

synaptic_LIF_poisson

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0.1 Stimulating a neuron with synaptic current

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[ ]: # Author #  
#####  
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# This code includes solution for Q2 and Q3 of Assignment.  
# References used:  
# https://compneuro.neuromatch.io/tutorials/W2D3_BiologicalNeuronModels/student/  
  ↪ W2D3_Tutorial1.html  
  
[ ]: # Importing dependencies  
import numpy as np  
import matplotlib.pyplot as plt  
%matplotlib inline  
plt.style.use('ggplot')
```

The dynamics of LIF neuron model under synaptic input can be explained by

$$C_m \frac{dv}{dt} = -g_L(V - E_L) + I_{\text{syn}(t)} + I_{\text{ext}} \quad (1)$$

where C_m is the membrane capacitance, g_L is the leak potential, I_{ext} is an external current injected in the neuron, $I_{\text{syn}(t)}$ is the synaptic current.

Assuming neuron has LIF model, we can estimate C_m as:

$$C_m = \tau_m / R_m \quad (2)$$

```
[ ]: # Define the model parameters  
def default_params(**kwargs):  
    params = {}  
  
    # Define neuron parameters  
    params['tau_m'] = 50.0 # Membrane time constant  
    params['g_L'] = 1e-6 # Leak resistance (in ohms)  
    params['E_L'] = 0.0 # Resting potential (in mV)  
    params['V_th'] = 20.0 # Threshold voltage (in mV)
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params['V_reset'] = 0.0 # Reset voltage (in mV)
params['V_init'] = 0.0 # initial potential [mV]
params['t_ref'] = 2.0 # Refractory period (in ms)

# Define the simulation parameters
params['dt'] = 0.1 # Time step (in ms)
params['T'] = 1000.0 # Simulation duration (in ms)

params['range_t'] = np.arange(0, params['T'], params['dt']) # Vector of
↳ discretized time points [ms]

# external parameters if any
for k in kwargs:
    params[k] = kwargs[k]

return params

```

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[ ]: def generate_poisson_spikes(params, rate, user_seed=False):
    """
    Generate poisson spike trains

    Args:
    params      : Parameters dictionary
    rate        : Poisson noise amplitude (Hz)
    n           : Number of Poisson spike trains
    user_seed   : Random seed. int or boolean
    """
    # get simulation parameters
    dt, range_t = params['dt'], params['range_t']
    Lt = range_t.size

    # set random seed
    if user_seed:
        np.random.seed(seed=user_seed)
    else:
        np.random.seed()

    # generate poisson spike trains
    I_pos = np.random.poisson(lam=rate * (dt/1000.), size=Lt)

    return I_pos

```

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[ ]: # Synaptic current waveform
def synaptic_current(I0, tau_syn, t):
    return I0 * np.sum(np.exp(-t/tau_syn))

```

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[ ]: def run_synaptic_LIF(params, I_ext):
    """
    Calculate coefficient of variation of interspike intervals as a function of
    stimulus rate

    Args:
    params      : Parameters dictionary
    rate_range  : Range of stimulus rates (Hz)
    num_trials  : Number of trials for each rate

    Returns:
    cv_list     : List of coefficients of variation for each rate
    """
    # Set parameters
    V_th, V_reset = params['V_th'], params['V_reset']
    tau_m, g_L = params['tau_m'], params['g_L']
    I0 = params['I0']
    tau_syn = params['tau_syn']
    V_init, E_L = params['V_init'], params['E_L']
    dt, range_t = params['dt'], params['range_t']
    t_ref = params['t_ref']
    Lt = range_t.size

    # Initialize voltage and current
    v = np.zeros(Lt)
    v[0] = V_init
    I_ext = I_ext * np.ones(Lt)
    I_syn = np.zeros_like(I_ext)
    tr = 0. # the count for refractory duration

    rec_spikes = [] # record spike times
    # Simulate LIF dynamics
    for it in range(Lt - 1):

        # Compute synaptic current
        if len(rec_spikes) > 0:
            t = (it*dt - rec_spikes)
            I_syn[it+1] = synaptic_current(I_syn[it], tau_syn, t)
        else:
            I_syn[it+1] = I_syn[it] + I0

        if tr > 0: # check if in refractory period
            v[it] = V_reset # set voltage to reset
            tr = tr - 1 # reduce running counter of refractory period

        elif v[it] >= V_th: # if voltage over threshold
            rec_spikes = np.append(rec_spikes, it*dt) # record spike event
```

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        v[it] = V_reset # reset voltage
        tr = t_ref / dt # set refractory time

        # calculate the increment of the membrane potential
        dv = (-(v[it] - E_L) + I_ext[it] / g_L + I_syn[it]) * (dt / tau_m)

        # Update the membrane potential
        v[it + 1] = v[it] + dv

        # Get spike times in ms
        rec_spikes = np.array(rec_spikes) * dt

    return rec_spikes

```

```

[ ]: def plot_firing_rate_vs_input(params, rate_range):
    """
    Plots the firing rate as a function of the input current.

    Args:
        params          : parameter dictionary
        rate_range      : firing rate range of the Poisson process

    Returns:
        None (plots the results)
    """

    spk_counts = np.zeros(len(rate_range))

    for idx in range(len(rate_range)):
        I_pos = generate_poisson_spikes(params, rate=rate_range[idx],
        ↪user_seed=2020)
        rec_spikes = run_synaptic_LIF(params, I_ext=I_pos)
        spk_counts[idx] = len(rec_spikes)

    # Plot the F-I curve i.e. Output firing rate as a function of input mean.
    plt.figure()
    plt.plot(rate_range, spk_counts, 'bo-', alpha=0.8, lw=2, dashes=(2, 2),
              label='Poisson input')
    plt.ylabel('Spike count')
    plt.xlabel('Average injected current (nA)')
    plt.legend(loc='best')
    plt.title('Firing Rate vs. Input Current')
    plt.savefig('Synaptic_Firing_rate(Hz)_vs_External_input_current(mA)_plot.
    ↪pdf')
    plt.show()

    return spk_counts

```

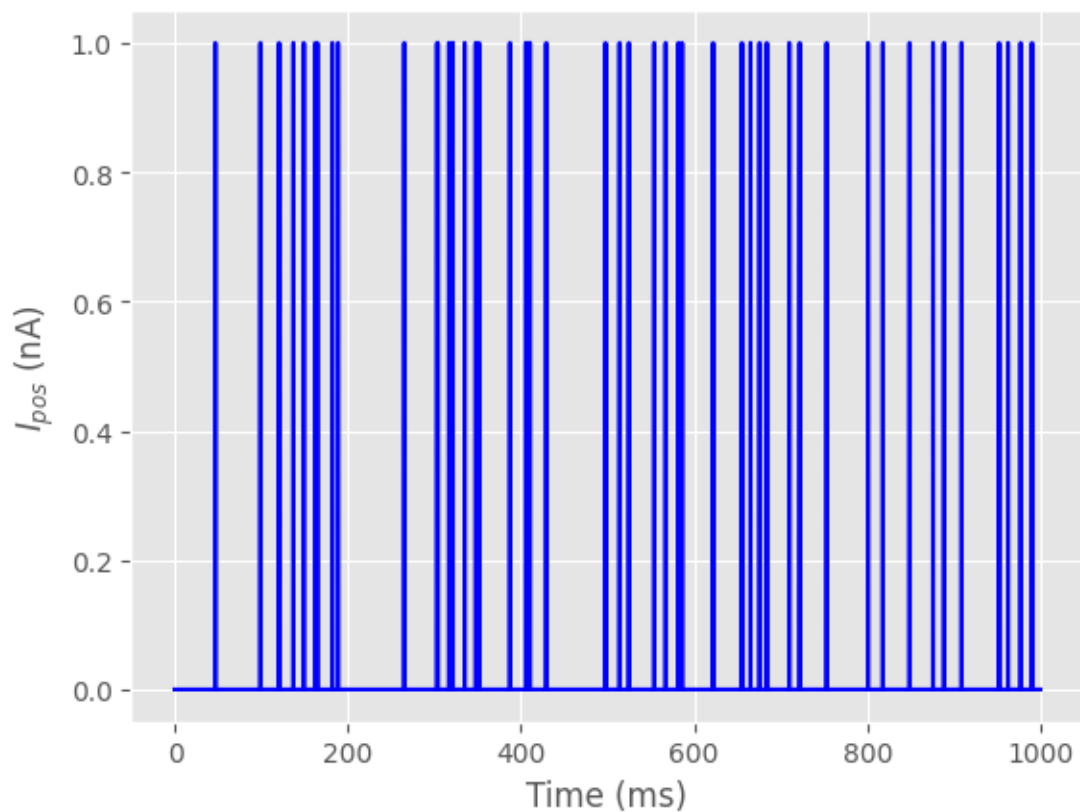
```
[ ]: # Default parameters
params = default_params(T=1000.)

# Add parameters
params['tau_syn'] = 5. # [ms]
params['I0'] = 20.

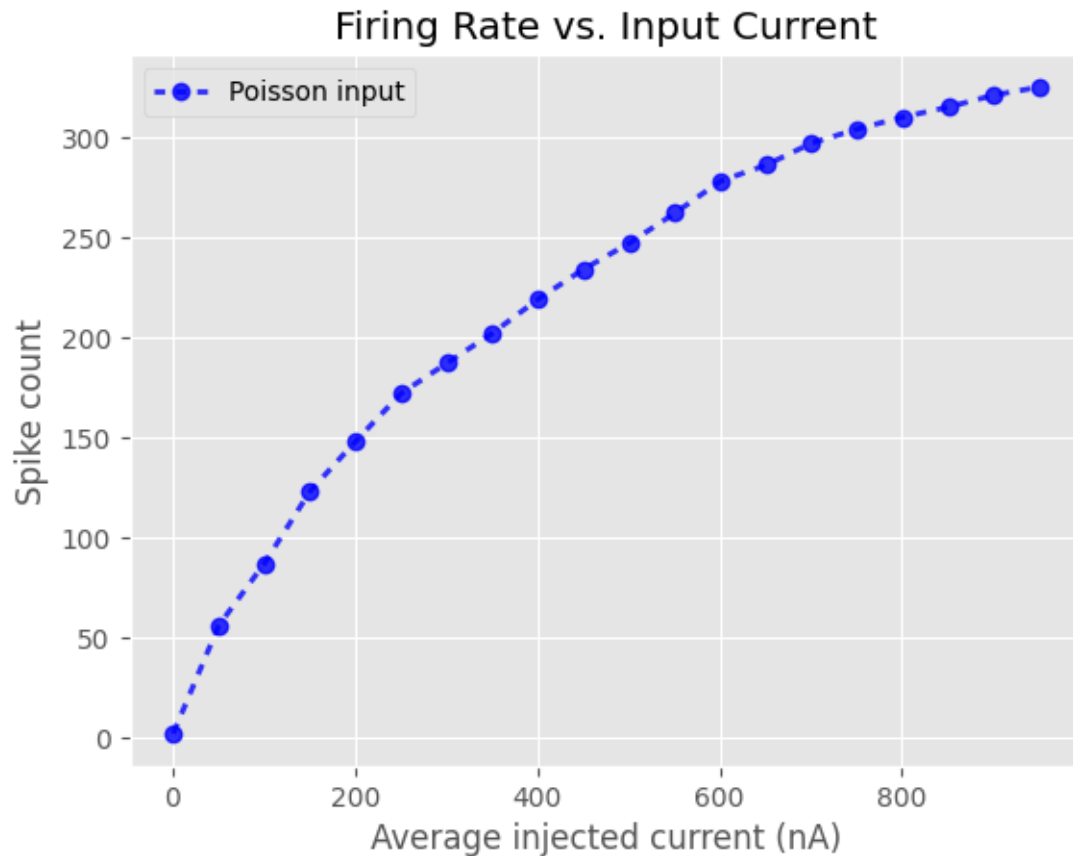
# Generate poisson spike trains
rate_range = np.arange(1, 1000, 50) # poisson rate range (Hz)

[ ]: # Plot example input spikes
I_pos = generate_poisson_spikes(params, rate=rate_range[1], user_seed=False)
plt.plot(params['range_t'], I_pos, 'b-')
plt.xlabel('Time (ms)')
plt.ylabel(r'$I_{pos}$ (nA)')

[ ]: Text(0, 0.5, '$I_{pos}$ (nA)')
```



```
[ ]: # Plot firing rate vs input current
spk_counts = plot_firing_rate_vs_input(params, rate_range)
```



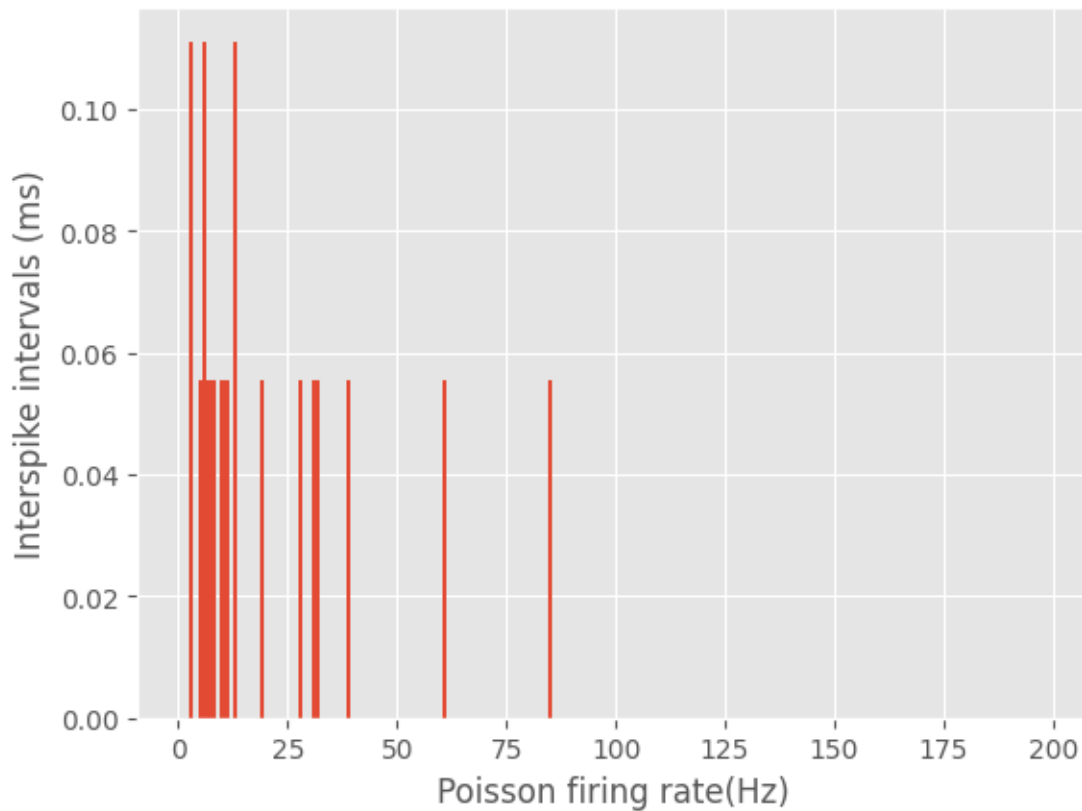
Note: F-I curve smoothens out if we increase firing rate.

0.2 Interspike interval distribution

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[ ]: spike_intervals = np.diff(spik_counts) # calculating interspike intervals by
      ↪ subtracting consecutive spikes
duration = params['T']
dt = params['dt']
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[ ]: # Compute histogram of the spike intervals
fig, ax = plt.subplots()
bins = 100
binSize = 1 # 1ms bins
x = np.arange(1, 200, binSize)
intervals_dist, bins = np.histogram(spike_intervals[spike_intervals<200],
      ↪ bins=x)
intervals_dist = intervals_dist / (np.sum(intervals_dist) * binSize) # normalize
ax.bar(bins[:-1], intervals_dist, width=binSize)
ax.set_xlabel('Poisson firing rate(Hz)')
```

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ax.set_ylabel('Interspike intervals (ms)')
plt.savefig('ISI_distribution_plot.pdf')
```



The histogram of the interspike intervals resembles an exponential probability distribution.

```
[ ]: def isi_cv_LIF(spike_times):
    """
    Calculates the inter-spike intervals (isi) and
    the coefficient of variation (cv) for a given spike_train
    Args:
        spike_times : (n, ) vector with the spike times (ndarray)
    Returns:
        isi          : (n-1,) vector with the inter-spike intervals (ms)
        cv           : coefficient of variation of isi (float)
    """
    if len(spike_times) >= 2:
        # Compute isi
        isi = np.diff(spike_times)
        # Compute cv
        cv = np.std(isi)/np.mean(isi)
    else:
```

```

isi = np.nan
cv = np.nan

return isi, cv

```

```

[ ]: cv_list = []
# Calculating coefficient of variance in interspike intervals
for rate in rate_range:
    I_pos = generate_poisson_spikes(params, rate)
    rec_spikes = run_synaptic_LIF(params, I_pos)
    _, cv = isi_cv_LIF(rec_spikes)
    cv_list.append(cv)

```

```

/var/folders/x2/w3k0335j38d_xs4_cjy_0lzh0000gn/T/ipykernel_15547/724215854.py:15
: RuntimeWarning: invalid value encountered in scalar divide
cv = np.std(isi)/np.mean(isi)

```

```

[ ]: plt.plot(rate_range, cv_list, 'coral')
plt.xlabel('Stimulus rate (Hz)')
plt.ylabel('Coefficient of variation')
plt.savefig('CV_plot.pdf')
plt.show()

```

