

Spiking Neuron Models

Problem 1: Leaky Integrate and Fire Model

(a) At steady state,

$$C \frac{dV(t)}{dt} = -g_L(V(t) - E_L) + I_{app}(t) = 0 \quad (1)$$

Hence,

$$V_s = \frac{I_0}{g_L} + E_L \quad (2)$$

A spike is issued only if $V_s \geq V_T$. Hence, Hence,

$$I_c = g_L(V_T - E_L) \quad (3)$$

Conductance is given in units of nS and voltages in units of mV, the units of the current will be pA. Plugging in the numbers, we get $I_c = 2700$ pA.

(b) Matlab function is below:

```

[xa,V]=simulate_lif(N,T,dt,Iapp)

iter=ceil(T/dt);
C=300;
gL=30;
EL=-70;
VT=20;

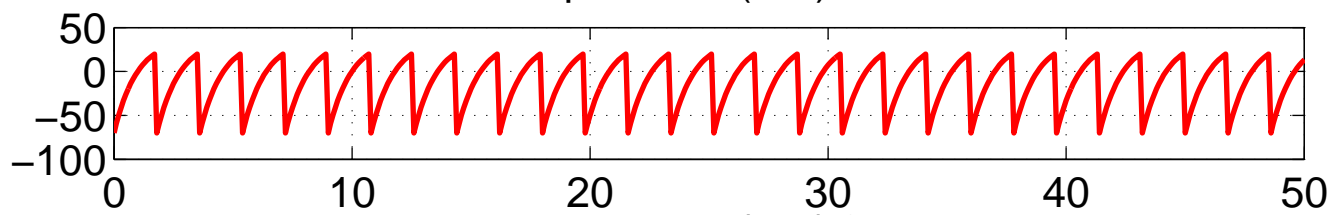
Ic=gL*(VT-EL);
iter=ceil(T/dt);
V=EL*ones(N,iter);

j=1; xa(1)=0;

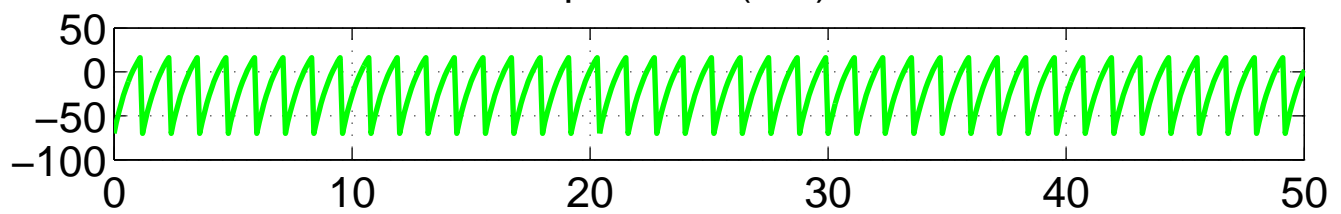
for j=1:iter-1
    V(:,j+1)=V(:,j)+ (dt/C)* (Iapp(:,j) - (gL*(V(:,j)-EL)));
    spind=sign(V(:,j+1)-VT);
    if max(spind)>0
        ind=find(spind>0); %indices of neuron that just spiked
        V(ind,j+1)=EL;
    end
    xa(j+1)=xa(j)+dt;
end
end

```

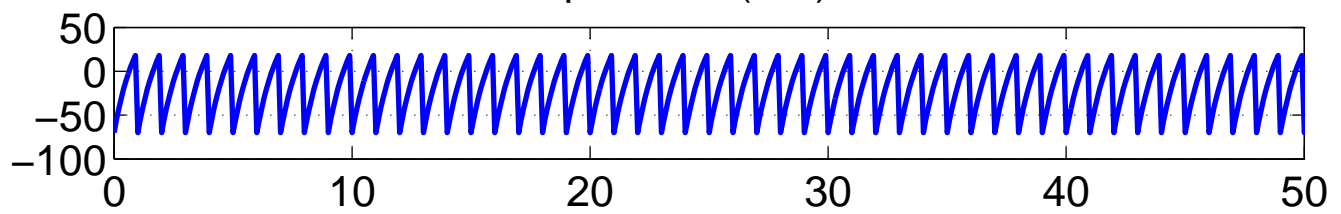
Membrane potential (mV) for neuron 2



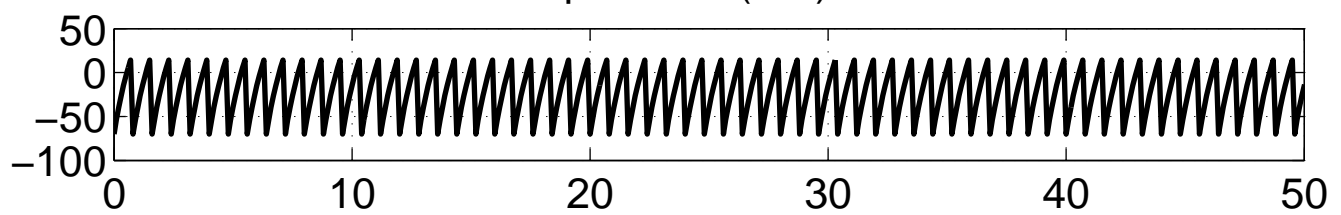
Membrane potential (mV) for neuron 4

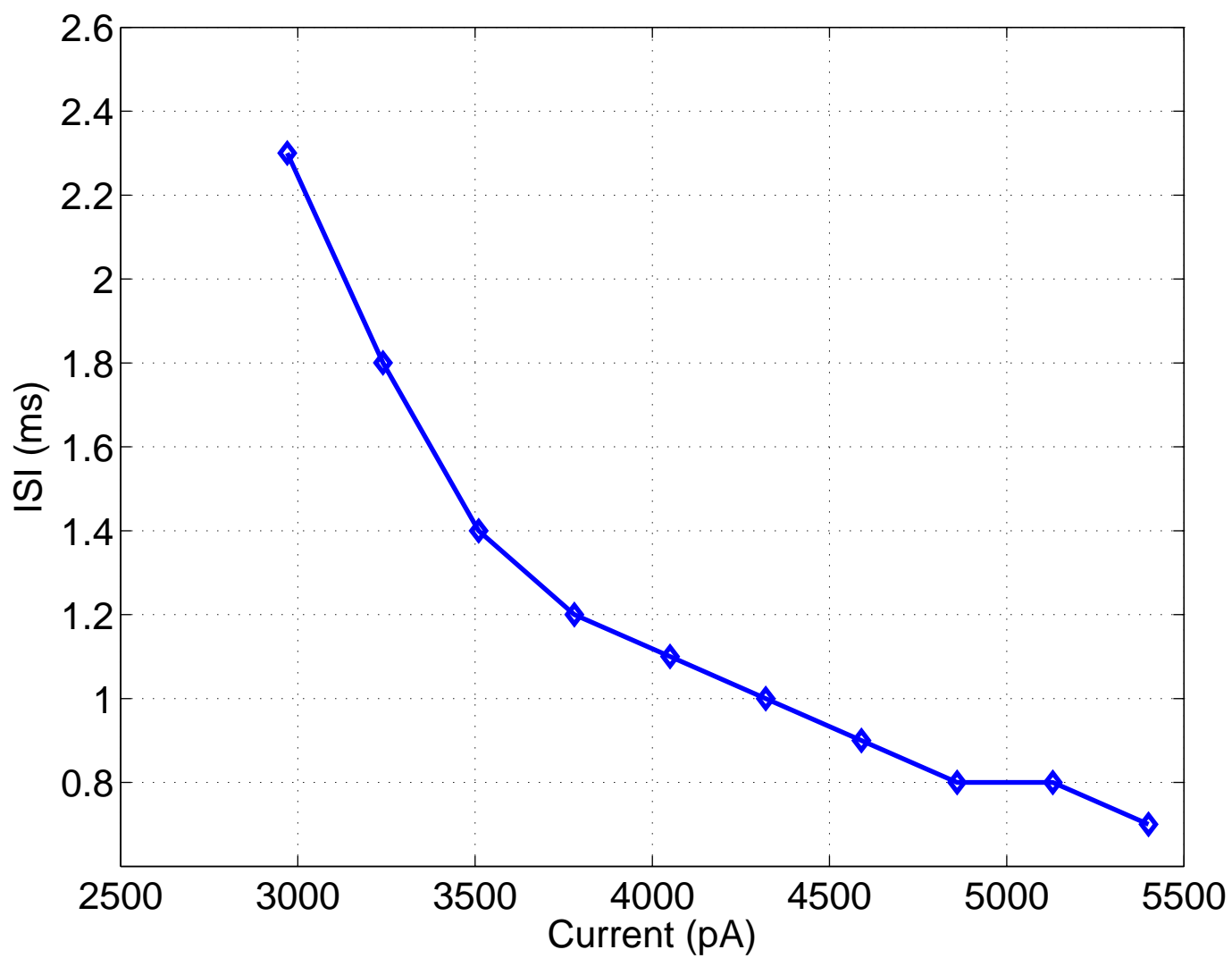


Membrane potential (mV) for neuron 6



Membrane potential (mV) for neuron 8





Problem 2: Izhikevich Model

(a) The steady state values are given by

$$0 = k_z(V_s - E_r)(V_s - E_t) - U_s \quad (4)$$

$$0 = a[b(V_s - E_r) - U_s] \quad (5)$$

Clearly, $V_s = E_r, U_s = 0$ is a solution. Also,

$$V_s = \frac{b}{k_z} + E_t \quad (6)$$

$$U_s = b\left(\frac{b}{k_z} + E_t - E_r\right) \quad (7)$$

is a solution.

(b) The equivalent difference equations are

$$C \frac{V(j+1) - V(j)}{\Delta t} = k_z(V(j) - E_r)(V(j) - E_t) - U(j) + I_{app}(j) \quad (8)$$

$$\frac{U(j+1) - U(j)}{\Delta t} = a[b(V(j) - E_r) - U(j)] \quad (9)$$

(c) The key in obtaining the correct behavior in these simulations is to have consistent set of units. One way to ensure this is to do a dimensional analysis check. With the numbers as defined below, every term in (8) has the units of pA and every term in (9) has the units of nA/s.

```
function [xa,V]=simulate_iz(N,T,dt,Iapp,type)
iter=ceil(T/dt);
```

```
if type==1
    st=['RS'];
    C=100;
    gz=0.7;
    Er=-60;
    Et=-40;
    a=0.03;
    b=-2;
    c=-50;
    d=100;
    vpeak=35;
elseif type==2
    st=['IB'];
    C=150;
    gz=1.2;
    Er=-75;
    Et=-45;
    a=0.01;
```

```

        b=5;
        c=-56;
        d= 130;
        vpeak=50;
    else
        st=['CH'];
        C=50;
        gz=1.5;
        Er=-60;
        Et=-40;
        a=0.03;
        b=1;
        c=-40;
        d=150;
        vpeak=25;
    end

    iter=ceil(500/dt);
    V=Er*ones(N,iter);
    U=0*ones(N,iter);

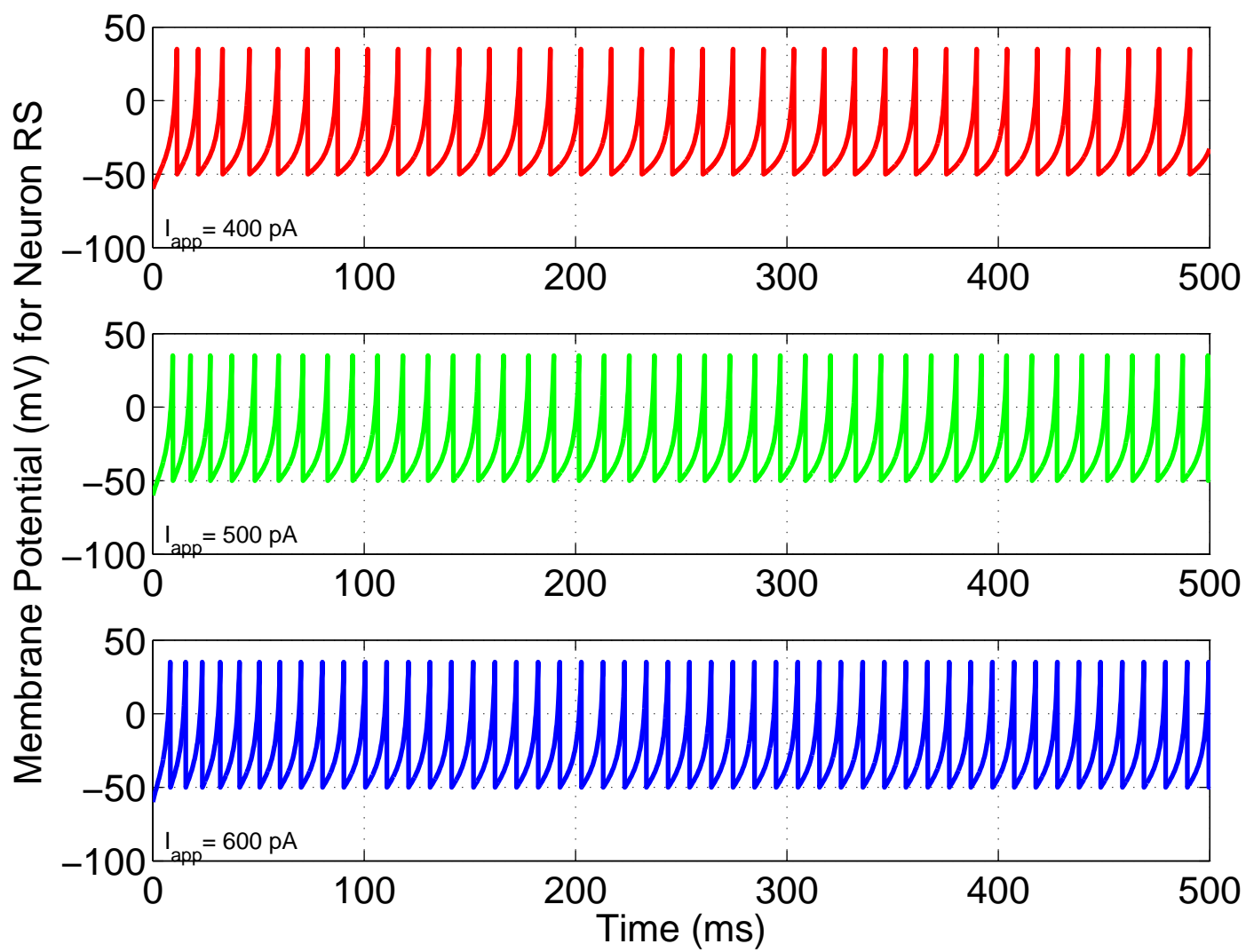
    j=1; xa(1)=0;

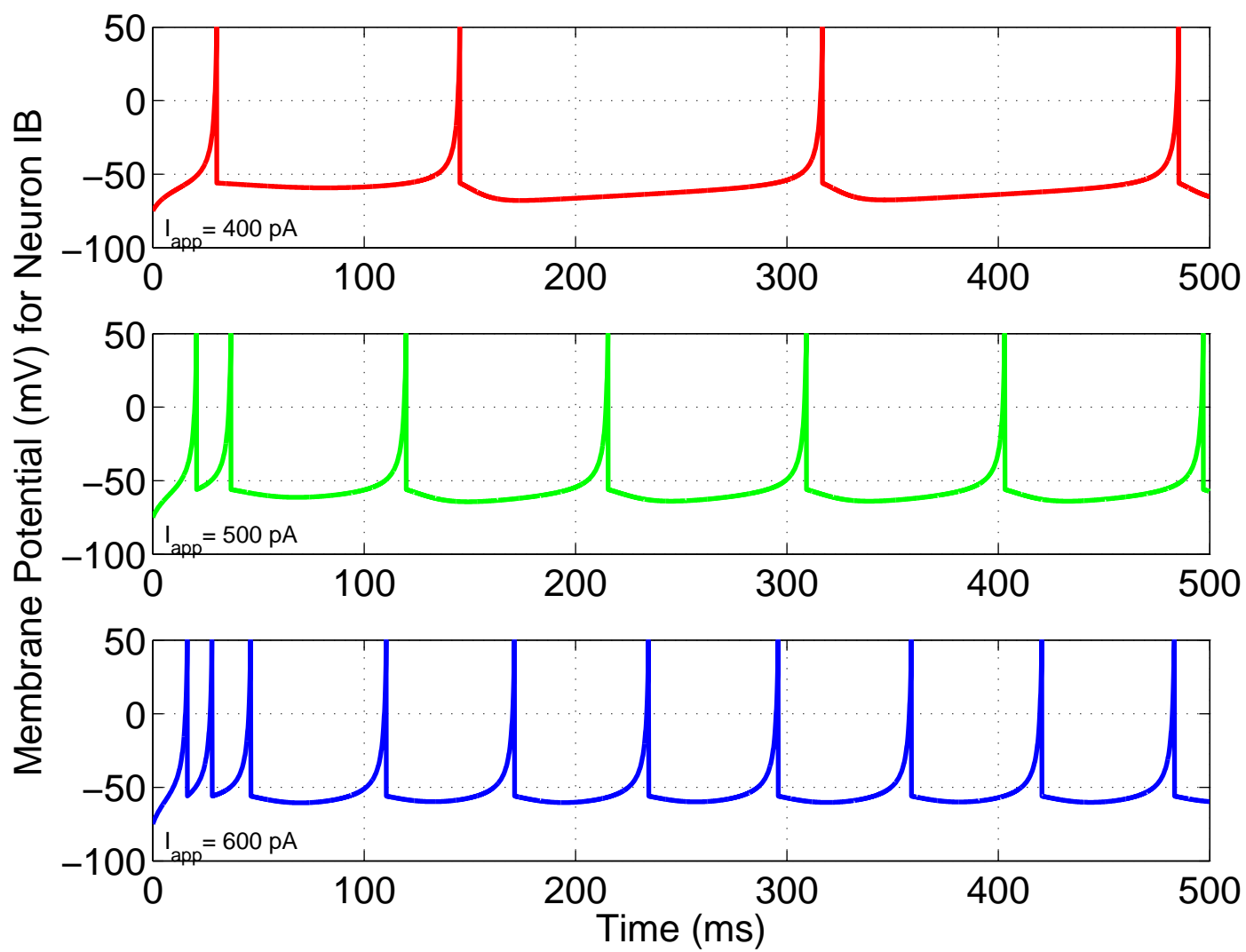
    for j=1:iter
        V(:,j+1)=V(:,j)+ (dt/C)* (Iapp(:,j) + (gz*(V(:,j)-Er).*(V(:,j)-Et)) - U(:,j));
        U(:,j+1)=U(:,j)+ (a*dt)* (-U(:,j) + (b*(V(:,j)-Er)) );

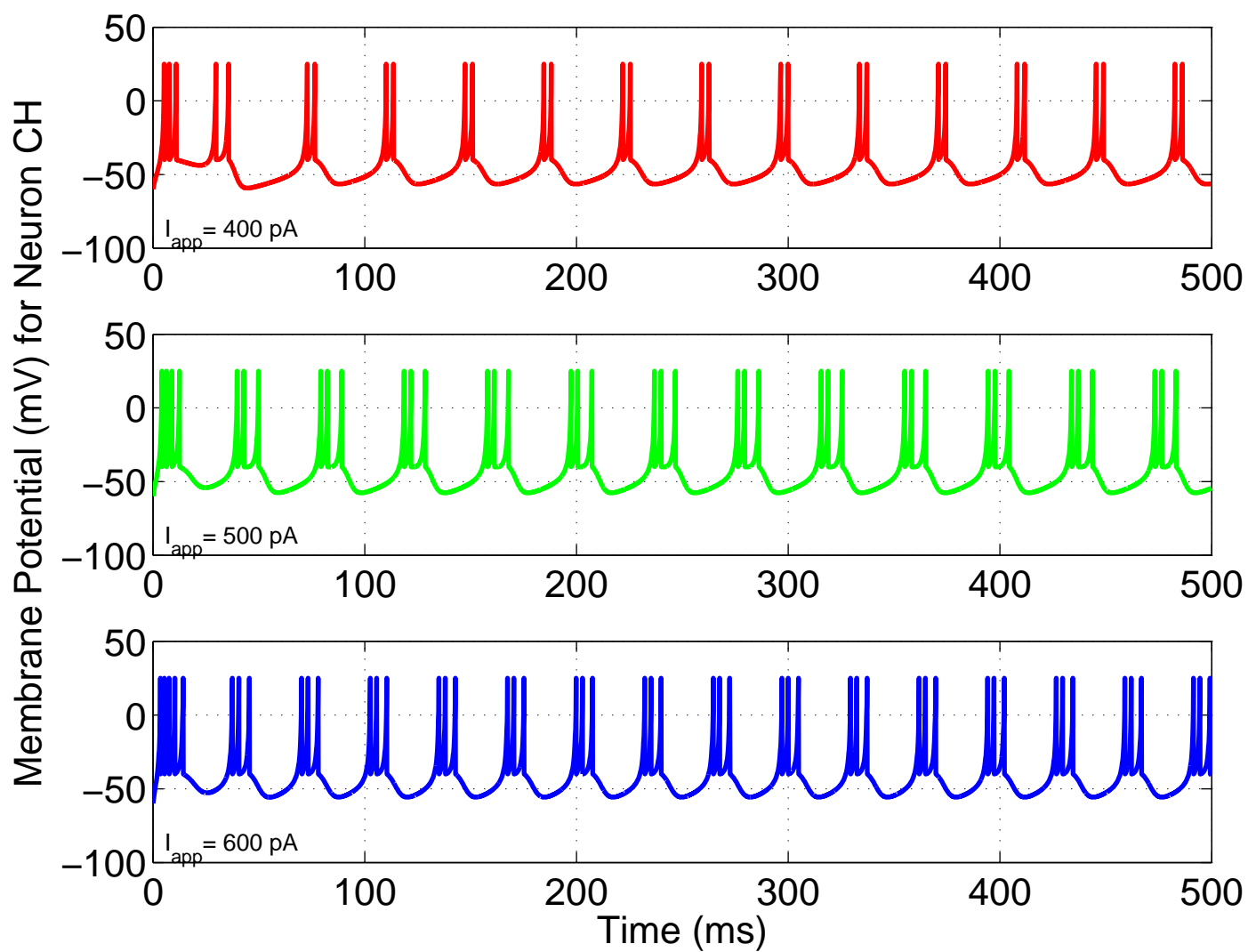
        spind=sign(V(:,j+1)-vpeak);
        if max(spind)>0
            ind=find(spind>0); %indices of neuron that just spiked
            V(ind,j)=vpeak;
            V(ind,j+1)=c;
            U(ind,j+1)=U(ind,j)+d;
        end
        xa(j+1)=xa(j)+dt;
    end

end

```







Problem 3: Adaptive Exponential Integrate-and-Fire Model

(a) The equivalent difference equations are

$$C \frac{V(j+1) - V(j)}{\Delta t} = -g_L(V(j) - E_L) + g_L \Delta_T \exp\left(\frac{V(j) - V_T}{\Delta_T}\right) - U(j) + I_{app}(j) \quad (10)$$

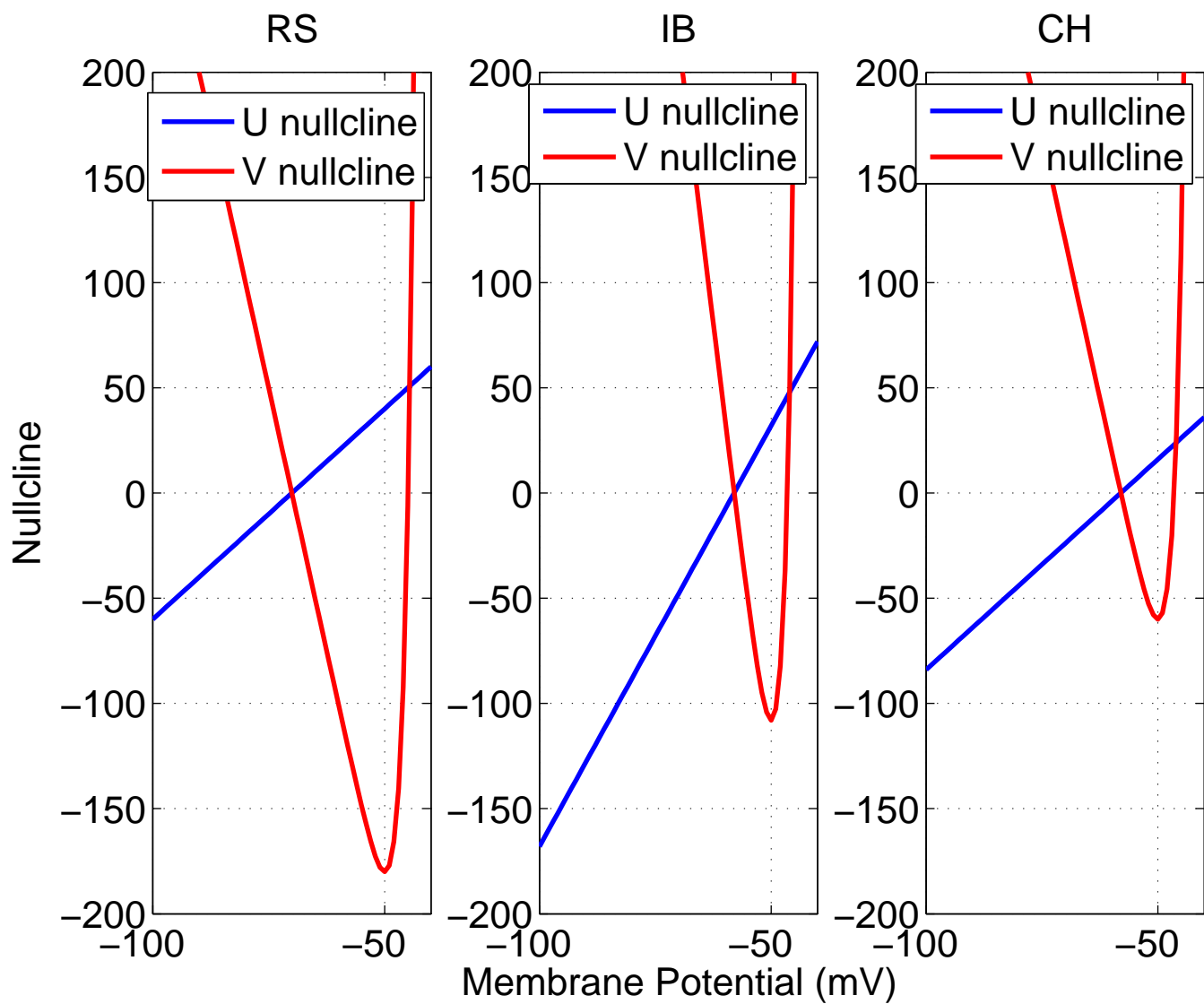
$$\tau_w \frac{U(j+1) - U(j)}{\Delta t} = a [V(j) - E_L] - U(j) \quad (11)$$

(b) To determine the steady state values of V and U for $I_{app} = 0$, we have to solve the following equations consistently.

$$U_s = -g_L(V_s - E_L) + g_L \Delta_T \exp\left(\frac{V_s - V_T}{\Delta_T}\right) \quad (12)$$

$$U_s = a [V_s - E_L] \quad (13)$$

The plot of U_s vs V_s for the two equations is given below. The points of intersections of the two lines are the critical points to the two differential equations. The resting potential is the intersection point where the two lines have opposite slopes. The other point of intersection is the critical point where an action potential is initiated.



```

function [Vs,Us]=aef_steady(ntype)

if ntype==1
    gL=10;
    EL=-70;
    Dt=2;
    a=2;
elseif ntype==2
    gL=18;
    EL=-58;
    Dt=2;
    a=4;
else
    gL=10;
    EL=-58;
    Dt=2;
    a=2;
end

v1=-50; v2=-100;

diff=abs(v2-v1); j=1;

while diff>1e-6
diff=abs(v2-v1);

    j=j+1;
    vm(j)=(v1+v2)/2;
    dif1=a*(v1-EL) - ((gL*Dt*exp((v1-Vt)/Dt)) - (gL*(v1-EL))) ;
    dif2=a*(v2-EL) - ((gL*Dt*exp((v2-Vt)/Dt)) - (gL*(v2-EL))) ;
    difm(j)=a*(vm(j)-EL) - ((gL*Dt*exp((vm(j)-Vt)/Dt)) - (gL*(vm(j)-EL))) ;

    if difm(j)*dif2>0
        v2=vm(j);
    else
        v1=vm(j);
    end
end

end

Vs=(v1+v2)/2;
Us=a*(Vs-EL);

```

From this, we obtain the steady state solutions as

RS: $V_s = -69.9999\text{mV}$, $U_s = 1.5102e - 004\text{mV}$

IB: $V_s = -57.9696\text{mV}$, $U_s = 0.1217\text{mV}$

CH: $V_s = -57.9690\text{mV}$, $U_s = 0.0620\text{mV}$

The other critical points are

RS: $V_s = -44.5481\text{mV}$, $U_s = 50.9039\text{mV}$

IB: $V_s = -46.0182\text{mV}$, $U_s = 47.9273\text{mV}$

CH: $V_s = -46.0622\text{mV}$, $U_s = 23.8755\text{mV}$

(c)

```
function [xa,V]=simulateaef(Iapp,T,dt,type)
N=1;
vpeak=0;
iter=T/dt;
if type==1
    C=200;
    gL=10;
    EL=-70;
    Vt=-50;
    Dt=2;
    a=2;
    tauw=30;
    b= 0 ;
    Vr=-58;
elseif type==2
    C=130;
    gL=18;
    EL=-58;
    Vt=-50;
    Dt=2;
    a=4;
    tauw=150;
    b= 120 ;
    Vr=-50;
else
    C=200;
    gL=10;
    EL=-58;
    Vt=-50;
    Dt=2;
    a=2;
    tauw=120;
    b= 100 ;
    Vr=-46;
end

[Vi,Ui]=aef_steady(type);
V=Vi*ones(N,iter);
U=Ui*ones(N,iter);
```

```

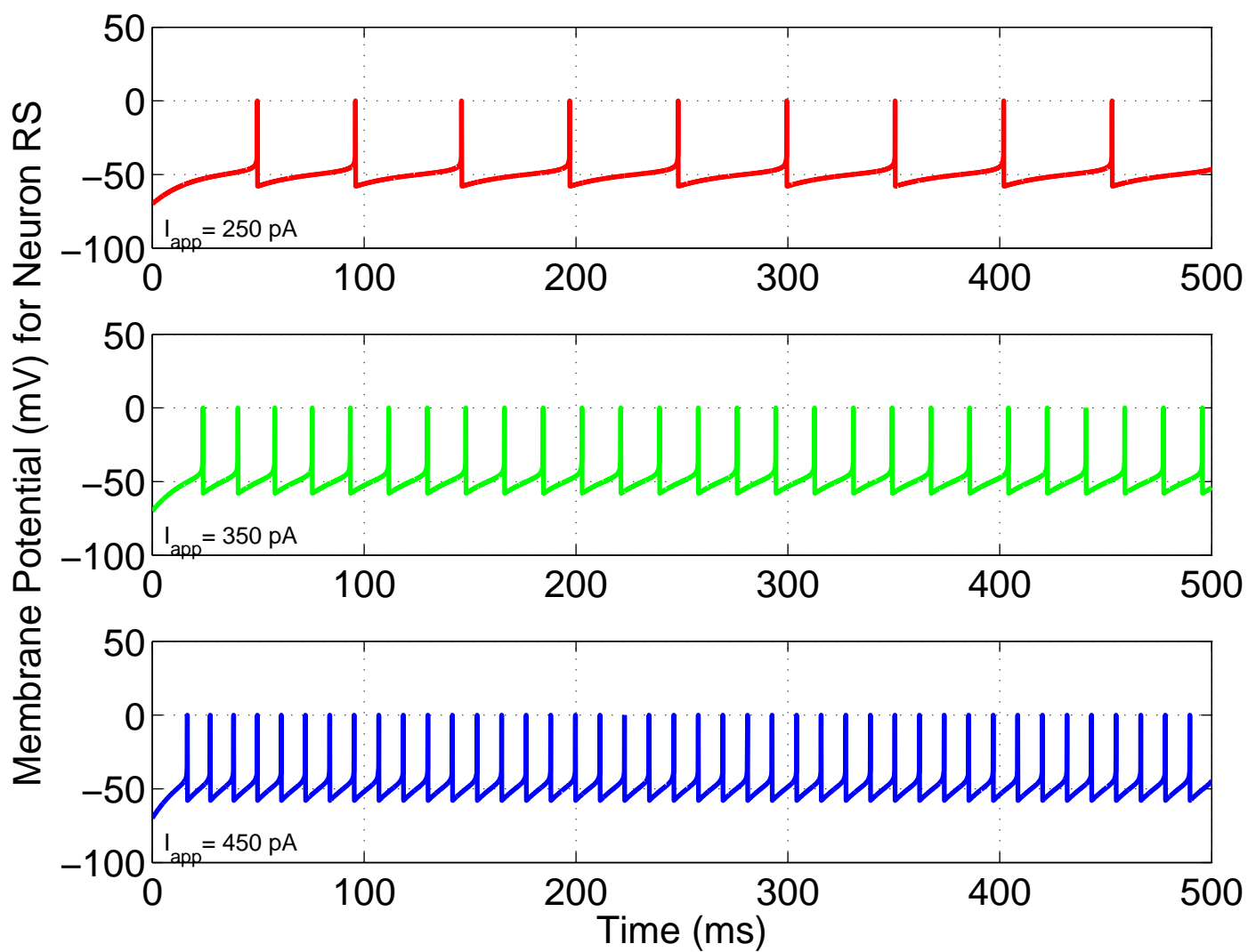
xa(1)=0; j=1;

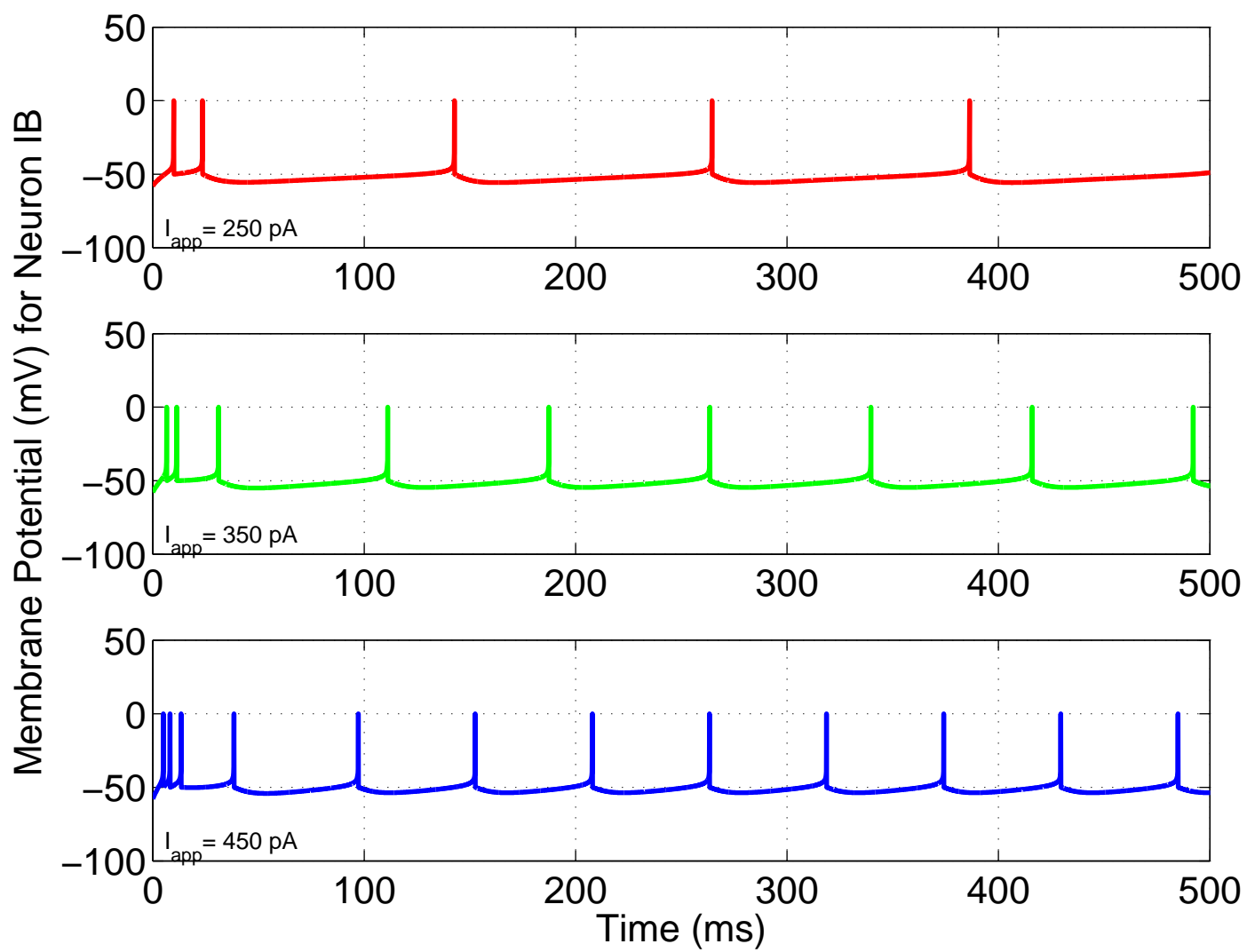
while max(xa)< T-dt
    V(:,j+1)=V(:,j)+(dt/C)*(Iapp(:,j)+(gL*Dt*exp((V(:,j)-Vt)/Dt))
        -(gL*(V(:,j)-EL))-U(:,j));
    U(:,j+1)=U(:,j)+(dt/tauw)*(-U(:,j)+(a*(V(:,j)-EL)));

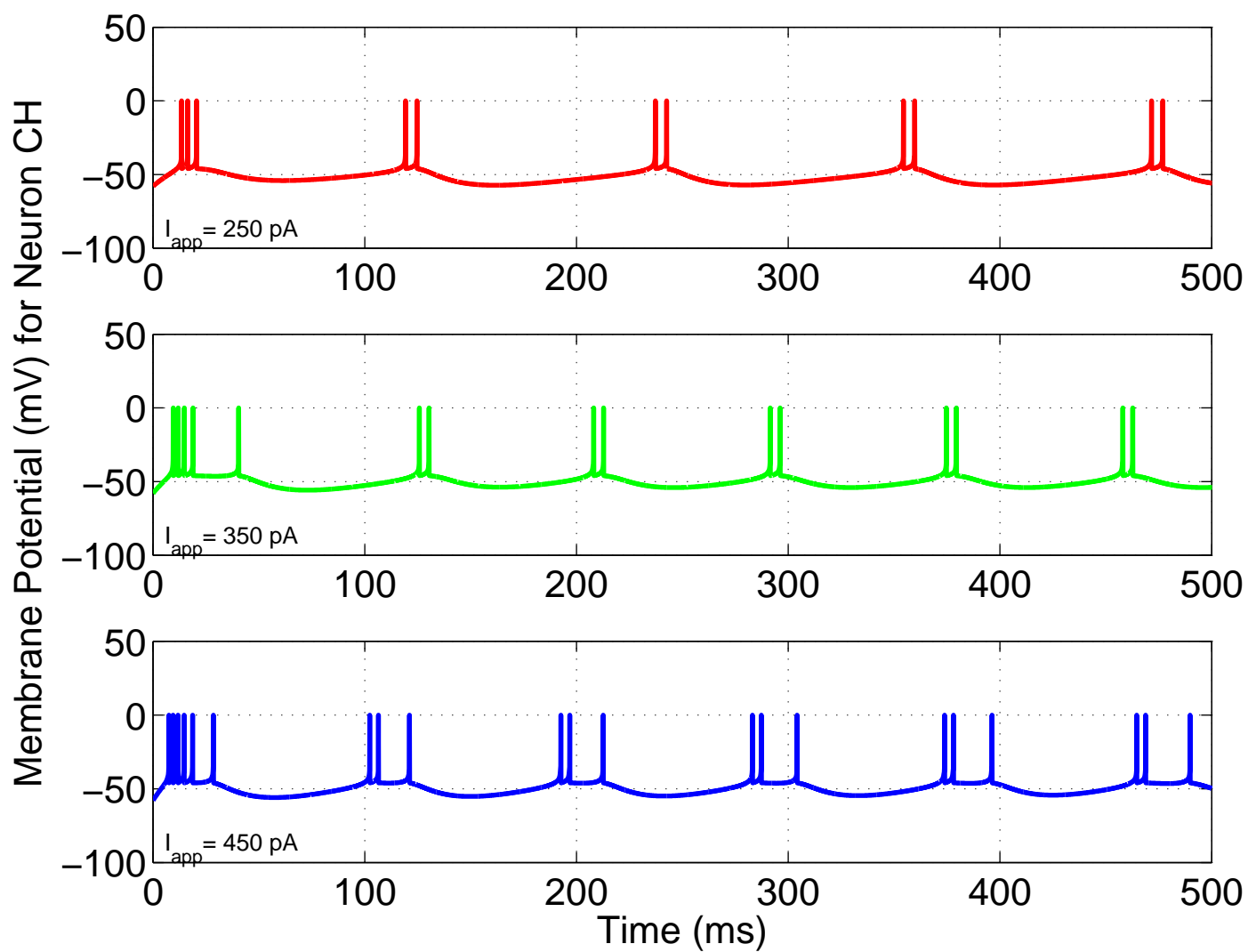
    spind=sign(V(:,j+1)-vpeak);
    if max(spind)>0
        ind=find(spind>0); %indices of neuron that just spiked
        V(ind,j)=vpeak;
        V(ind,j+1)=Vr;
        U(ind,j+1)=U(ind,j)+b;
    end

    xa(j+1)=xa(j)+dt;
    j=j+1;
end

```







Problem 4: Spike energy based on Hodgkin-Huxley neuron model

(a) The code used to calculate the membrane potential and power is below. We will assume a typical cell area of $10^4 \mu\text{m}^2$:

```
clc;clear;
close all

Area=1e4*1e-8 %1e-8; %(in cm^2)

cm = 1*Area; % membrane capacitance
dt = 0.01; % time step
T=100;
I0= 10*Area;
Iext = [zeros(1,ceil(T/dt)),I0*ones(1,3*ceil(T/dt)),zeros(1,ceil(T/dt))]; % ex

ENa = 50; %55.17;
EK = -77;% 72.14;
El =-55;% 49.42;

gNamax = 120*Area; %max conductances
gKmax = 36*Area;
glmax = 0.3*Area;

niter = ceil(T/dt); % max number of iterations

% Intial conditions
vm(1) = -60; % intial membrane voltage
V(1) = vm;

% calculating alphas abd betas for the intial membrane voltage
alpn = 0.01*(vm+55)/(1-exp(-(vm+55)/10));
alpm = 0.1*(vm+40)/(1-exp(-(vm+40)/10));
alph = 0.07*exp(-0.05*(vm+65));

betn = 0.125*exp(-(vm+65)/80);
betm = 4*exp(-0.0556*(vm+65));
beth = 1/(1+exp(-0.1*(vm+35)));

% intial m,n & h
n = alpn/(alpn+betn);
```

```

m = alpm/(alpm+betm);
h = alph/(alph+beth);
M=m;N=n;H=h;

% the loop for calculation of membrane voltage dynamics is here.
for i = 1:length(Iext)-1,

    dm = dt*(alpm*(1-m)-betm*m);
    dn = dt*(alpn*(1-n)-betn*n);
    dh = dt*(alph*(1-h)-beth*h);

    m = m + dm;
    n = n + dn;
    h = h + dh;

    gNa = gNamax*m^3*h;
    gK = gKmax*n^4;
    gl = glmax;

    INa = gNa*(vm-ENa);
    IK = gK*(vm-EK);
    Il = gl*(vm-El);

    dvm = (dt/cm)*(Iext(i)-( INa + IK + Il));

    vm = vm + dvm;
    V(i+1) = vm;
    H(i+1) = h;
    M(i+1) = m;
    N(i+1) = n;

    alpn = 0.01*(vm+55)/(1-exp(-(vm+55)/10));
    alpm = 0.1*(vm+40)/(1-exp(-(vm+40)/10));
    alph = 0.07*exp(-0.05*(vm+65));

    betn = 0.125*exp(-(vm+65)/80);
    betm = 4*exp(-0.0556*(vm+65));
    beth = 1/(1+exp(-0.1*(vm+35)));

    INa(i+1)=gNa*m^3*h*(vm-ENa);
    IK(i+1)=gK*n^4*(vm-EK);
    IL(i+1)=gl*(vm-El);

    Pe(i+1)=Iext(i+1)*V(i+1);

```

```

PNa(i+1)=INa(i+1)*(V(i+1)-ENa);
PK(i+1)=IK(i+1)*(V(i+1)-EK);
PL(i+1)=IL(i+1)*(V(i+1)-EL);
Pc(i+1)=V(i+1)*(Iext(i+1)-( INa(i+1) + IK(i+1) + IL(i+1)));

end
tim = dt*(1:length(Iext)+1);
h=figure('Position',[100 100 820 700])
subplot(2,1,1)

set(gca,'fontsize',26)
plot(tim(1:end-1),(V)/1,'b','linewidth',2);
ylabel('V (mV)');
axis([0,5*T,-100,100])

subplot(2,1,2)
set(gca,'fontsize',26)
plot(tim(1:end-1),1e6*Iext,'r','linewidth',2);
axis([0,5*T,min(1e6*Iext),max(1e6*Iext)])
ylabel('I (pA)');
xlabel('Time (ms)');

h=figure('Position',[100 100 820 700])
subplot(3,1,1)
set(gca,'fontsize',26)
plot(tim(1:end-1),N,'r','linewidth',2);
ylabel('n')

subplot(3,1,2)
set(gca,'fontsize',26)
plot(tim(1:end-1),M,'b','linewidth',2);
ylabel('m')

subplot(3,1,3)
set(gca,'fontsize',26)
plot(tim(1:end-1),H,'g','linewidth',2);
ylabel('h')
xlabel('Time (ms)');

[pks,locs]=findpeaks(V);

```

```

ti=(locs(pks>0));
t1=(ti(2)-100) ;
t2=(ti(3)-100);

%In pJ
ENa=trapz(tim(t1:t2)-tim(t1), PNa(t1:t2))
EK=trapz(tim(t1:t2)-tim(t1), PK(t1:t2))
El=trapz(tim(t1:t2)-tim(t1), PL(t1:t2))

Ec=trapz(tim(t1:t2)-tim(t1), Pc(t1:t2))
Ettotal=(ENa+EK+El+Ec)

h=figure('Position',[100 100 820 700]);
set(gca,'fontsize',26)
plot(tim(t1:t2), PNa(t1:t2),'r','linewidth',2);
hold on
plot(tim(t1:t2), PK(t1:t2),'g','linewidth',2);
hold on
plot(tim(t1:t2), PL(t1:t2),'b','linewidth',2);
hold on
plot(tim(t1:t2), Pc(t1:t2),'k','linewidth',2);
grid

ylabel('Power(nW)');
xlabel('Time (ms)');
legend('Na','K','L','C')
axis([tim(t1),tim(t2),-.50,2 ])

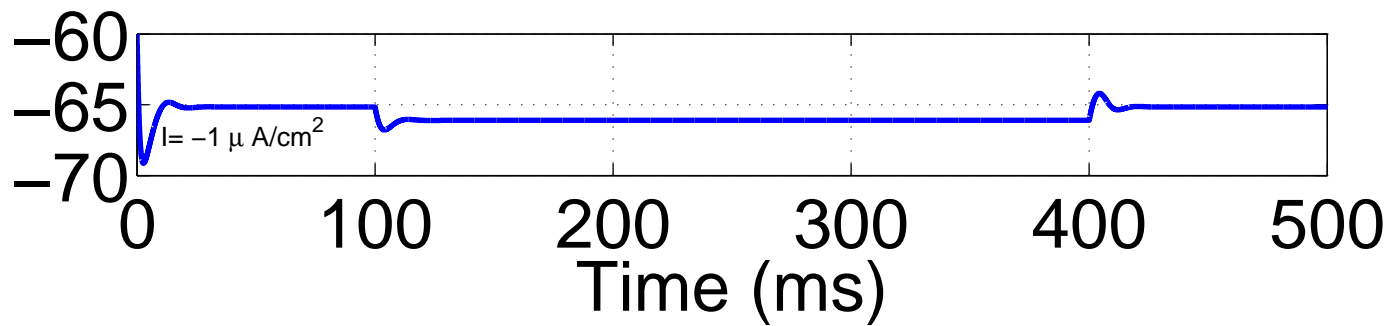
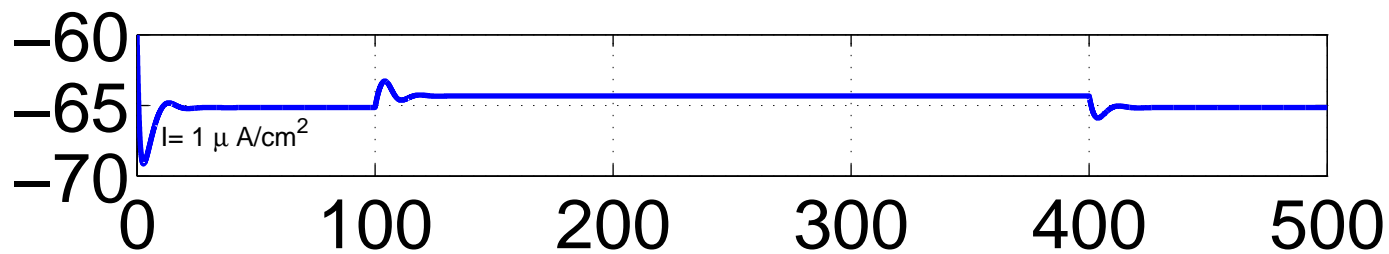
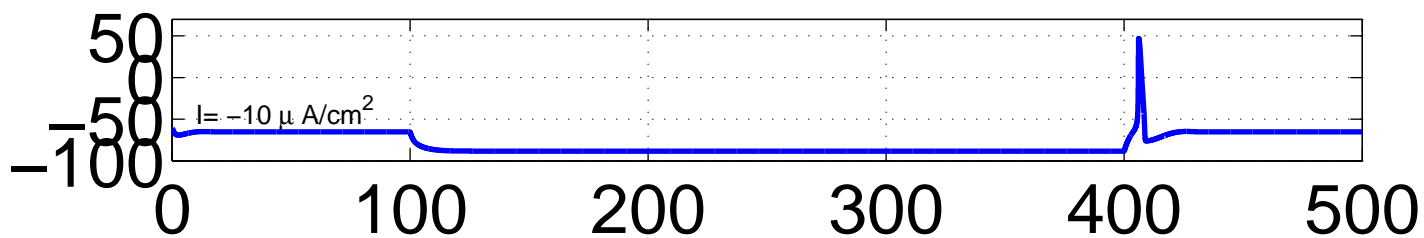
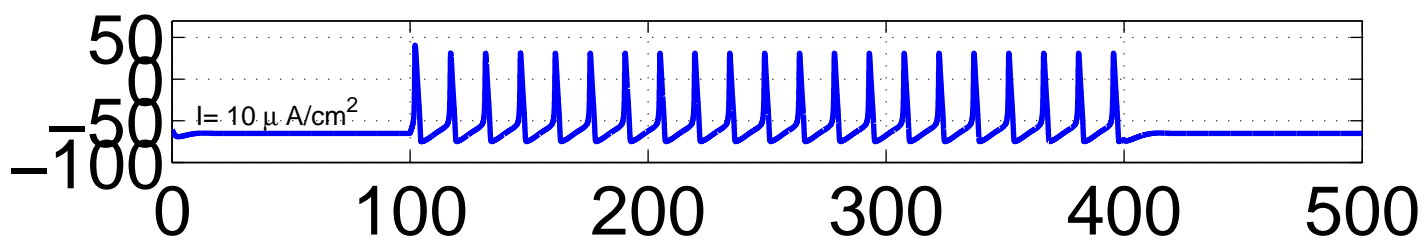
h=figure('Position',[100 100 820 700]);
subplot(3,1,1)
set(gca,'fontsize',26)
plot(tim(t1:t2),N(t1:t2),'r','linewidth',2);
ylabel('n')
axis([tim(t1),tim(t2),0,1])
grid

subplot(3,1,2)
set(gca,'fontsize',26)
plot(tim(t1:t2),M(t1:t2),'b','linewidth',2);
ylabel('m')

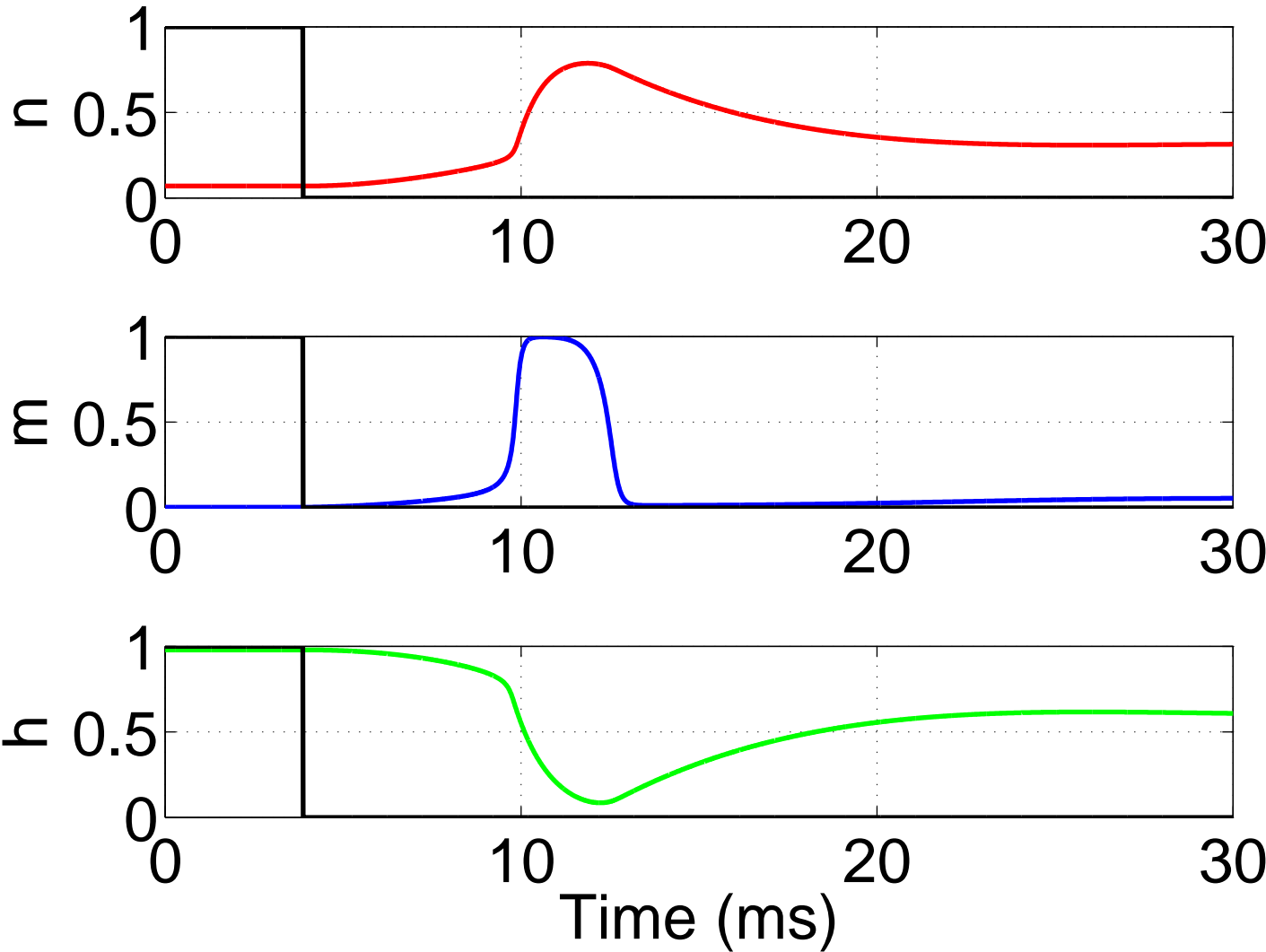
```

```
axis([tim(t1),tim(t2),0,1])  
grid
```

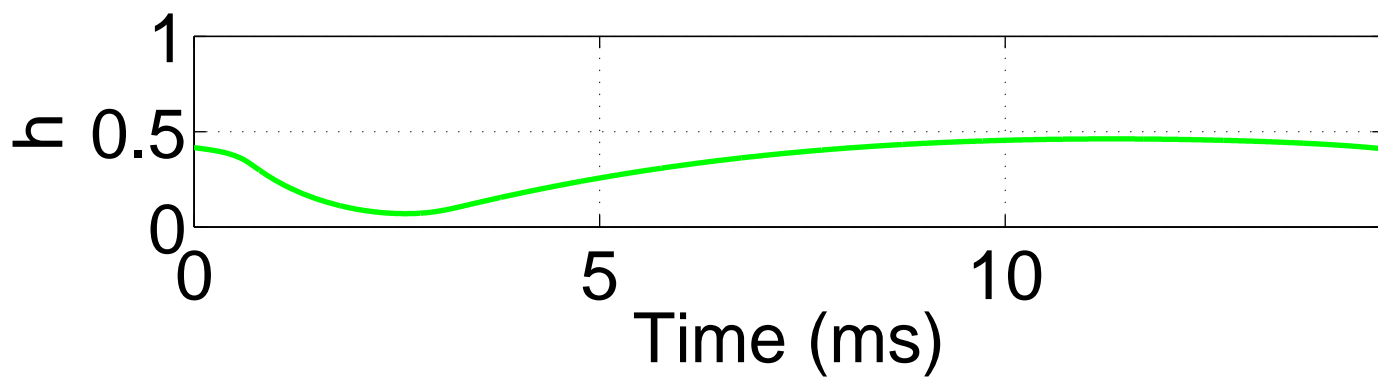
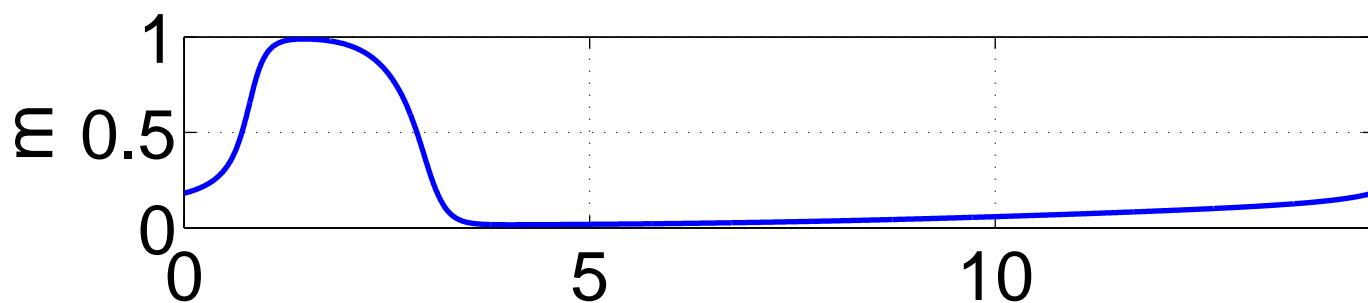
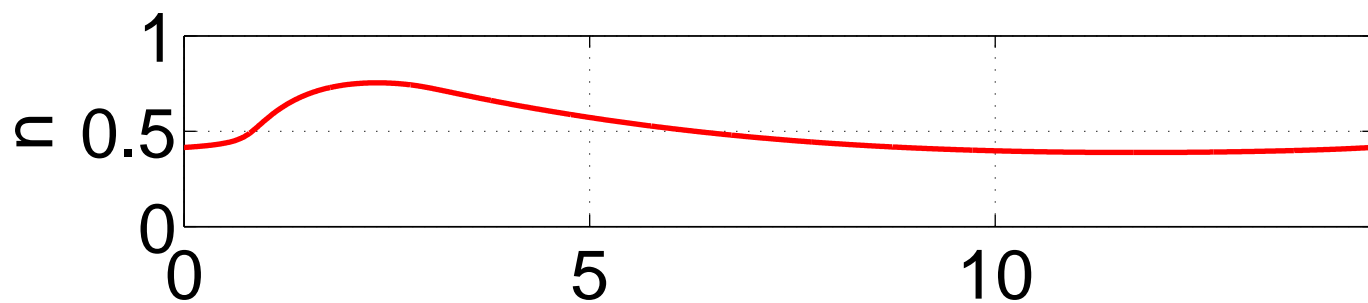
```
subplot(3,1,3)  
set(gca,'fontsize',26)  
plot(tim(t1:t2),H(t1:t2),'g','linewidth',2);  
ylabel('h')  
xlabel('Time (ms)');  
grid  
axis([tim(t1),tim(t2),0,1])
```



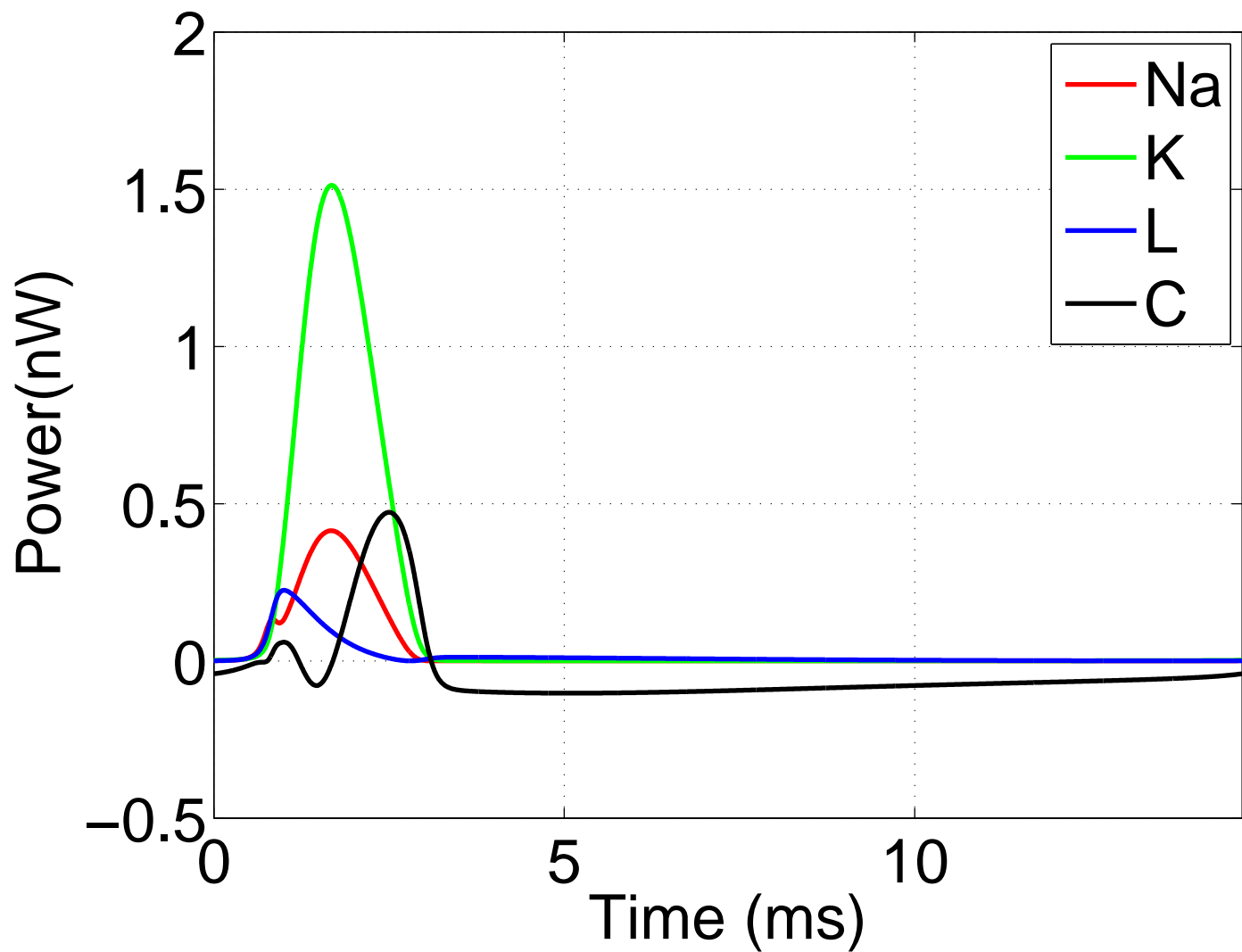
Activation variables during Post-inhibitory rebound



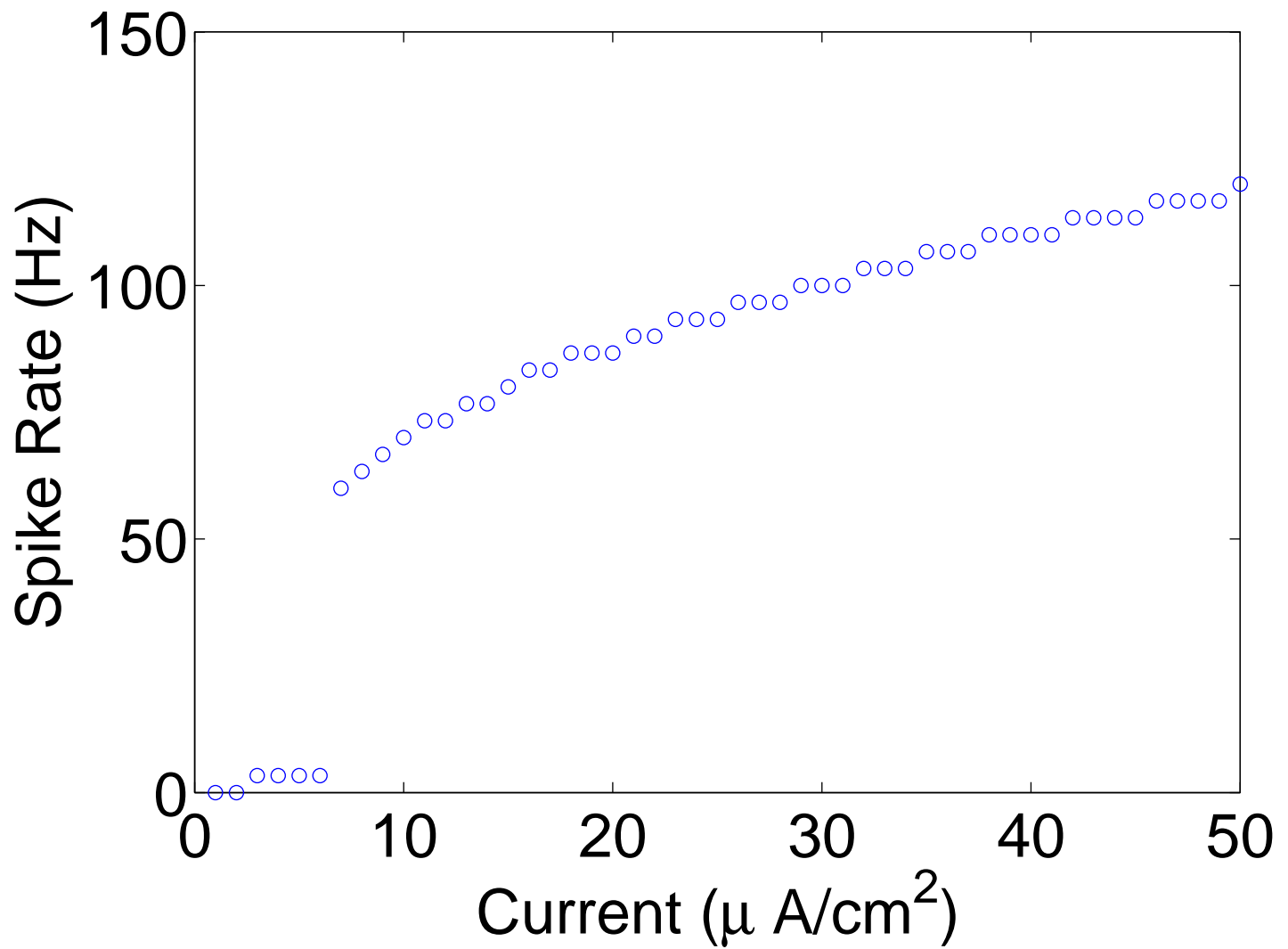
Activation variables during Action Potential



Instantaneous Power during Action Potential



Spike rate



Mechanism for Post-inhibitory rebound: See <http://genesis-sim.org/node/816>

Power calculation: After numerical integration, the total energy dissipated in the channel during one cycle of the action potential is calculated to be about $0.20814e$ fJ for a patch of area $1\mu\text{m}^2$. This translates to about 2pJ for an average neuron with diameter of $100\mu\text{m}$.