

ITN_Assignment2

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1 Assignment 2

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1.0.1 Intro to Theoretical Neuroscience

Problem 1

a) Integrating using Euler

```
In [1]: import numpy as np
import matplotlib.pyplot as plt

def diff_x(x, t):
    return -0.5*x + 1 + np.sin(3*t)

# initial conditions x(0) = 5
# and we have euler integration  $x(t + \delta t) = x(t) + \delta t * dx(t)/dt$ 
def integrate_euler(start, end, step_size):
    steps = int(np.ceil((end-start)/step_size))
    result = np.zeros(steps+1)
    result[0] = 5
    step_range = np.array(range(0, steps+1))
    t_range = step_range*step_size
    for i in step_range[1:]:
        t = t_range[i]
        result[i] = result[i-1] + diff_x(result[i-1], t)*step_size
    return [t_range, result]

rng, res = integrate_euler(0.0, 5.0, 0.5)
plt.plot(rng, res)
rng, res = integrate_euler(0.0, 5.0, 0.01)
plt.plot(rng, res)
plt.show()
```

<Figure size 640x480 with 1 Axes>

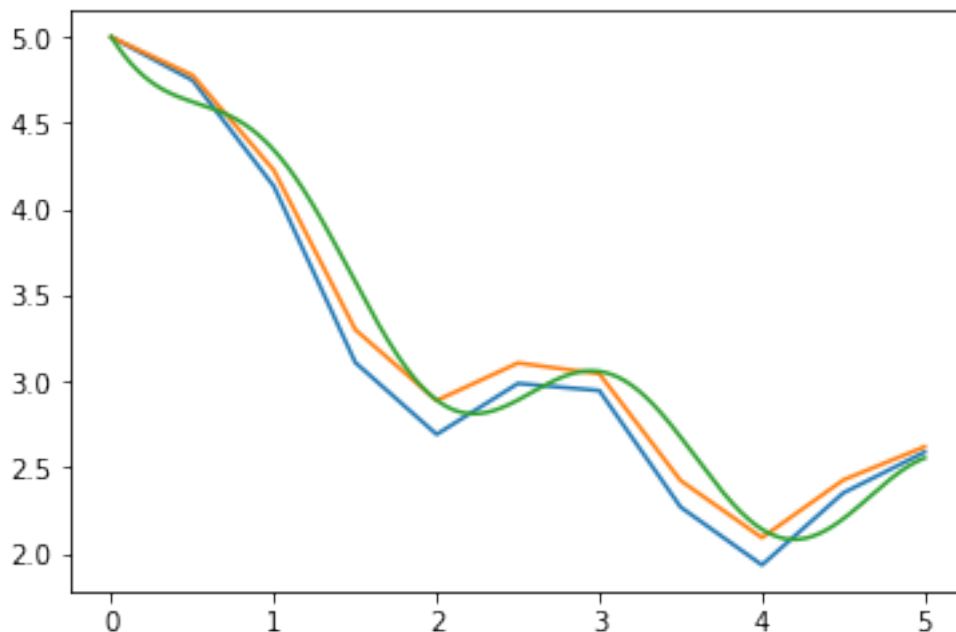
b) Integrating using exponential integration scheme

```
In [2]: def a(t):
        return (1 + np.sin(3*t))*2.0

def integrate_exp(start, end, step_size):
    # tau = 2 from equation
    ss_over_tau = step_size/2.0
    steps = int(np.ceil((end-start)/step_size))
    result = np.zeros(steps+1)
    result[0] = 5
    step_range = np.array(range(0, steps+1))
    t_range = step_range*step_size
    for i in step_range[1:]:
        t = t_range[i]
        result[i] = a(t) + (result[i-1] - a(t))*np.exp(-ss_over_tau)
    return [t_range, result]

rng_euler, res_euler = integrate_euler(0.0, 5.0, 0.5)
plt.plot(rng_euler, res_euler)
rng_exp, res_exp = integrate_exp(0.0, 5.0, 0.5)
plt.plot(rng_exp, res_exp)
rng, res = integrate_euler(0.0, 5.0, 0.01)
plt.plot(rng, res)
plt.show()

idx = [i in rng_exp for i in rng]
print('Euler squared error', ((res_euler-res[idx])**2).sum())
print('Exp squared error', ((res_exp-res[idx])**2).sum())
```



Euler squared error 0.5717565737544987

Exp squared error 0.28193457388216314

Blue: euler, delta_t = 0.5 Orange: exp, delta_t = 0.5 Green: euler, delta_t = 0.01

Using green as baseline, we see that Orange (exp) does better than euler with the same time step in approximating green. (Check squared error)

Problem 2

```
In [3]: c_m = 10.0e-3
        g_L = 1.0
        E_L = -70e-3
        g_ex = 0.5
        E_ex = 0.0
        V_th = -54e-3
        V_reset = -80e-3

        tau_ex = 10.0e-3

        def a(s):
            return (g_L*E_L+g_ex*s*E_ex)/(g_L+g_ex*s)

        def tau(s):
            return c_m/(g_L+g_ex*s)

        times = [0.1,0.2,0.23,0.3,0.32,0.4,0.41, np.inf]

        def integrate_V(start, end, step_size):
            ti = 0
            steps = int(np.ceil((end-start)/step_size))
            V = np.zeros(steps+1)
            s = np.zeros(steps+1)
            V[0] = E_L # start at E_L
            s[0] = 0 # at V_reset
            step_range = np.array(range(0, steps+1))
            t_range = step_range*step_size
            for i in step_range[1:]:
                t = t_range[i]
                # calc s
                s[i] += s[i-1]*np.exp(-step_size/tau_ex)
                if t >= times[ti]:
                    s[i] += 1
                    ti +=1
            tmp_a = a(s[i])
```

```

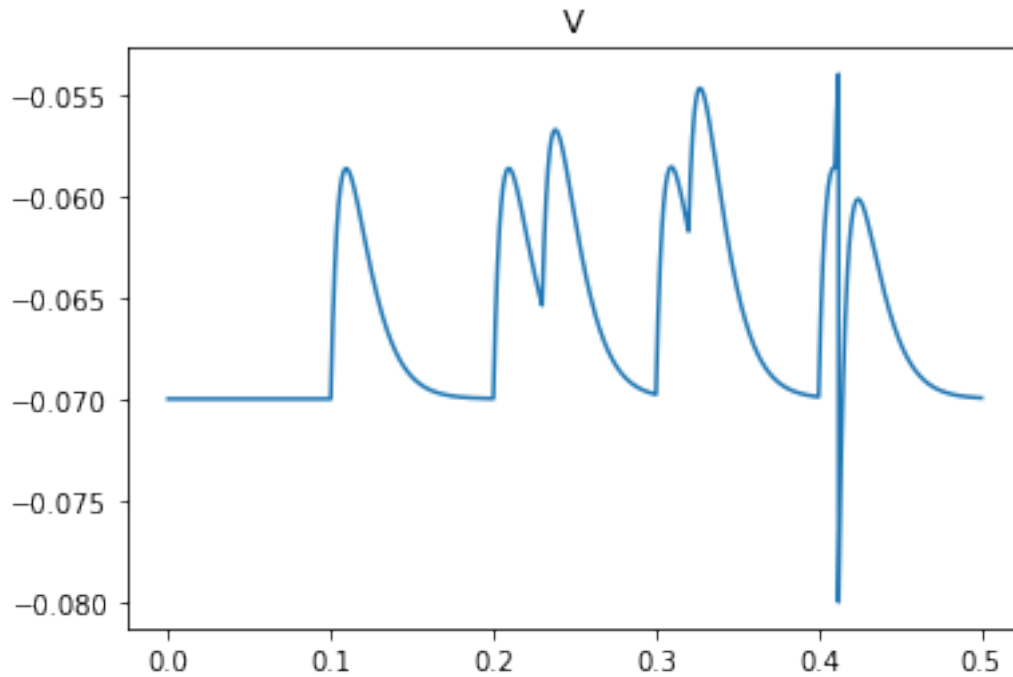
    tmp_tau = tau(s[i])
    V[i] = tmp_a + (V[i-1] - tmp_a)*np.exp(-step_size/tmp_tau)
    if V[i] >= V_th:
        V[i] = V_reset
    I = g_ex*s*(V-E_ex)
    return [t_range, V, I]

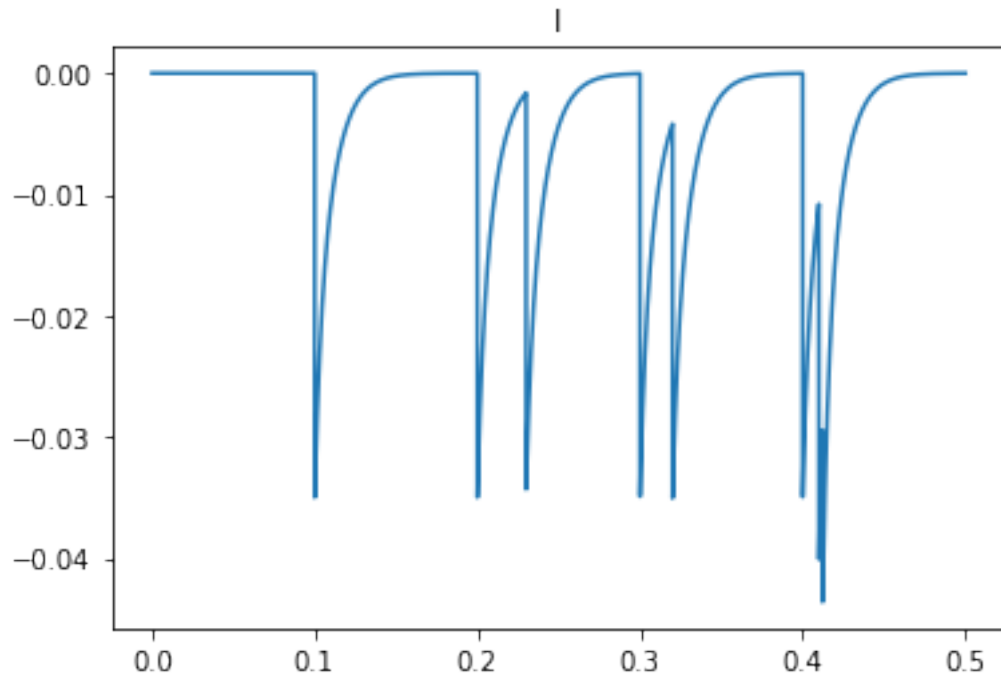
```

```

In [4]: rng, V, I = integrate_V(0.0, 0.5, 0.00001)
plt.plot(rng, V)
plt.title('V')
plt.show()
plt.plot(rng, I)
plt.title('I')
plt.show()

```





optional more advanced problem

In [0]:

Problem 2

a)

In [5]: *## Doing this in milli-seconds and mV this time*

```
tau_m = 10
```

```
E = -56
```

```
V_th = -54
```

```
V_reset = -80
```

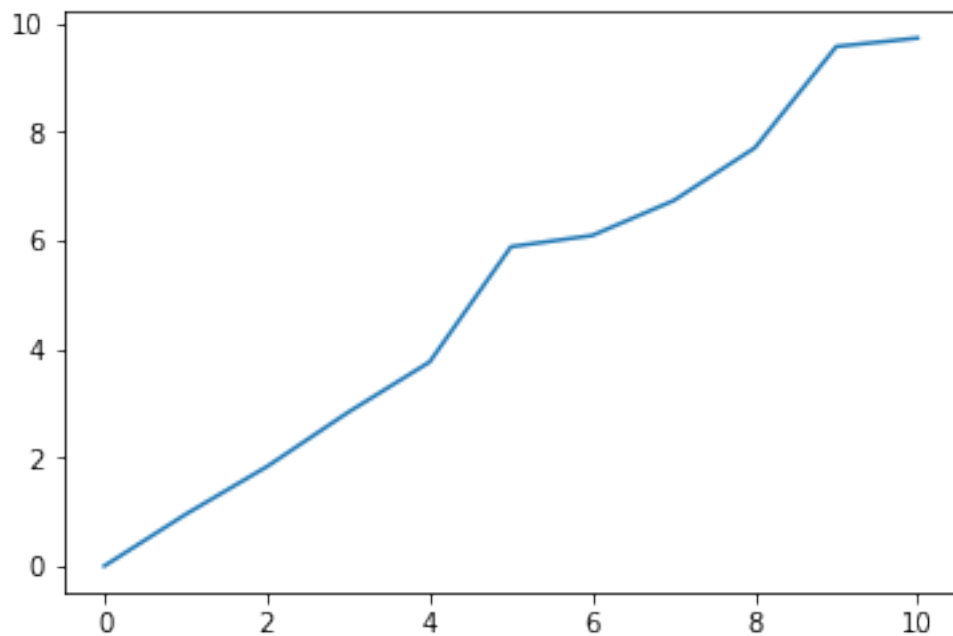
```
def integrate_with_euler(start, end, step_size, sigma_v, I, V_th=V_th):
    steps = int(np.ceil((end-start)/step_size))
    normal_dist = np.random.normal(size=steps)
    D = (sigma_v**2)*tau_m/step_size
    result = np.zeros(steps+1)
    result[0] = E
    step_range = np.array(range(0, steps+1))
    t_range = step_range*step_size
    firing_count = 0
    for i in step_range[1:]:
        t = t_range[i]
```

```

d_V = (E - result[i-1] + np.sqrt(2*D)*normal_dist[i-1] + I)/tau_m
result[i] = result[i-1] + d_V*step_size
if result[i] >= V_th:
    result[i] = V_reset
    firing_count += 1
return [t_range, result, firing_count/(end-start)*1000.0]

devs = []
for sigma_v in range(0, 11):
    _, V, _ = integrate_with_euler(0, 1000.0, 0.1, sigma_v, I=0, V_th=np.inf)
    devs.append(V)
plt.plot(np.arange(0, 11), np.std(devs, axis=1))
plt.show()

```

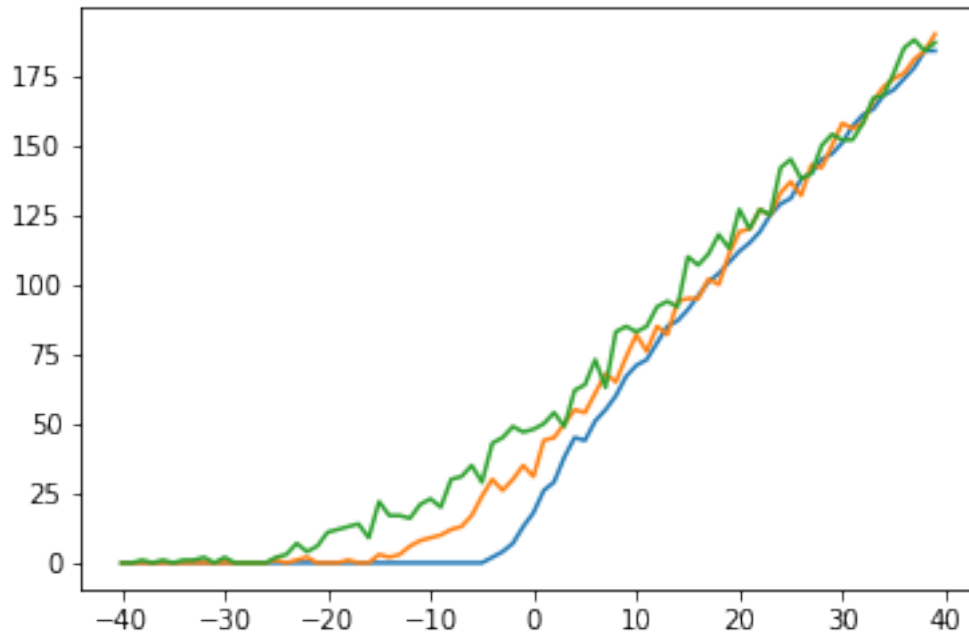


b)

```

In [6]: for sigma_v in [2, 6, 10]:
rang = np.arange(-40, 40)
rates = []
for I in rang:
    _, _, firing_rate = integrate_with_euler(0, 1000.0, 0.1, sigma_v, I=I, V_th=V_th)
    rates.append(firing_rate)
plt.plot(rang, rates)
plt.show()

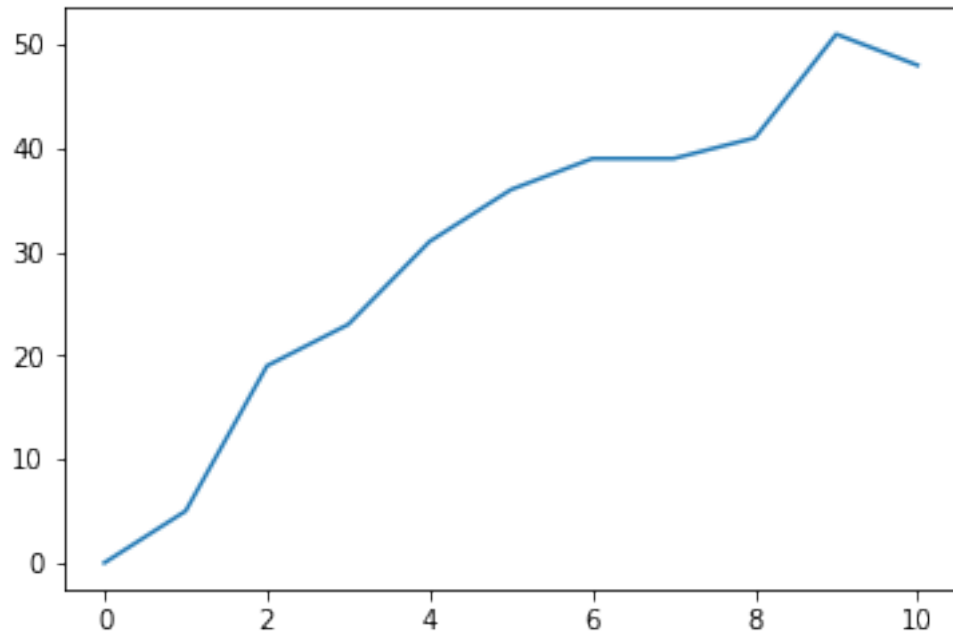
```



In this model, the error doesn't affect the signal for high I.

c)

```
In [7]: rates = []
rang = np.arange(11)
for sigma_v in rang:
    _, _, firing_rate = integrate_with_euler(0, 1000.0, 0.01, sigma_v, I=0, V_th=V_th)
    rates.append(firing_rate)
plt.plot(rang, rates)
plt.show()
```



The firing rate increases as the σ_v increases even though the current is kept at 0.