:

```
1 plt.style.use("seaborn")
2 plt.plot(x,y)
3 plt.xlabel("x")
4 plt.ylabel("F(x)")
```

Out[4]:

Text(0, 0.5, 'F(x)')



```
1 As y = f(x - 5)^2 => Gradient of this is => 2(x-5) which is simply the differentiation w.r.t x of the function
```

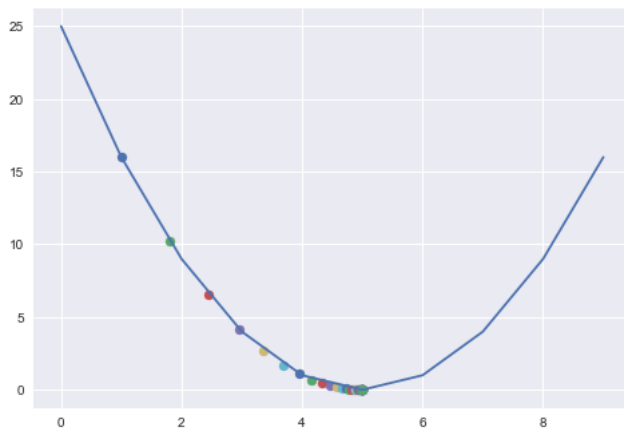
Implementing Gradient Descent

In [5]:

```

1 random_value = 0
2 Learning_rate = 0.1
3 error = []
4
5 plt.plot(x,y)
6
7 for i in range(50):
8     gradient = 2*(random_value-5)
9     random_value = random_value - (Learning_rate * gradient)
10
11     Y = (random_value-5)**2
12     error.append(Y)
13
14     print("Random Value -> ", random_value, "\nCost -> ", Y)
15
16     plt.scatter(random_value,Y)
17
18     print("\n=====")
19
20

```



1 we can observe that the step gradients are converging because the magnitude of $(\frac{\partial f(x)}{\partial x})$ is decreasing and can observe how our gradient descent is converging

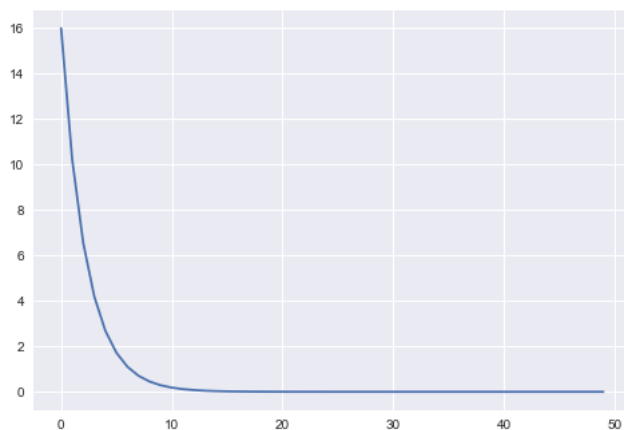
Plotting the errors

In [6]:

```
1 plt.plot(error)
```

Out[6]:

```
[<matplotlib.lines.Line2D at 0x1ca20387df0>]
```



1 this demonstrates that as we increase the number of steps, after around 10, the error is being saturated