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In [1]: import numpy as np
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Floating point arithmetic

The Machine epsilon is defined as the smallest floating point number such that it holds: $\text{fl}(1+\text{eps}) > 1$. Compute eps. Tips: use a while structure.

```
In [2]: eps = 1.0 # initialize epsilon
while 1 + eps > 1:
    eps /= 2
eps *= 2
print(eps)

2.220446049250313e-16
```

Let's consider the sequence $a_n = \left(1 + \frac{1}{n}\right)^n$. It is well known that: $\lim_{x \rightarrow \infty} a_n = e$, where e is the Nepero number. Choose different values for n , compute a_n and compare it to the real value of the Nepero number. What happens if you choose a large value of n ?

```
In [4]: from matplotlib.pyplot import subplots

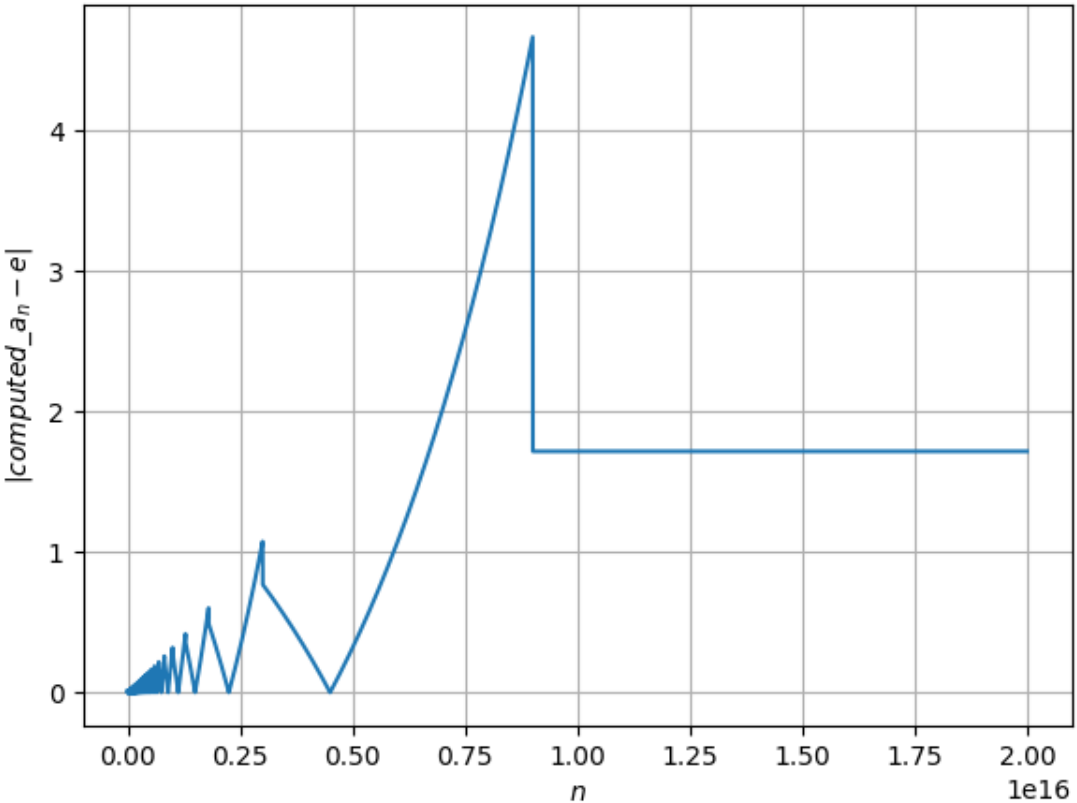
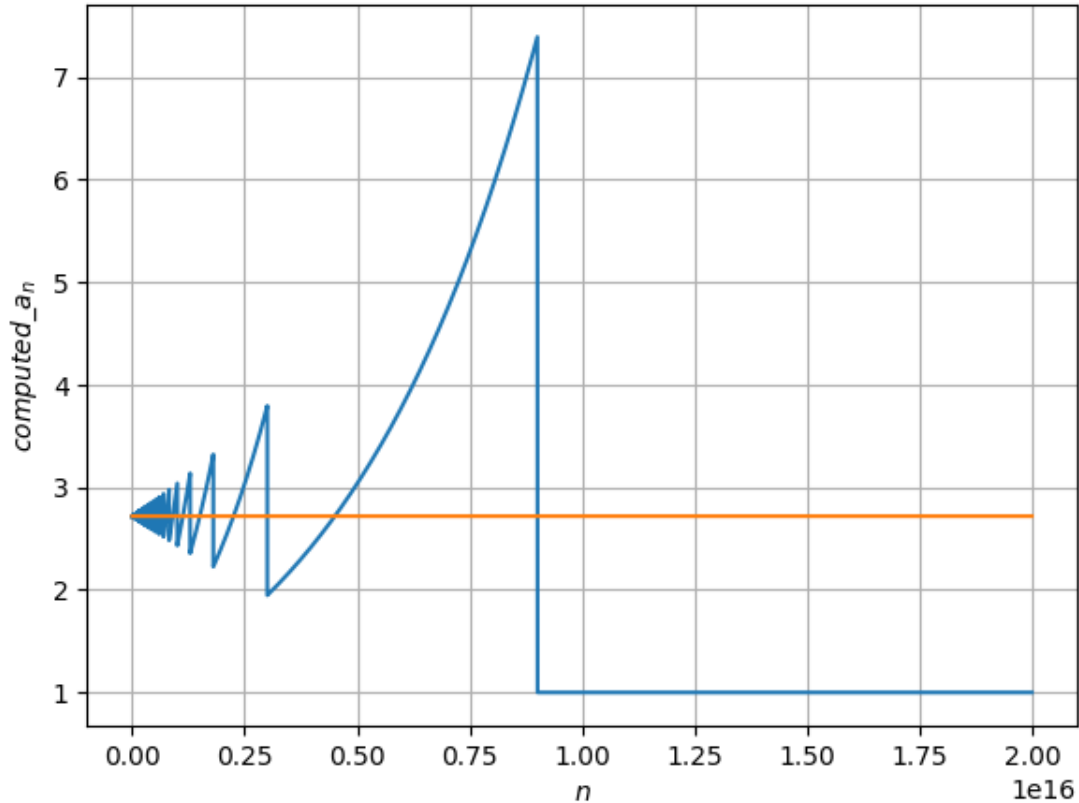
#n_values = np.arange(1, 100, 1)
#n_values = np.arange(10, 1e14, 1e8)
#n_values = np.arange(100, 5e15, 1e10)
n_values = np.arange(100, 2e16, 1e10)

a_n = (1 + (1/n_values))**n_values
abs_err = abs(a_n - np.e)

fig, axs = subplots(figsize=(15,5), nrow=1, ncol=2)

ax = axs[0]
x = range(1, len(n_values)+1)
ax.plot(n_values, a_n)
ax.plot(n_values, np.full(len(a_n), np.e))
ax.set_xlabel(r"$n$")
ax.set_ylabel(r"$computed\_a\_n$")
#add legend
ax.grid()

ax = axs[1]
ax.plot(n_values, abs_err)
ax.set_xlabel(r"$n$")
ax.set_ylabel(r"$|computed\_a\_n - e|$")
ax.grid()
```



Compute the rank of A and B and their eigenvalues. Are A and B full-rank matrices? Can you infer some relationship between the values of the eigenvalues and the full-rank condition? Please, corroborate your deduction with other examples.

```
In [5]: A = np.array([[4, 2], [1, 3]])
        B = np.array([[4, 2], [2, 1]])

rk_A = np.linalg.matrix_rank(A) #np.linalg.matrix_rank uses SVD decomposition to compute the rank
rk_B = np.linalg.matrix_rank(B)
print('Rk(A) = ' + str(rk_A) + ', so A is full-rank')
print('Rk(B) = ' + str(rk_B) + ', so A is not full-rank'+ '\n')

eigenvals_A = np.linalg.eigvals(A)
eigenvals_B = np.linalg.eigvals(B)
print("Eigenvalue of A: " + str(eigenvals_A))
print("Eigenvalue of B: " + str(eigenvals_B))

Rk(A) = 2, so A is full-rank
Rk(B) = 1, so A is not full-rank

Eigenvalue of A: [5. 2.]
Eigenvalue of B: [5. 0.]
```

We can observe that a square matrix has full rank if and only if all its eigenvalues are non-zero. For example the matrix $A = \begin{bmatrix} 1 & 2 \\ 0 & 3 \end{bmatrix}$ is full-rank with eigenvalues 1 and 3. The matrix

$$B = \begin{bmatrix} 1 & 2 \\ 0 & 0 \end{bmatrix} \tag{2}$$

is not full-rank with eigenvalues 1 and 0.

```
In [ ]:
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