PSET_2

February 26, 2020

Q1

$$E = E^o + \frac{RT}{nF} ln\Big(\frac{[Ox]}{[R]}\Big) = E^o + \frac{0.059}{n} log_{10}\Big(\frac{[Ox]}{[R]}\Big) \tag{1}$$

$$\tag{2}$$

$$Fe^{3+} + e^- \rightarrow Fe^{2+} \Leftarrow \boxed{\texttt{Gain of electron - Reduction}} \tag{3}$$

$$\tag{4}$$

$$[Re] + e^- \rightarrow [Ox] \tag{5}$$

$$\tag{6}$$

$$Answer\ (both\ valid): \tag{7}$$

$$\boldsymbol{c)\ \ E = E^o + \frac{0.059}{n} log_{10}\Big(\frac{[Fe^{3+}]}{[Fe^{2+}]}\Big)} \tag{8}$$

$$\tag{9}$$

$$\boldsymbol{d)\ \ E = E^o - \frac{0.059}{n} log_{10}\Big(\frac{[Fe^{2+}]}{[Fe^{3+}]}\Big)} \tag{10}$$

Q2

$$E_{probe} = K + 0.059 \cdot log[F^-]$$

$$E_{probe} = V \cdot q \Leftarrow \boxed{\texttt{charge ratiometric, so regarded q as 1}}$$

$$V_0 = K + 0.059 \cdot log[F^-]$$

$$0.112V = K + 0.059 \cdot log[0.01]$$

$$0.112V = K - 0.059 \Leftarrow \boxed{\texttt{K = 0.23V}}$$

$$V_t = K + 0.059 \cdot log[F_t^-]$$

$$0.203V = 0.23V + 0.059 \cdot log[F_t^-]$$

$$-0.027 = 0.059 \cdot log[F_t^-]$$

$$Answer:$$
$$\boldsymbol{[F_t^-] = 0.35M}$$

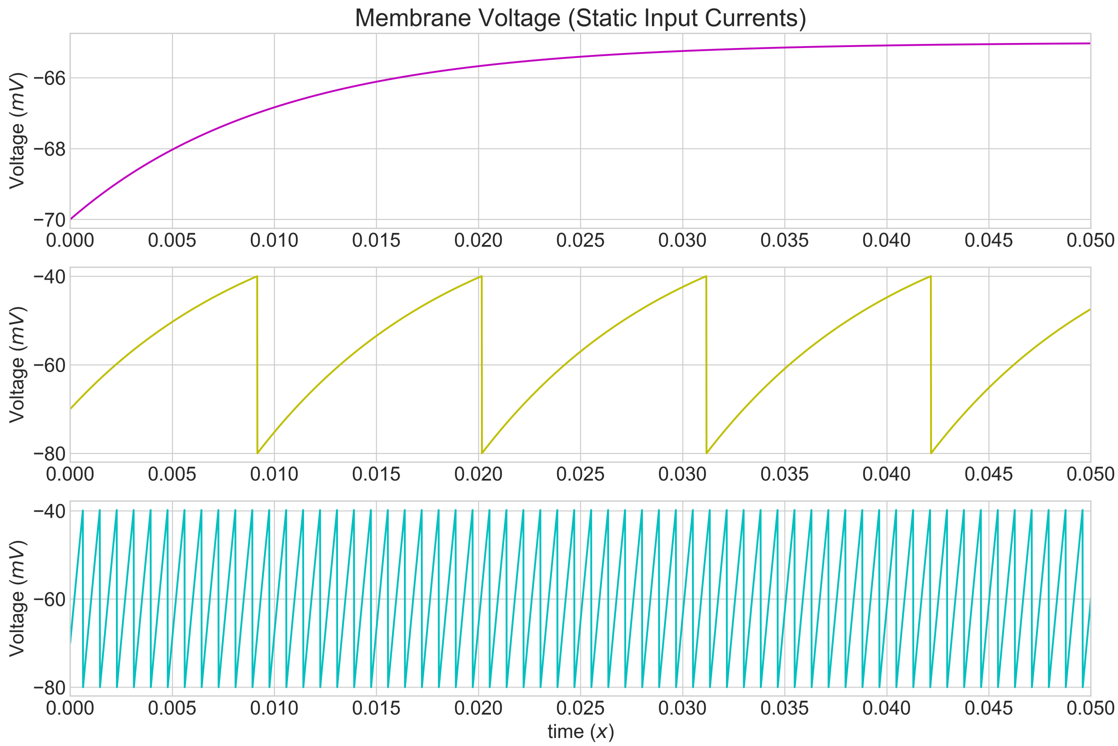
Q3 - Part 1

[7]:
```
import numpy as np
import pylab as plt
from scipy.integrate import odeint

# conversion constants
nA = 1e9; us = 1e-6; MOhms = 1e6; pF = 1e-12; mV = 1e-3; mS = 1e-3; S = 1
pA = 1e-12; nA = 1e-9

# styling
TITLE_FONT_SIZE = 20; LABEL_FONT_SIZE = 16; DPI = 458; FIG_SIZE=(15,10);␣
↪LINE_WIDTH = 3
plt.rcParams.update({'font.size': LABEL_FONT_SIZE})
plt.style.use('seaborn-whitegrid')
character_color = np.asarray(['m', 'y', 'c', 'b', 'g', 'r'])
```

[8]:
```
dt = 10*us # time step
t_max = 1*S # total time
time = np.arange(0, t_max, dt)
R_m = 50*MOhms # Membrane Resistance
C_m = 200*pF # Membrane capacitance
V_e = -70*mV # Equilibrium voltage
```

```
V_reset = -80*mV # Hyperpolarization/reset voltage
V_th = -40*mV # Threshold voltage
Input_Is = np.asarray([100*pA, 1*nA, 10*nA])

def voltageModel(dt, R_m, C_m, V_e, V_reset, V_th, Input_Is):
    V_m = np.zeros(time.shape)
    V_m[0] = V_e

    for i in range(1, V_m.shape[0]):
        dV = dt * (-1/R_m * (V_m[i-1]-V_e) + Input_Is)/C_m
        V_m[i] = V_m[i-1] + dV

        if V_m[i-1] >= V_th:
            V_m[i] = V_reset

    return V_m

plt.figure(1, figsize=FIG_SIZE, dpi=DPI)
for I in range(Input_Is.shape[0]):
    plt.subplot(3, 1, I+1)
    if I == 0:
        plt.title('Membrane Voltage (Static Input Currents)',␣
↪fontsize=TITLE_FONT_SIZE)
    plt.plot(time, voltageModel(dt, R_m, C_m, V_e, V_reset, V_th, Input_Is[I])/
↪mV, character_color[I])
    plt.ylabel('Voltage ($mV$)')
    plt.xticks(np.arange(0, t_max, 5*mS), fontsize=LABEL_FONT_SIZE)
    plt.xlim((0,50*mS))

plt.xlabel('time ($x$)')
```

[8]:
```
Text(0.5, 0, 'time ($x$)')
```

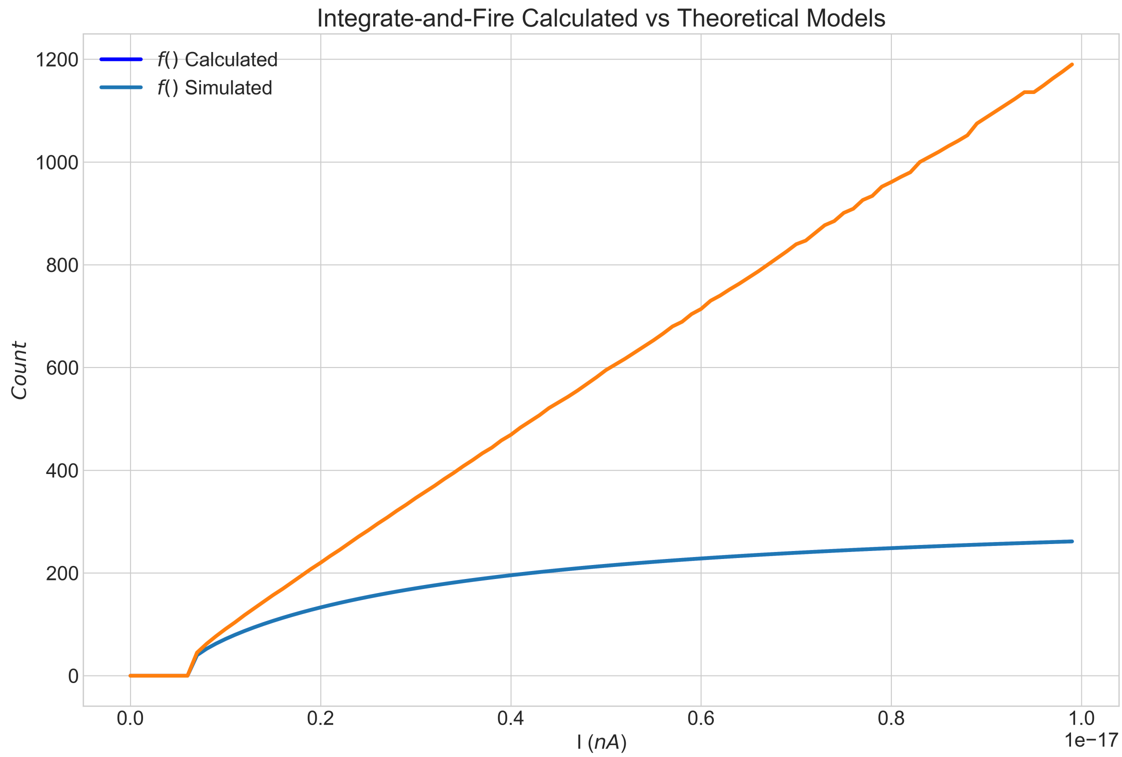


Q3 - Part 2

[9]:
```
T_ref = 3*mS # Refractory time
R_m = 50*MOhms # Membrane Resistance
C_m = 200*pF # Membrane capacitance
V_e = -70*mV # Equilibrium voltage
V_reset = -80*mV # Hyperpolarization/reset voltage
V_th = -40*mV # Voltage threshold

def frequencyModel(T_ref, R_m, C_m, V_e, V_reset, V_th, I):
    Ith = (V_th - V_e)/R_m
    if I < Ith:
        return 0

    f = (T_ref + R_m*C_m*np.log((I*R_m+V_e-V_reset)/(I*R_m+V_e-V_th)))**-1

    return f

I = np.arange(0, 10*nA, 100*pA)
f_calc = np.zeros(I.shape[0])

for i in range(f_calc.shape[0]):
    f_calc[i] = frequencyModel(T_ref, R_m, C_m, V_e, V_reset, V_th, I[i])
```

```
plt.figure(1, figsize=FIG_SIZE, dpi=DPI)
plt.title('Integrate-and-Fire Calculated vs Theoretical Models',␣
 ↪fontsize=TITLE_FONT_SIZE)
plt.plot(I*nA,f_calc, character_color[3], linewidth=LINE_WIDTH)
plt.xlabel("I ($nA$)")
plt.ylabel('Membrane Voltage (mV)')
plt.xticks(fontsize=LABEL_FONT_SIZE)

dt = 10*us
t_max = 1*S
time = np.arange(0, t_max, dt)

def spikeIFModel(dt, R_m, C_m, V_e, V_reset, V_th, Input_Is):
    V_m = np.zeros(time.shape)
    Spike_count = 0
    V_m[0] = V_e
    # Let's figure out dV
    for i in range(1, V_m.shape[0]):
        dV = dt * (-1/R_m * (V_m[i-1]-V_e) + Input_Is)/C_m
        V_m[i] = V_m[i-1] + dV

        if V_m[i-1] >= V_th:
            V_m[i] = V_reset
            Spike_count += 1

    return Spike_count

f_sim = np.zeros(I.shape)
for i in range(I.shape[0]):
    f_sim[i] = spikeIFModel(dt, R_m, C_m, V_e, V_reset, V_th, I[i])

plt.figure(1, figsize=FIG_SIZE, dpi=DPI)
plt.plot(I*nA, f_calc, linewidth=LINE_WIDTH)
plt.plot(I*nA, f_sim, linewidth=LINE_WIDTH)
plt.title('Integrate-and-Fire Calculated vs Theoretical Models',␣
 ↪fontsize=TITLE_FONT_SIZE)
plt.xlabel("I ($nA$)")
plt.ylabel("$Count$")
plt.legend(['$f()$ Calculated', '$f()$ Simulated'], loc="best")
```

```
[9]: <matplotlib.legend.Legend at 0x243ec34f048>
```



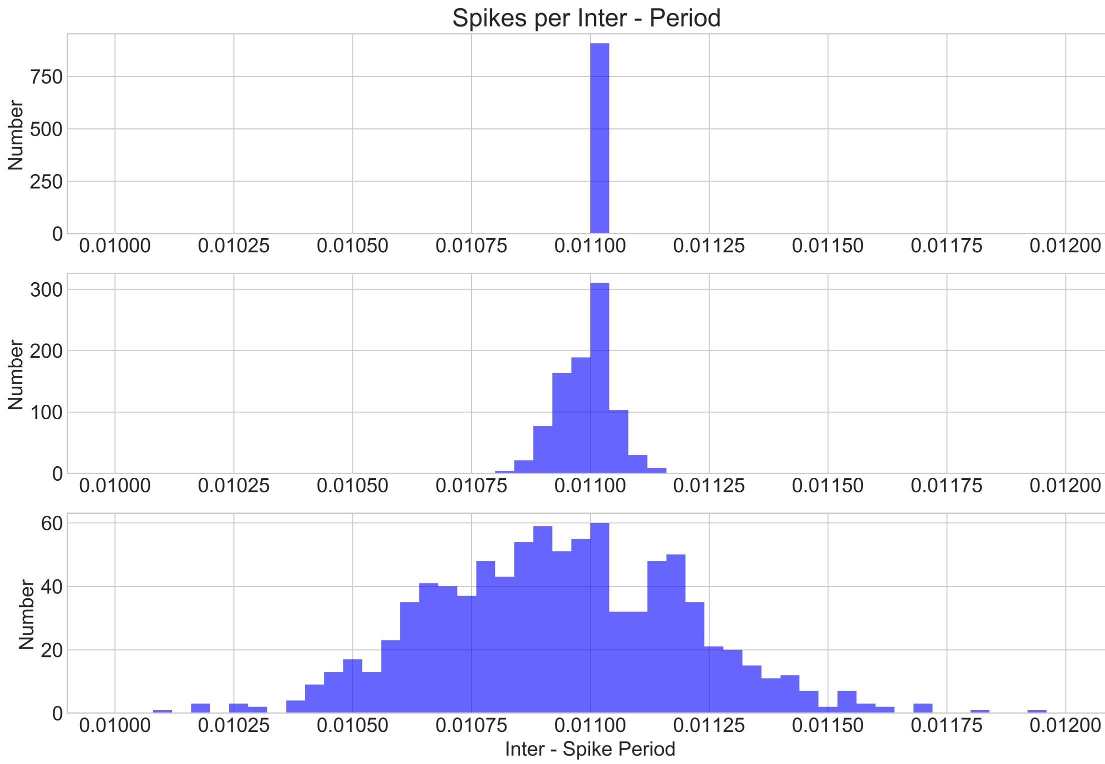
Though in the current resolution on the graph these may not seem to closely resemble eachother, if you zoom into the initial changes in count versus current the calculated and simulated present great identity to eachother. This is further examined by eliminating the differentiating factor T_ref by dropping the value to 0 induces these curves to match eachother, showing introducing a refractory period improves modeling of biological models. Nevertheless, the calculated and simulated frequency models converge greatly.

Q3 - Part 3

[10]:
```
dt = 10*us # Time step
t_max = 10*S # Total time
time = np.arange(0, t_max, dt) # Time array
T_ref = 3*mS # Refactory time
R_m = 50*MOhms #Membrane Resistance
C_m = 200*pF # Membrane capacitance
V_e = -70*mV # Equilibrium voltage
V_reset = -80*mV # Hyperpolarization/reset voltage
V_th = -40*mV # Threshold voltage

def gaussSpikeIFModel(dt, R_m, C_m, V_e, V_reset, V_th, Isigma):
    V_m = np.zeros(time.shape)
    Imean = 1000*pA
    Inoisy = np.random.normal(Imean, Isigma, (time.shape[0], 1))
    Spike_count = 0
```

```
    V_m[0] = V_e

    spikes = []
    last_spike_idx = 0
    for i in range(1, V_m.shape[0]):
        dV = dt* (-1/R_m * (V_m[i-1]-V_e) + (Inoisy[i] if Inoisy[i] > 0 else␣
↪0))/C_m
        V_m[i] = V_m[i-1] + dV

        if V_m[i-1] >= V_th:
            V_m[i] = V_reset
            spikes.append((i-last_spike_idx)*dt)
            last_spike_idx = i

    return np.asarray(spikes) # inter-spike periods

Input_Is = np.asarray([0, 100*pA, 500*pA])

plt.figure(1, figsize=FIG_SIZE, dpi=DPI)
for I in range(Input_Is.shape[0]):
    plt.subplot(3, 1, I+1)
    if I == 0:
         plt.title("Spikes per Inter - Period ", fontsize=TITLE_FONT_SIZE)
    spikecounts = gaussSpikeIFModel(dt, R_m, C_m, V_e, V_reset, V_th,␣
↪Input_Is[I])
    plt.hist(spikecounts, bins=50, range=(0.010, 0.012),␣
↪color=character_color[3], alpha=0.6)
    plt.ylabel("Number")

plt.xlabel("Inter - Spike Period")
```

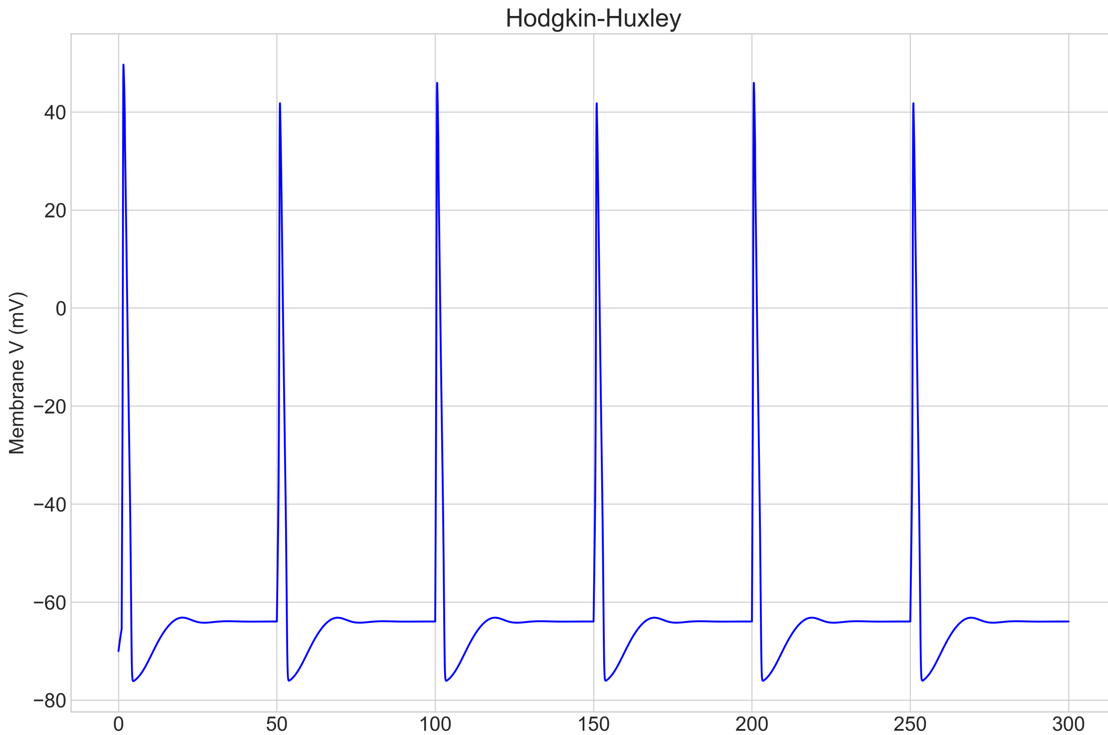
```
[10]: Text(0.5, 0, 'Inter - Spike Period')
```

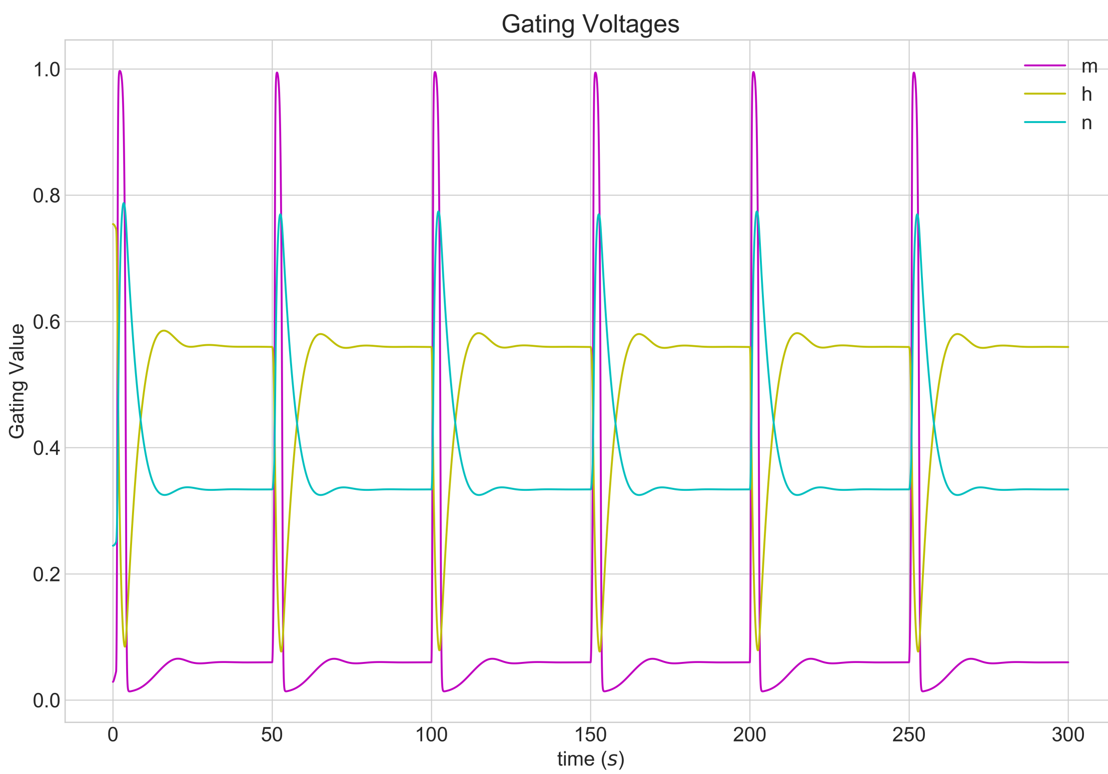


[11]:
```
Alpha_n = lambda V_m:0.01*(V_m+55)/(1-np.exp(-(V_m+55)/10))
Alpha_m = lambda V_m: 0.1*(V_m+40)/(1-np.exp(-(V_m+40)/10))
Alpha_h = lambda V_m: 0.07*np.exp(-(V_m+65)/20)

Beta_n = lambda V_m: 0.125*np.exp(-(V_m+65)/80)
Beta_m = lambda V_m: 4*np.exp(-(V_m+65)/18)
Beta_h = lambda V_m: 1/(1+np.exp(-(V_m+35)/10))

# Set up constants
Cm = 1
gk = 36
gna = 120
gl = 0.3
Vna = 50
Vk = -77
Vl = -50
V_e = -70

# I = Cdv/dt + Ik + Ina + Il
# dV/dt = 1/C_m * (I- (Ik + Ina + Il))
Ik = lambda gk, n, V_m, Vk: gk*n**4*(V_m-Vk)
Ina = lambda gna, m, h, V_m, Vna: gna*m**3*h*(V_m-Vna)
Igl = lambda gl, V_m, Vl: gl*(V_m-Vl)
```

```
m_inf = Alpha_m(V_e)/(Alpha_m(V_e)+Beta_m(V_e))
h_inf = Alpha_h(V_e)/(Alpha_h(V_e)+Beta_h(V_e))
n_inf = Alpha_n(V_e)/(Alpha_n(V_e)+Beta_n(V_e))

dt = 1*mS
time = np.arange(0,300*S,dt)

def Vm_eq(X, time):
    V_m, m, h, n = X
    I = 150.0*((time>1) - (time > 2)) + 50 * ((time > 50) - (time>51)) + 150.
 ↪0*((time>100) - (time > 101)) + 50 * ((time > 150) - (time>151))+ 150.
 ↪0*((time > 200) - (time>201)) + 50 * ((time > 250) - (time>251))+ 150.
 ↪0*((time>300) - (time > 301)) + 50 * ((time > 350) - (time>351))
    dVdt = 1/Cm * (I-Ik(gk, n, V_m, Vk) - Ina(gna, m, h, V_m, Vna) - Igl(gl,␣
 ↪V_m, Vl))
    # determine gating values
    dmdt = Alpha_m(V_m)*(1-m)-Beta_m(V_m)*m
    dhdt = Alpha_h(V_m)*(1-h)-Beta_h(V_m)*h
    dndt = Alpha_n(V_m)*(1-n)-Beta_n(V_m)*n
    return dVdt, dmdt, dhdt, dndt

def HodgkinHuxley(time):
    s = odeint(Vm_eq,  [V_e,  m_inf, h_inf, n_inf], time)
    return s

plt.figure(1, figsize=FIG_SIZE, dpi=DPI)
plt.title('Hodgkin-Huxley', fontsize=TITLE_FONT_SIZE)
plt.plot(time, HodgkinHuxley(time)[:,0], character_color[3])
plt.ylabel('Membrane V (mV)')

plt.figure(2, figsize=FIG_SIZE, dpi=DPI)
plt.plot(time, HodgkinHuxley(time)[:,1], character_color[0])
plt.plot(time, HodgkinHuxley(time)[:,2], character_color[1])
plt.plot(time, HodgkinHuxley(time)[:,3], character_color[2])
plt.legend(['m', 'h', 'n'], loc="best")
plt.title('Gating Voltages', fontsize=TITLE_FONT_SIZE)
plt.xlabel('time ($s$)')
plt.ylabel('Gating Value')
```

[11]:
```
Text(0, 0.5, 'Gating Value')
```


Hodgkin-Huxley
Membrane V (mV)
40
20
0
−20
−40
−60
−80
0
50
100
150
200
250
300


Gating Voltages
m
h
n
Gating Value
1.0
0.8
0.6
0.4
0.2
0.0
0
50
100
150
200
250
300
time (s)