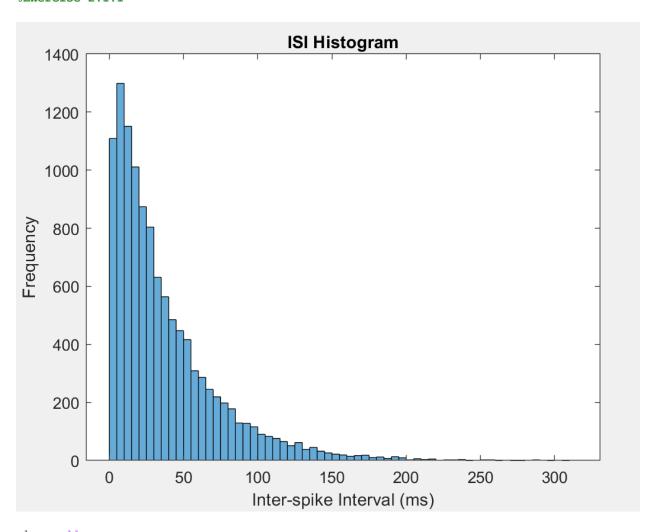
Jason Wong $1090292616\ 15/11/2017$ %Note, if you would like to run the plots, please comment out the close(x) lines <code>%Exercise 2.1.1</code>



```
clear all
close all
load Spike_data_1.mat
spikeData1=d;
load Spike data 2.mat
spikeData2=d;
load Spike_data_3.mat
spikeData3=d;
ISI = [];
n trials = 1000;
T = 500;
for k=1:n trials
    spike_times = find(spikeData1(k,:) == 1);
    isi0 = diff(spike_times);
    ISI = [ISI, isi0];
end
figure(1)
histogram(ISI);
title('ISI Histogram')
```

```
ylabel('Frequency')
xlabel('Inter-spike Interval (ms)')
close(1)
```

%Exercise 2.1.2

%The neuron fires earlier rather than later.

*Because the curve fits inside a negative exponential curve, therefore the *neuron is Poissonnian and follows Poisson distribution. This is because *neurons fire randomly in a fixed average rate.

%Exercise 2.1.3

%meanTime=36.9525ms, that means it takes on average 36.9525 milliseconds for %the neuron to fire again

meanTime = mean(ISI);

%Exercise 2.1.4

%modeTime=2ms, that means the most common time it took for the neuron to %fire again was 2ms modeTime = mode(ISI);

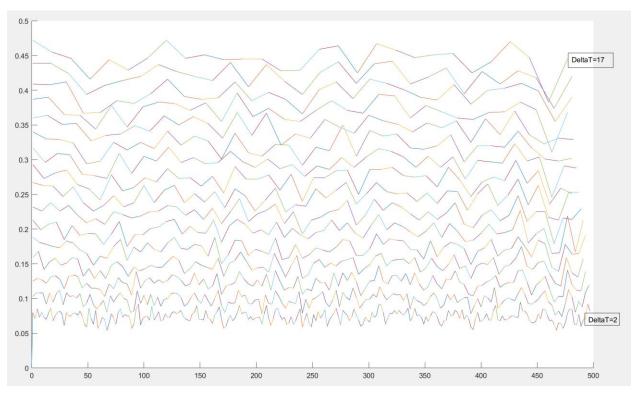
%Exercise 2.1.5

%meanTime(36.9525ms) is not equal to modeTime(2ms), because the average is %not the same as the mode

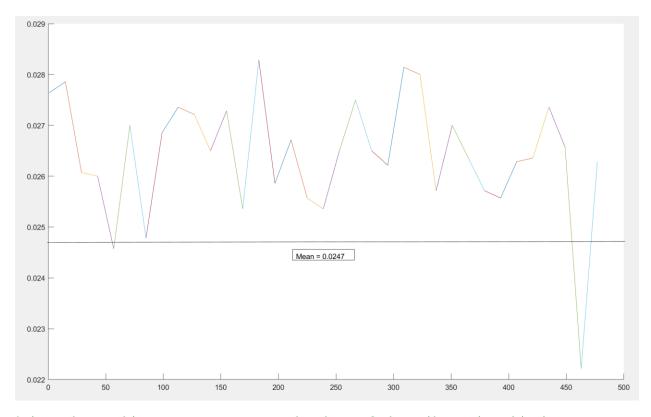
%Exercise 2.1.6

%I binned each set of spikes into bins seperated by deltaT, and by changing %deltaT from 2 to 20ms, I could remove noise until a clear spike train %emerged.

%Plotting deltaT from 2ms to 20ms

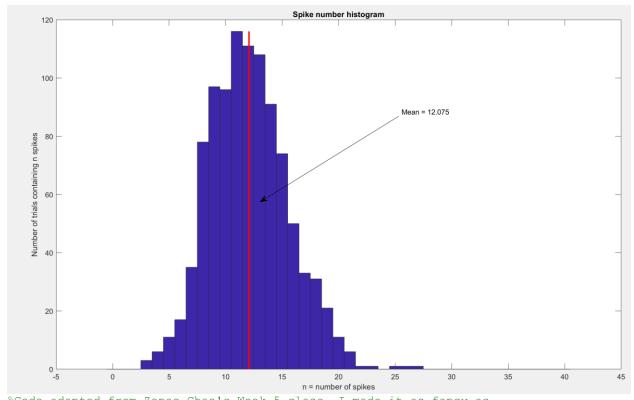


%Plotting deltaT=14 as it's the most clearest, note the ~12 spikes



```
%Given the ISI histogram, I can guess the shape of the spike train, this is
%because I can guess the average ISI using the histogram and then I will
%know the average intervals between each spike in the spike train.
figure(2)
prob spike = sum(spikeData1,1)/n trials;
prevDot = 0;
prevT = 0;
hold on
for deltaT=2:1:17
    for t=1:deltaT:500-deltaT
        sumSpikes=0;
        for i=t:1:t+deltaT
            sumSpikes=sumSpikes+prob_spike(i);
        end
        if t>prevT
            plot([prevT t], [prevDot sumSpikes]);
        prevT = t;
        prevDot = sumSpikes;
    end
end
hold off
close(2);
figure(11)
prob_spike = sum(spikeData1,1)/n_trials;
for deltaT=14:1:14 %deltaT = 14 has the clearest spike train with n cloests to 12.344
    for t=1:deltaT:500-deltaT
        sumSpikes=0;
        for i=t:1:t+deltaT
            sumSpikes=sumSpikes+prob_spike(i);
        sumSpikes=sumSpikes/deltaT
```

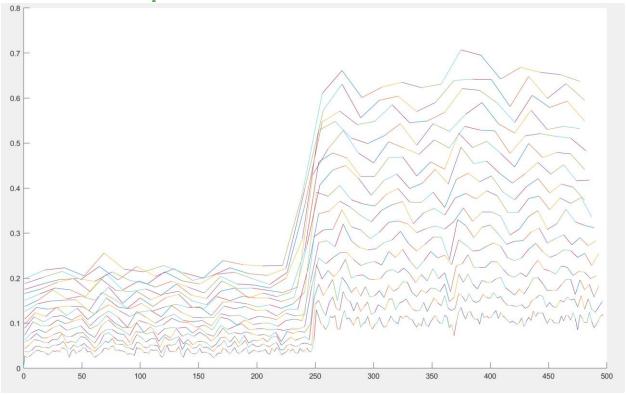
```
if t>prevT
            plot([prevT t], [prevDot sumSpikes]);
        prevT = t;
       prevDot = sumSpikes;
    end
end
hold off
close(11);
%Exercise 2.2.1
%This sums up every spike, and divides it by n trials for the average spike
%per trial
%avgSpikePerTrial = 12.0750
avgSpikePerTrial = sum(sum(spikeData2))/n trials;
%Exercise 2.2.2
%prob_Spike is an array of all the probability for a neuron to spike
%mean(prob_spike) gives the average firing rate probability
%avgProb Spike = 0.0241
prob_spike = sum(spikeData2)/n_trials;
avgProb spike = mean(prob spike);
%Exercise 2.2.3
%Average ISI is found by using the mean function on the ISI.
%meanTime = 35.6245
ISI2 = [];
n trials = 1000;
T = 500;
for k=1:n trials
    spike_times = find(spikeData2(k,:) == 1);
    isi0 = diff(spike times);
    ISI2 = [ISI2, isi0];
meanTime = mean(ISI2);
%Exercise 2.2.4
```

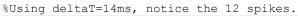


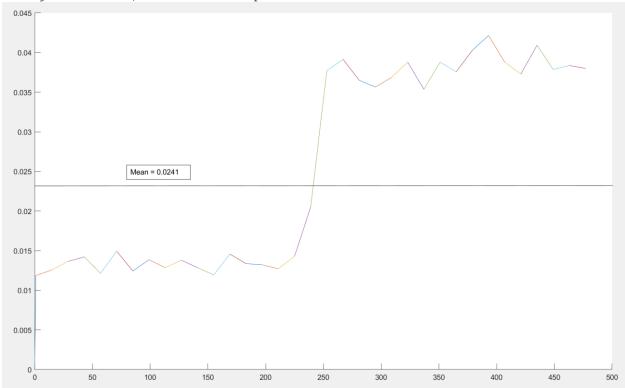
```
%Code adapted from Zenas Chao's Week 5 class. I made it as fancy as
%possible
%n avg = 12.075 spikes
n_spikes_per_trial = sum(spikeData2,2);
n avg = mean(n spikes per trial);
figure(3)
hist(n spikes per trial, [0:1:40])
xlabel('n = number of spikes');
ylabel('Number of trials containing n spikes');
title('Spike number histogram')
hold on
plot([n_avg, n_avg], [0, max(hist(n_spikes_per_trial, [0:1:40]))], 'r', 'LineWidth',
2)
x = [0.6 \ 0.43];
y = [0.7 \ 0.5];
annotation('textarrow',x,y,'String','Mean = 12.075') %Labels the mean
hold off
close(3)
%Exercise 2.2.5
figure(4)
hold on
histogram(ISI2);
title('ISI Histogram')
xlabel('Inter-spike Interval (ms)');
ylabel('Frequency');
line([meanTime, meanTime], ylim, 'LineWidth', 1, 'Color', 'k'); %Plots the mean of the
histogram
x = [0.4 \ 0.25];
y = [0.7 \ 0.5];
annotation('textarrow',x,y,'String','Mean = 35.6245') %Labels the mean
hold off
```

close(4)

%Exercise 2.2.6
%Look at 2.1.6 for explanation of this code







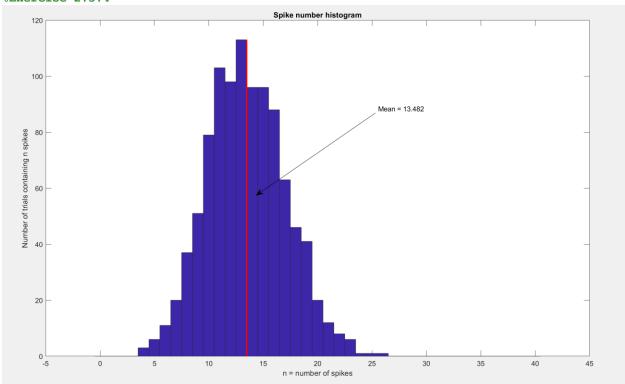
figure(5)
prob_spike = sum(spikeData2,1)/n_trials;

```
prevDot = 0;
prevT = 0;
hold on
for deltaT=2:1:17
    for t=1:deltaT:500-deltaT
        sumSpikes=0;
        for i=t:1:t+deltaT
            sumSpikes=sumSpikes+prob spike(i);
        if t>prevT
            plot([prevT t], [prevDot sumSpikes]);
        end
        prevT = t;
        prevDot = sumSpikes;
    end
end
hold off
close(5);
figure(10)
prob spike = sum(spikeData2,1)/n trials;
prevDot = 0;
prevT = 0;
hold on
for deltaT=14:1:14 %DeltaT = 14ms for the most clearest spike train
    for t=1:deltaT:500-deltaT
        sumSpikes=0;
        for i=t:1:t+deltaT
            sumSpikes=sumSpikes+prob spike(i);
        sumSpikes=sumSpikes/deltaT
        if t>prevT
            plot([prevT t], [prevDot sumSpikes]);
        prevT = t;
        prevDot = sumSpikes;
    end
end
hold off
close(10);
%Exercise 2.2.7
%Average ISI: Data1 = 36.9525ms, Data2 = 35.6245ms (Exercise 2.1.3 and
%Exercise 2.2.3 results)
%Data1 > Data 2
%Most common ISI: Data1 = 2ms 0.0251 occurance rate, Data2 = 1ms 0.0315
%occurance
%Occurance rate found using numel(find(ISI == 2))/numel(ISI)
%Data 1 mode > Data 2 mode. Data 2 occurance rate > Data 1 occurance
%Average probability of spiking: Data1 = 0.0247, Data2 = 0.0241
%This is very similar so it can't be used to reliably differentiate between Data 1 and
%Looking at the spike train of Data 2, the spike probability almost tripled
%halfway through the trial. This indicates that the subject
%underwent a stimulus event at the halfway point.
%For the spike train of Data 1, there wasn't such stimulus event, and the
%spike train had a single average.
Because Data 1 doesn't have a jump in spike probability which is present
%in Data 2, my conclusion is that they display different activity.
%Exercise 2.3.8
```

%Activity in data2 has a massive jump halfway into the trial, most likely %because of stimulation from an external source. Other than that there %weren't a lot of unique identifiers.

```
%Exercise 2.3.1
%This sums up every spike, and divides it by n trials for the average spike
%per trial
%avgSpikePerTrial = 13.4820
avgSpikePerTrial = sum(sum(spikeData3))/n_trials;
%Exercise 2.3.2
prob\_Spike is an array of all the probability for a neuron to spike
%mean(prob_spike) gives the average firing rate probability
%avgProb Spike = 0.0270
prob spike = sum(spikeData3)/n trials;
avgProb spike = mean(prob spike);
%Exercise 2.3.3
%Average ISI is found by using the mean function on the ISI.
meanTime = 32.4087
ISI3 = [];
n trials = 1000;
T = 500;
for k=1:n trials
    spike times = find(spikeData3(k,:) == 1);
    isi0 = diff(spike times);
    ISI3 = [ISI3, isi0];
end
meanTime = mean(ISI3);
```

%Exercise 2.3.4

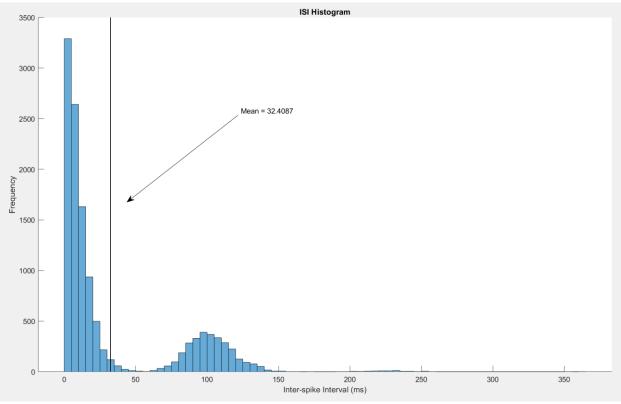


Code adapted from Zenas Chao's Week 5 class. I made it as fancy as possible

n spikes per trial = sum(spikeData3,2);

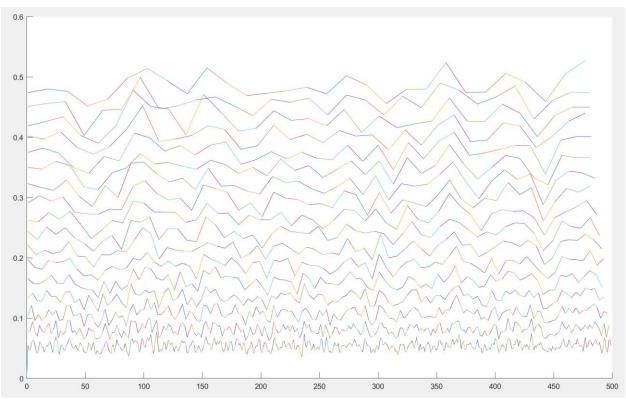
```
n_avg = mean(n_spikes_per_trial);
figure(6)
hist(n_spikes_per_trial, [0:1:40])
xlabel('n = number of spikes');
ylabel('Number of trials containing n spikes');
title('Spike number histogram')
hold on
plot([n_avg, n_avg], [0, max(hist(n_spikes_per_trial, [0:1:40]))], 'r', 'LineWidth', 2)
x = [0.6 0.43];
y = [0.7 0.5];
annotation('textarrow',x,y,'String','Mean = 13.482') %Labels the mean
hold off
close(6)
```

%Exercise 2.3.5



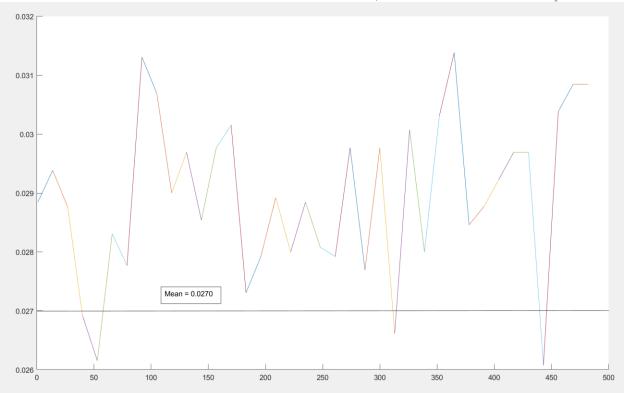
```
%Mean ISI = 32.4087
%There are two peaks, which allows us to separate data 3 from 1 and 2
figure(7)
hold on
histogram(ISI3);
title('ISI Histogram')
xlabel('Inter-spike Interval (ms)');
ylabel('Frequency');
line([meanTime, meanTime], ylim, 'LineWidth', 1, 'Color','k'); %Plots the mean of the
histogram
x = [0.4 0.25];
y = [0.7 0.5];
annotation('textarrow',x,y,'String','Mean = 32.4087') %Labels the mean
hold off
close(7)
```

%Exercise 2.3.6



\$I binned each set of spikes into bins seperated by deltaT, and by changing \$deltaT from 2 to 20ms, I could remove noise until a clear spike train \$emerged.

%The deltaT is 13ms for the clearest line for data 3, notice the 13 visible spikes



figure(8)

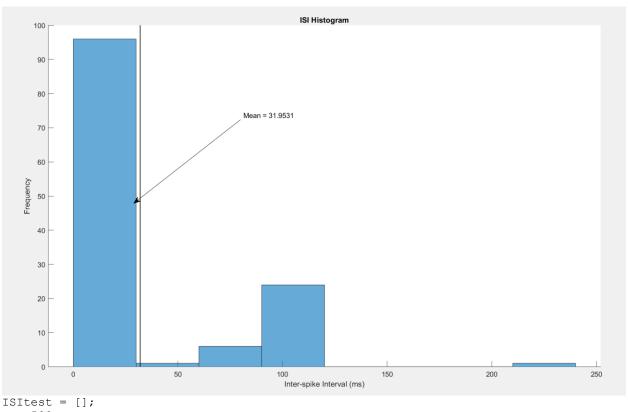
```
prob spike = sum(spikeData3,1)/n trials;
prevDot = 0;
prevT = 0;
hold on
for deltaT=1:1:17 %Plotting for deltaT = 1 to 17
    for t=1:deltaT:500-deltaT
        sumSpikes=0;
        for i=t:1:t+deltaT
            sumSpikes=sumSpikes+prob spike(i);
        if t>prevT
            plot([prevT t], [prevDot sumSpikes]);
        prevT = t;
        prevDot = sumSpikes;
    end
end
hold off
close(8)
figure (9)
hold on
for deltaT=[13] %Plotting for deltaT = 13 has the most clear spike train
    for t=1:deltaT:500-deltaT
        sumSpikes=0;
        for i=t:1:t+deltaT
            sumSpikes=sumSpikes+prob spike(i);
        sumSpikes=sumSpikes/deltaT
        if t>prevT
            plot([prevT t], [prevDot sumSpikes]);
        prevT = t;
        prevDot = sumSpikes;
    end
end
hold off
close(9);
%Exercise 2.3.7
%The ISI histogram for data 3 is very different to those of 1 and 2, as it
%has another peak at about 100ms whereas 1 and 2 doesn't have that peak.
%Looking at the spike train of Data 3, it is very similar to Data 1, but
%the important difference is that the amplitude of the spikes in Data 3 are
%much greater.
Average ISI: Data1 = 36.9525ms, Data2 = 35.6245ms Data3 = 32.4087ms
%Data1 > Data2 > Data3 in speed
%Most common ISI: Data1 = 2ms 0.0251 occurance rate, Data2 = 1ms 0.0315
%occurance Data3 = 1ms 0.0731 occurance rate
%Occurance rate found using numel(find(ISI == 2))/numel(ISI)
%Data 1 mode > Data 2 mode
%Average probability of spiking: Data1 = 0.0247, Data2 = 0.0241 Data3 =
%0.0270
%Data3 has a higher one than data1 and 2, so this can be used to
%differentiate it out of the other two
%Exercise 2.3.8
%Data 3 has a similar spike train to Data 1, the identifying feature for
%this data set is that for the ISI histogram, there is a second peak at
%100ms. The second peak may be because the neuron only fires at regular intervals.
```

%Exercise 2.4.1

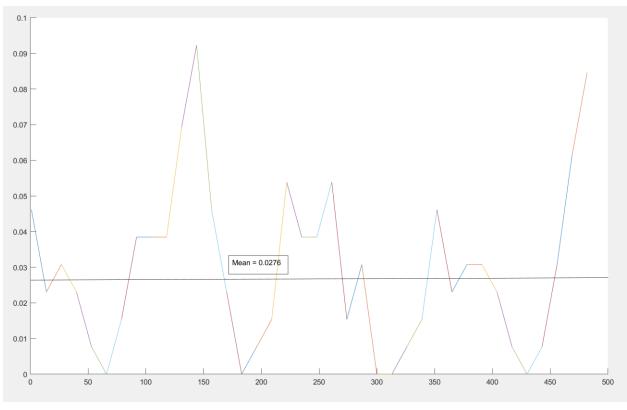
```
%I will look at the ISI histogram and the spike train data, if the spike %probability suddenly increases halfway through the trials, then it must be %C2. If the ISI histogram has two peaks, then it must be C3. If it doesn't %have two then it must be C1 load('Spike_data_test.mat') spikeDataTest = d_test; n_trials = 10;
```

%The ISI Histogram shows two peaks, so it must be C3

is very similar to C3 (0.0270) so it must be C3.



```
T = 500;
for k=1:n_trials
    spike_times = find(spikeDataTest(k,:) == 1);
    isi0 = diff(spike times);
    ISItest = [ISItest,isi0];
end
figure(13)
hold on
histogram (ISItest);
title('ISI Histogram')
xlabel('Inter-spike Interval (ms)');
ylabel('Frequency');
line([meanTime, meanTime], ylim, 'LineWidth', 1, 'Color', 'k'); %Plots the mean of the
histogram
x = [0.4 \ 0.25];
y = [0.7 \ 0.5];
annotation('textarrow',x,y,'String','Mean = 32.4087') %Labels the mean
hold off
%close(13)
%Checking the spike train data, there doesn't seem to be an increase of
%spike probability halfway through the trial so it definitely isn't C2, also the mean
```



```
figure(15)
hold on
prob_spike = sum(spikeDataTest,1)/10;
for deltaT=[13] %Plotting for deltaT = 13ms has the most clear spike train
    for t=1:deltaT:500-deltaT
        sumSpikes=0;
        for i=t:1:t+deltaT
            sumSpikes=sumSpikes+prob_spike(i);
        end
        sumSpikes=sumSpikes/deltaT;
        if t>prevT
            plot([prevT t], [prevDot sumSpikes]);
        end
        prevT = t;
        prevDot = sumSpikes;
    end
end
hold off
%close(15);
```