

## section\_2

February 7, 2025

### 1 Section 2. Error Analysis Using Plots

This notebook is designed to guide you through plotting error for rootfinding methods.

Please submit your answers to the questions below along with screenshots of the plots you make while working through this.

Let's continue working with the very simple equation  $f(x) = x^2 - 2$  with solution  $p = \sqrt{2}$ .

Since we know the exact solution to this equation, we can compare the exact answer to the value of the approximation at each iteration and calculate the absolute error:

$$\text{Absolute Error} = |\text{Approximate Solution} - \text{Exact Solution}|.$$

The code cell below defines the function we developed in class to carry out the bisection method with one small change.

```
[1]: import numpy as np

def my_bisection_method(my_func,a,b,TOL,exact_root):

    #check if initial point satisfies the initial criteria
    if my_func(a) == 0:
        print("The root is {}".format(a))
        return a

    if my_func(b) == 0:
        print("The root is {}".format(b))
        return b

    if my_func(a)*my_func(b) > 0:
        print("Sign of f(a) and sign of f(b) must be opposite")
        return None

    curr_iter = 1
    error_vec = []

    #Looping Bisection Method
    while (b-a)/2 > TOL:
```

```

print("The current iteration is {}".format(curr_iter))
print("The current error is = {}".format((b-a)/2))
p = a + (b-a)/2
abs_error = np.abs(p-exact_root)
error_vec.append(abs_error)
print("The approximation p_k = {}".format(p))
print("f(p_k) = {}".format(my_func(p)))

if my_func(p)==0:
    print("The root is {}".format(p))
    return p

if my_func(a)*my_func(p)>0:
    a = p
else:
    b = p

curr_iter = curr_iter + 1
print("Current error is {}".format((b-a)/2))
return p, error_vec

```

**Question 1:** How many inputs does the function have? What are they? How many outputs does the function have? What are they?

- The function has inputs (my\_func,a,b,TOL,exact\_root), so five total, all required
  - The function we’re trying to find a root for
  - The lower bound
  - The upper bound
  - The tolerance we’re looking to get the root within
  - The actual root we’re comparing against
- The function prototype isn’t explicitly declared, but, analyzing the return statement ‘return p, error\_vec’ we can determine that we have two outputs
  - The independent value of the root we’ve found
  - A list of error values (a vector)

Next I will show you how to call a function with two outputs.

```

[2]: import numpy as np

my_func = lambda x: x**2-2
a = 0
b = 2
tol = .001
exact_root = np.sqrt(2)

approx_root, error_vec = my_bisection_method(my_func,a,b,tol,exact_root)

error_vec_length = len(error_vec)

```

```
print("The length of the error vector is {}".format(error_vec_length))
```

```
The current iteration is 1
The current error is = 1.0
The approximation p_k = 1.0
f(p_k) = -1.0
The current iteration is 2
The current error is = 0.5
The approximation p_k = 1.5
f(p_k) = 0.25
The current iteration is 3
The current error is = 0.25
The approximation p_k = 1.25
f(p_k) = -0.4375
The current iteration is 4
The current error is = 0.125
The approximation p_k = 1.375
f(p_k) = -0.109375
The current iteration is 5
The current error is = 0.0625
The approximation p_k = 1.4375
f(p_k) = 0.06640625
The current iteration is 6
The current error is = 0.03125
The approximation p_k = 1.40625
f(p_k) = -0.0224609375
The current iteration is 7
The current error is = 0.015625
The approximation p_k = 1.421875
f(p_k) = 0.021728515625
The current iteration is 8
The current error is = 0.0078125
The approximation p_k = 1.4140625
f(p_k) = -0.00042724609375
The current iteration is 9
The current error is = 0.00390625
The approximation p_k = 1.41796875
f(p_k) = 0.0106353759765625
The current iteration is 10
The current error is = 0.001953125
The approximation p_k = 1.416015625
f(p_k) = 0.005100250244140625
Current error is 0.0009765625
The length of the error vector is 10
```

**Question 2:** Is the number of iterations consistent with the theory?

Next we will make plots of the absolute error on the vertical axis and the iteration number on the

horizontal axis.

Run the code cell below to redefine the bisection method function so it carries out more iterations than needed. This way our plots will look better.

```
[3]: import numpy as np
#this code block defines the function called my_bisection_method
#INPUTS: function my_func, a, b, TOL, exact root
#OUTPUTS: root or a error message
def my_bisection_method(my_func,a,b,TOL,exact_root,max_iter):
    if my_func(a) == 0:
        print("The root is {}".format(a))
        return a#this end the function
    if my_func(b) == 0:
        print("The root is {}".format(b))
        return b#this end the function
    if my_func(a)*my_func(b) > 0:
        print("Sign of f(a) and sign of f(b) must be opposite")
        return None#this end the function
    curr_iter = 1
    error_vec = []
    for i in range(max_iter):#current interval divided by two bigger than tol
        print("The current iteration is {}".format(curr_iter))
        print("The current error is = {}".format((b-a)/2))
        p = a + (b-a)/2
        #compute /approximate root - exact root/
        abs_error = np.abs(p-exact_root)
        error_vec.append(abs_error)
        print("The approximation p_k = {}".format(p))
        print("f(p_k) = {}".format(my_func(p)))
        if my_func(p)==0:
            print("The root is {}".format(p))
            return p#this end the function
        if my_func(a)*my_func(p)>0:
            a = p
        else:
            b = p
        curr_iter = curr_iter + 1
    print("Current error is {}".format((b-a)/2))
    return p, error_vec
```

```
[4]: import numpy as np
#define the inputs
my_func = lambda x: x**2-2
a = 0
b = 2
tol = .001
exact_root = np.sqrt(2)
```

```
max_iter = 30
#call the function
approx_root, error_vec = □
    ↪ my_bisection_method(my_func,a,b,tol,exact_root,max_iter)
```

```
The current iteration is 1
The current error is = 1.0
The approximation p_k = 1.0
f(p_k) = -1.0
The current iteration is 2
The current error is = 0.5
The approximation p_k = 1.5
f(p_k) = 0.25
The current iteration is 3
The current error is = 0.25
The approximation p_k = 1.25
f(p_k) = -0.4375
The current iteration is 4
The current error is = 0.125
The approximation p_k = 1.375
f(p_k) = -0.109375
The current iteration is 5
The current error is = 0.0625
The approximation p_k = 1.4375
f(p_k) = 0.06640625
The current iteration is 6
The current error is = 0.03125
The approximation p_k = 1.40625
f(p_k) = -0.0224609375
The current iteration is 7
The current error is = 0.015625
The approximation p_k = 1.421875
f(p_k) = 0.021728515625
The current iteration is 8
The current error is = 0.0078125
The approximation p_k = 1.4140625
f(p_k) = -0.00042724609375
The current iteration is 9
The current error is = 0.00390625
The approximation p_k = 1.41796875
f(p_k) = 0.0106353759765625
The current iteration is 10
The current error is = 0.001953125
The approximation p_k = 1.416015625
f(p_k) = 0.005100250244140625
The current iteration is 11
The current error is = 0.0009765625
The approximation p_k = 1.4150390625
```

```

f(p_k) = 0.0023355484008789062
The current iteration is 12
The current error is = 0.00048828125
The approximation p_k = 1.41455078125
f(p_k) = 0.0009539127349853516
The current iteration is 13
The current error is = 0.000244140625
The approximation p_k = 1.414306640625
f(p_k) = 0.0002632737159729004
The current iteration is 14
The current error is = 0.0001220703125
The approximation p_k = 1.4141845703125
f(p_k) = -8.200109004974365e-05
The current iteration is 15
The current error is = 6.103515625e-05
The approximation p_k = 1.41424560546875
f(p_k) = 9.063258767127991e-05
The current iteration is 16
The current error is = 3.0517578125e-05
The approximation p_k = 1.414215087890625
f(p_k) = 4.314817488193512e-06
The current iteration is 17
The current error is = 1.52587890625e-05
The approximation p_k = 1.4141998291015625
f(p_k) = -3.8843369111418724e-05
The current iteration is 18
The current error is = 7.62939453125e-06
The approximation p_k = 1.4142074584960938
f(p_k) = -1.726433401927352e-05
The current iteration is 19
The current error is = 3.814697265625e-06
The approximation p_k = 1.4142112731933594
f(p_k) = -6.474772817455232e-06
The current iteration is 20
The current error is = 1.9073486328125e-06
The approximation p_k = 1.4142131805419922
f(p_k) = -1.0799813026096672e-06
The current iteration is 21
The current error is = 9.5367431640625e-07
The approximation p_k = 1.4142141342163086
f(p_k) = 1.6174171832972206e-06
The current iteration is 22
The current error is = 4.76837158203125e-07
The approximation p_k = 1.4142136573791504
f(p_k) = 2.687177129701013e-07
The current iteration is 23
The current error is = 2.384185791015625e-07
The approximation p_k = 1.4142134189605713

```

```

f(p_k) = -4.056318516632018e-07
The current iteration is 24
The current error is = 1.1920928955078125e-07
The approximation p_k = 1.4142135381698608
f(p_k) = -6.845708355740499e-08
The current iteration is 25
The current error is = 5.960464477539063e-08
The approximation p_k = 1.4142135977745056
f(p_k) = 1.0013031115363447e-07
The current iteration is 26
The current error is = 2.9802322387695312e-08
The approximation p_k = 1.4142135679721832
f(p_k) = 1.583661290993632e-08
The current iteration is 27
The current error is = 1.4901161193847656e-08
The approximation p_k = 1.414213553071022
f(p_k) = -2.6310235545778937e-08
The current iteration is 28
The current error is = 7.450580596923828e-09
The approximation p_k = 1.4142135605216026
f(p_k) = -5.236811428943611e-09
The current iteration is 29
The current error is = 3.725290298461914e-09
The approximation p_k = 1.414213564246893
f(p_k) = 5.29990096254096e-09
The current iteration is 30
The current error is = 1.862645149230957e-09
The approximation p_k = 1.4142135623842478
f(p_k) = 3.154454475406965e-11
Current error is 9.313225746154785e-10

```

```

[5]: # next we will plot the absolute error on the vertical axis and the iteration
      ↪ number on the horizontal axis
import matplotlib.pyplot as plt
# "formula" for making a scatterplot
# plt.plot(x_vals, y_vals)
# in this case the x_vals should be the set of iterations x_vals =
      ↪ [1,2,3,4,5,6,7,8,9,10]
# there are many ways to make collections of numbers like this in Python
# we will keep using numpy
x_vals = np.arange(1,11)
print(x_vals)
# notice it leaves off the last number

```

```
[ 1  2  3  4  5  6  7  8  9 10]
```

```
[6]: # the best way to set up this vector is below
```

```

x_vals = np.arange(1, len(error_vec)+1) # the len(error_vec) function returns the
↳ # of things in the error_vec array
print(x_vals)
# now we make the plot
plt.scatter(x_vals, error_vec)
# add some titles so we know what we are looking at
plt.xlabel("Iteration Number")
plt.ylabel("Absolute Error")

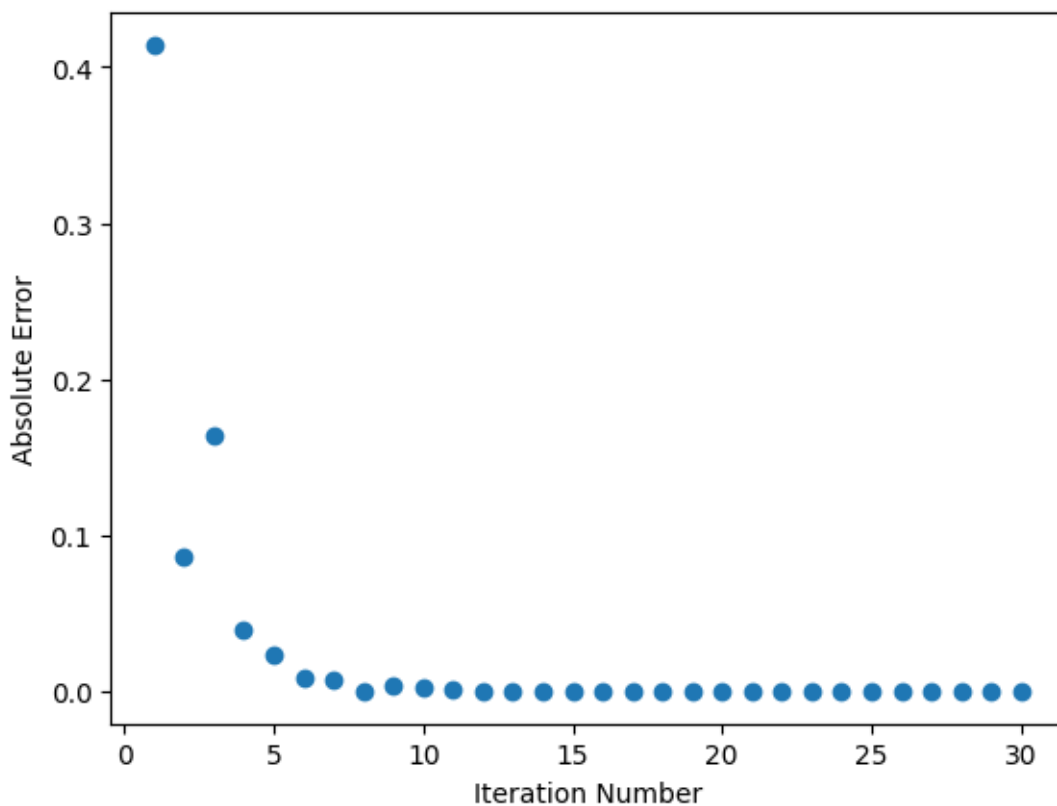
```

```

[ 1  2  3  4  5  6  7  8  9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24
 25 26 27 28 29 30]

```

```
[6]: Text(0, 0.5, 'Absolute Error')
```



As expected, the absolute error follows an exponentially decreasing trend. Namely,  $\frac{b-a}{2^n}$ .

Notice that it isn't a completely smooth curve since we will have some jumps in the accuracy just due to the fact that sometimes the root will be near the midpoint of the interval and sometimes it won't be.

```

[7]: # we can add the theoretical curve to our plot
a = 0

```

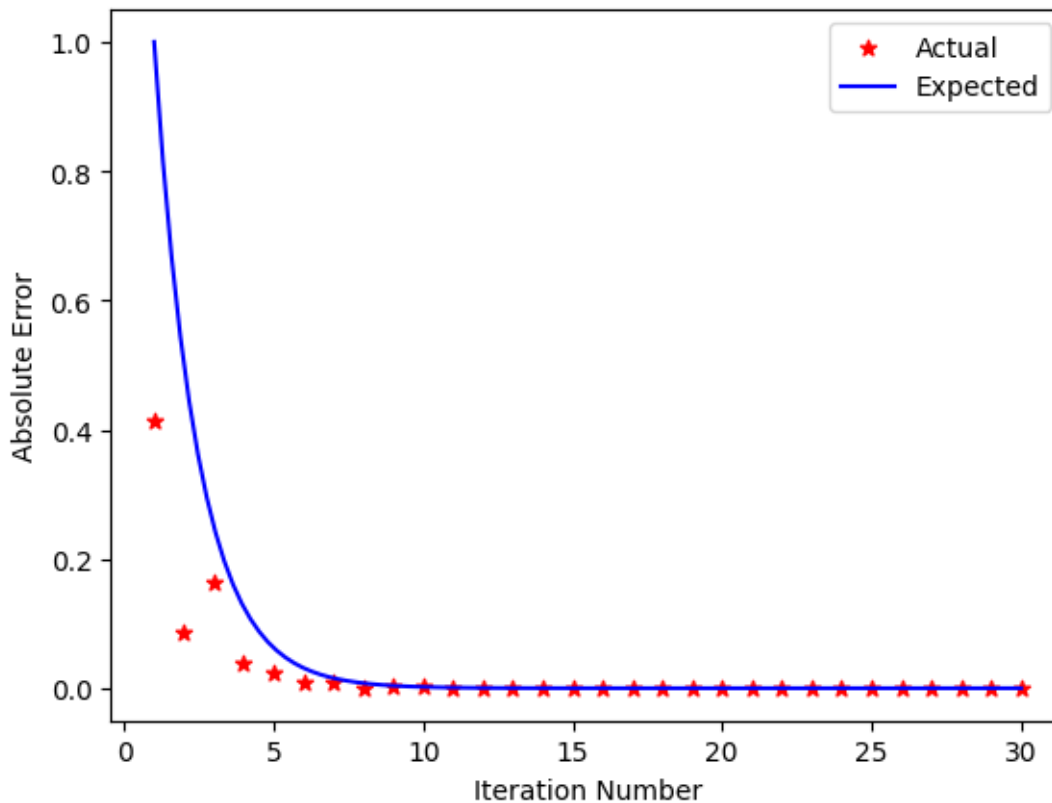


```

b = 2
error_func = lambda n: (b-a)/(2**n)
# now we make the plot with the error vals we computed vs. this function
plt.scatter(x_vals, error_vec, label = "Actual", marker = '*', color = 'r')
# I added a label so we could make a legend, we can also specify the marker
↳ type and color
smooth_x_vals = np.linspace(1,max_iter,100)
plt.plot(smooth_x_vals, error_func(smooth_x_vals), label = "Expected", color = 'b'
↳ 'b')
plt.xlabel("Iteration Number")
plt.ylabel("Absolute Error")
# this uses the labels I defined in the plot call above
plt.legend()

```

[7]: <matplotlib.legend.Legend at 0x1f7411ad190>



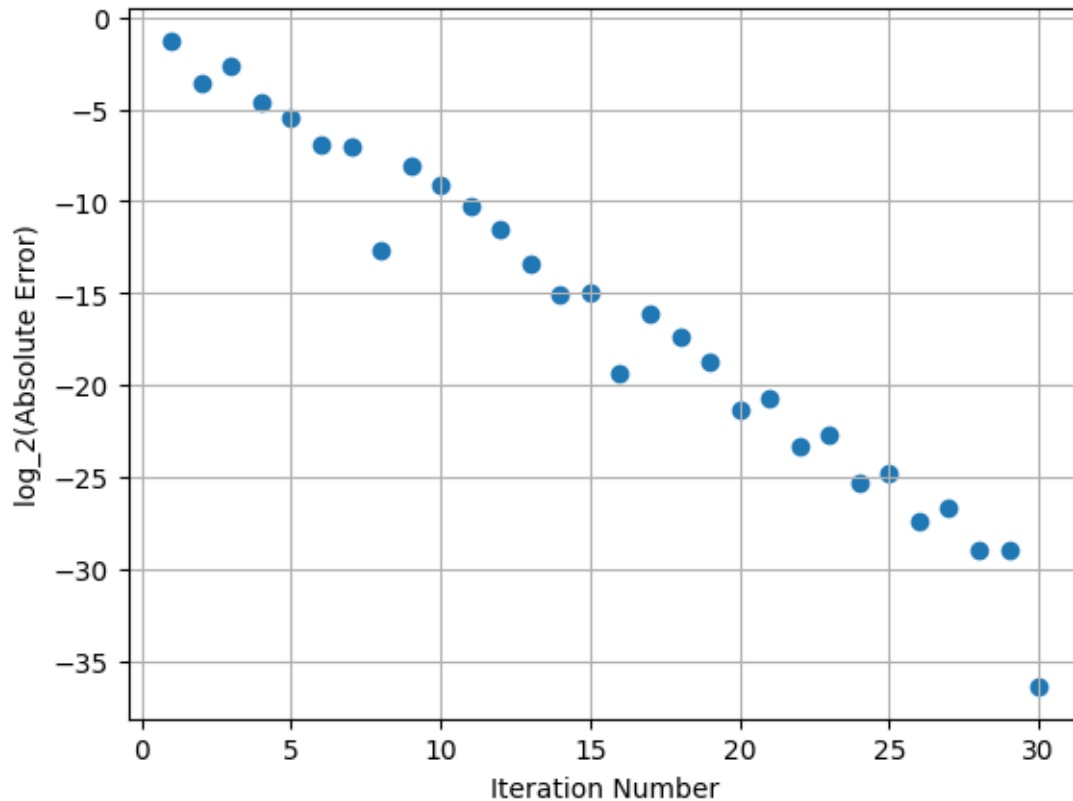
Without the theory, it would be really hard to determine what the exact behavior is in the exponential plot above. We know that the error will divide by 2 at every step, so if we instead plot

$$\log_2(\text{errorvec})$$

against the iteration number we should see a linear trend.

```
[8]: plt.scatter(x_vals, np.log2(error_vec))  
      # Add grid lines  
      plt.grid(True)  
      plt.xlabel("Iteration Number")  
      plt.ylabel("log_2(Absolute Error)")
```

```
[8]: Text(0, 0.5, 'log_2(Absolute Error)')
```



**Question 3:** What is the slope of the line in the plot above (estimate it)? Why does that slope make sense?

Another plot we can use to determine how an algorithm is behaving as it progresses is to compare the absolute error at iteration  $k$  to the absolute error at iteration  $k + 1$ .

Run the code below to print the length of the `error_vec`.

```
[9]: print(len(error_vec))
```

30

```
[10]: # let's learn how to pull out values at certain indices of a vector like error_vec
      ↪ error_vec
      error_at_k = error_vec[0:len(error_vec)-1] #this pulls out the first 29 values
      # Note that: 1) we start at 0 not 1, so the indices of the vector go from 0-29
      # in python this code will go up to but not include the value in the 29th and
      ↪ last index
      print(len(error_at_k))
```

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**Question 4:** How would you pull out the first five values?

```
[11]: error_first_5_values = error_vec[0:5]
      print(len(error_first_5_values))
```

5

**Question 5:** Make a vector called `error_at_k_plus_1` with the last 29 values of `error_vec`.

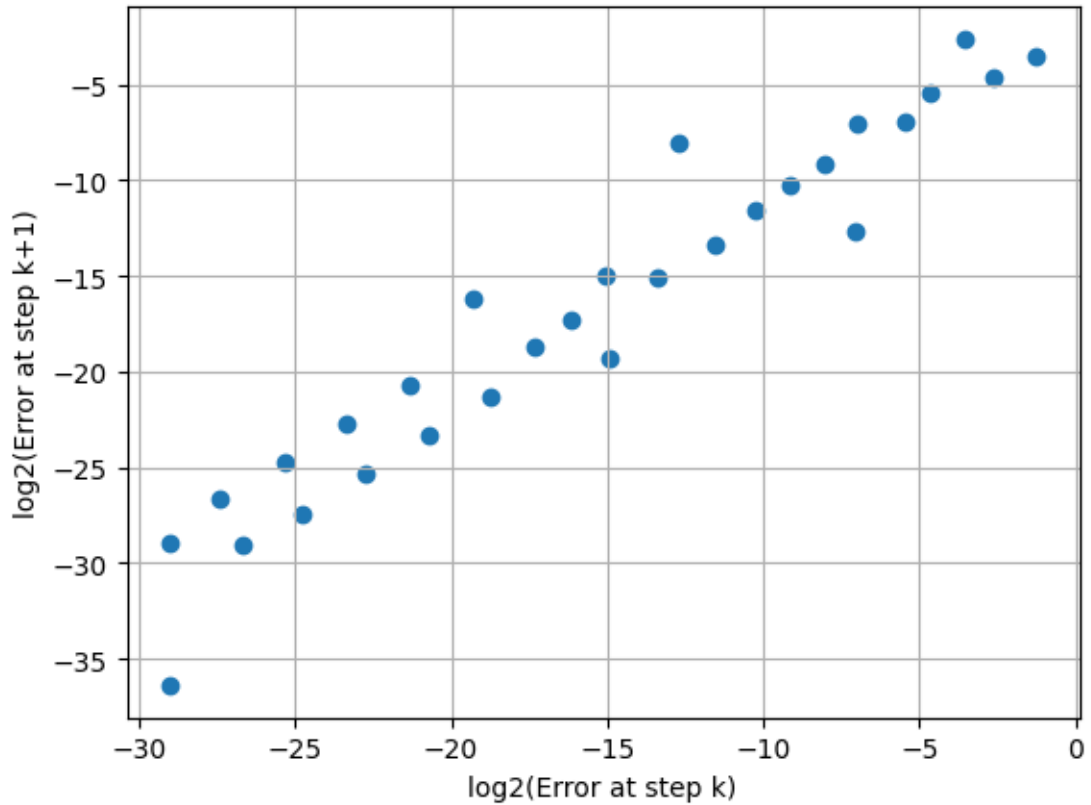
```
[12]: error_at_k_plus_1 = error_vec[1:len(error_vec)]
      print(len(error_at_k_plus_1))
```

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**Question 6:** Which values does the line above pull out?

```
[13]: # write code to plot error_at_k_plus_1 vs. error_at_k
      plt.scatter(np.log2(error_at_k), np.log2(error_at_k_plus_1))
      # Add grid lines
      plt.grid(True)
      plt.xlabel("log2(Error at step k)")
      plt.ylabel("log2(Error at step k+1)")
```

```
[13]: Text(0, 0.5, 'log2(Error at step k+1)')
```



**Question 7** Interpret the graph above.

The graph is displaying linear behaviour, a straight line, let's break down the details. - We can see along the left side that our dependent variable is  $\log_2(x)$  of the step ahead - We can see along the bottom side that our independent variable is  $\log_2(x)$  of the current step - We're taking the error at both - Taken all together, we can see that the subsequent step is twice as accurate, half as inaccurate - We're demonstrating here the behaviour of the bisection method, we're throwing out half the search space each time, so, we're converging at  $O(\log_2(n))$