AE-410 GPS (Assignment-1 Acquisition)

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Some important Relations / Properties

- $C_i(\eta-\tau)$ is the PRN code (similar to a vector/array with 1023 dimensions/elements) for i-th satellite, delayed by τ chips
- $C_i(\eta) = C_i(\eta + 1023)$ i.e. PRN codes are periodic with period = 1023 chips = 1 millisecond
- $\frac{C_i(\eta) \oplus C_j(\eta au)}{1023} = 1$ {iff i=j and au = 0, = 0 otherwise}
- $\bullet \hspace{0.1in} \oplus$ in the above point denotes correlation, i.e. element wise multiplication:
 - $C_i(\eta) \oplus C_j(\eta \tau) = \sum_{r=0}^{r=1022} C_i(\eta)[r] \cdot C_j(\eta \tau)[r]$
 - $C_i(\eta)[r]$ is a scalar value having 0 or 1 value

Theory for serial search

• we know the signal received at the antenna is:

 $s(t) = \sqrt{2P}D(t-\tau)C_j(t-\tau)cos((w_L+w_D)t+\theta)$ where : - P is power - τ is delay ($\tau = \tau_{int} + \tau_{frac}$) - w_L is L-5 band frequency and w_D is doppler frequency

• After RF front-end processing we get:

$$I = \sqrt{2P_1}D(t-\tau)C_i(t-\tau)cos((w_D)t + \theta)$$

$$Q = \sqrt{2P_1}D(t- au)C_j(t- au)sin((w_D)t+ heta)$$

where: - I is Inphase samples - Q is Quadrature samples

Note: slight abuse of notation is used as we defined C for discrete values while here we are considering it as continous, but this discrepancy will be resolved in program as we have input data as discrete samples

Now,

for serial search we need to get 3 parameters

- i the satellite number
- $\tau_{frac} \in [0, 1022]$
- $w_D \in [-10kHz: 500Hz: 10kHz]$

Steps

- Firstly, we will rewrite signal as $S=I+Qj=\sqrt{2P_1}D(t-\tau)C(t-\tau)e^{jw_dt}$
- Then, we will multiply S by $C(t-\hat{\tau})e^{-j\hat{w}_dt}$
- We will get:

$$Corr = \sqrt{2P_1}D(t-\tau)C_j(t-\tau) \oplus C_i(t-\hat{\tau})e^{j(w_d-\hat{w}_d)t}$$

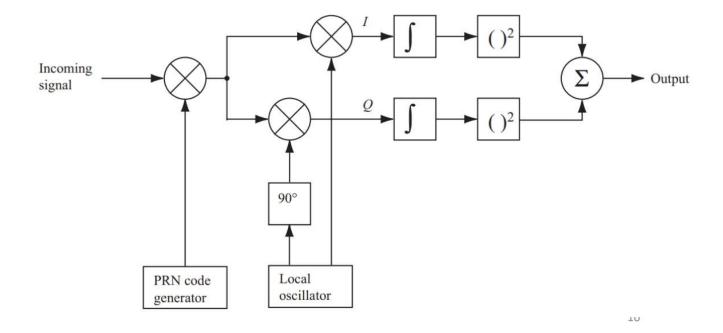
it will attain it's maximum when:

$$w_d - \hat{w}_d = 0$$

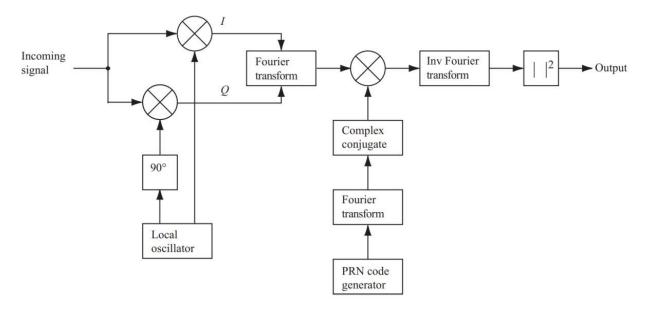
$$\tau=\hat{\tau}$$

i=j= satellite number of incoming signal

Serial Search



Parallel code phase search



Data set utilized

RTL_SDR The data set recorded with a RTL_SDR. This collection includes the GPS L1 band with 2.048MHz bandwidth and 2.048MSPS@8bit I/Q sampling rate. The data was collected on Sep 10, 2017 also in Oegstgeest, Netherlands. The acquired signals belong to the GPS L1 band (PRN 5, 13, 15, 20, 28, 30).

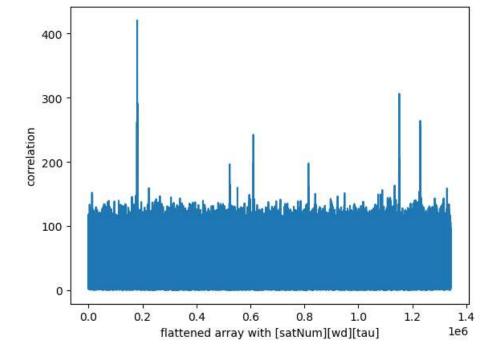
link for the dataset: https://sdr.ion.org/RTL_SDR/RTLSDR_Bands-L1.uint8

Implementation

```
In []: # importing necessary Libraries
    import numpy as np
    import matplotlib.pyplot as plt
    import scipy
    import numba as nb

In []: # constants
    n = 2 # times sampLing frequency
    wds = np.linspace(-10000,10000,41)
    t = np.linspace(0,1,1023*n)*1e-3
    # GPS PRN Extraction
    codesFile = scipy.io.loadmat("codes_L1CA.mat")
    codes = np.array(codesFile['codes_L1CA'])
```

```
In [ ]: # reading datafile having sampling frequency 1*1.023 MHz
        raw_data = np.fromfile("RTLSDR_Bands-L1.uint8", dtype=np.uint8)
In [ ]: # took 2*2*1023 data points because we need 2*1023 data points for each I and Q
        samples = raw_data[0:4092]
In [ ]: # stadard format of sdr data defined on the website
        I = samples[0::2]
        Q = samples[1::2]
In [ ]: # converting to float to have easier operations
        I= I.astype(np.float64)
        Q= Q.astype(np.float64)
In [ ]: # reducing it to ensure the sign bits are intact
        I = I - 128
        Q = Q - 128
In [ ]: # normalizing the chips
        I = I/128
        Q = Q/128
In [ ]: # Serial search
        def modPRN(PRN,n,tau):
            function to get modified PRN for some sampling frequency & delay
                - PRN is the PRN code for single satellite with 1023 length
                - n = SamplingFrequency/1.023e6
                - PRN code for 1 millisecond with length = 1023*n
                - This PRN code is delayed by tau chips
            # return np.roll(np.repeat(PRN,n),-tau)
            return np.repeat(np.roll(PRN,-tau),n)
        def serialSearch(I, Q, knownPRNs, Fs, wds=wds):
            An algo that serially finds correlation of input I and Q with all the
            PRN-ID (satellite number), wds and delays
            rcvdSignal = np.array(I+Q*1j)
            n = int(Fs/1.023e6)
            t = np.linspace(0,1,n*1023)*1e-3
            totalSatellites = knownPRNs.shape[1]
            corr = np.zeros((totalSatellites,len(wds),1023))
            for x in range(0,totalSatellites):
                for y in range(0,len(wds)):
                    for z in range(0,1023):
                         generatedSignal = modPRN(knownPRNs[:,x],n,z)*np.array(np.exp(-1j*wds[y]*t*2*np.pi))
                         corr[x,y,z] = np.abs(np.sum(generatedSignal*rcvdSignal))
            return corr
In [ ]: %%time
        corrSerial = serialSearch(I,Q,codes,2*1.023e6)
       CPU times: total: 45 s
       Wall time: 3min 53s
In [ ]: corrSerial = corrSerial
In [ ]: plt.plot(corrSerial.flatten())
        plt.ylabel("correlation")
        plt.xlabel("flattened array with [satNum][wd][tau]")
Out[]: Text(0.5, 0, 'flattened array with [satNum][wd][tau]')
```



In the above figure it is clearly visible that there are some peaks, if we want to extract the indices of those peaks, we can set some threshold eg. threshold = 180

```
peaksIndices = np.argwhere(corrSerial>180)
In [ ]: peaksIndices
Out[]: array([[
                      11, 330],
                  4,
                      12, 330],
                  4, 13, 330],
               [ 12,
                     19, 325],
               [ 14,
                      22, 173],
                 14,
                      23, 173],
                [ 19,
                      17, 858],
                [ 27,
                      17, 729],
                [ 27,
                      18, 729],
               [ 27,
                     19, 729],
                 29,
                      12, 525],
                [ 29, 13, 525]], dtype=int64)
```

It is clear that we are able to extract the PRN-IDs, since the index starts from 0 so instead of 5, 13, 15 these are 4, 12, 14

Parallel code phase search

For this we will be leveraging numba, which is a **just in time compiler" for python, it can handle parallel computations easily

Note:

The following code will not work if you don't have cuda setup with numba

```
CPU times: total: 8.86 s
       Wall time: 38.2 s
In [ ]: corrParr = np.sqrt(corrR**2+corrI**2)
In [ ]: plt.plot(corrParr.flatten())
Out[]: [<matplotlib.lines.Line2D at 0x161941051d0>]
        400
       300
       200
       100
          0
                                                                       1.2
              0.0
                        0.2
                                 0.4
                                           0.6
                                                    0.8
                                                              1.0
                                                                                 1.4
```

We can clearly see that we have gotten the same plot, we can again set a similar threshold and can get the required params

1e6

```
In [ ]: peaksIndicesParr = np.argwhere(corrParr>180)
In [ ]: peaksIndicesParr
Out[]: array([[ 4, 11, 330],
                  4, 12, 330],
                  4,
                     13, 330],
               [ 12,
                      19, 325],
                 14,
                      22, 173],
                      23, 173],
                 14,
                 19,
                      17, 858],
                 27,
                      17, 729],
                 27,
                      18, 729],
                 27,
                      19, 729],
                 29,
                      12, 525],
                [ 29, 13, 525]], dtype=int64)
```

It can also be noted that the parallel code phase search is much faster than serial search by 10x

Note:

In []: **%%time**

corrR,corrI = parallelSearch(I,Q,codes,wds,n,a)

for delay, I have used it as the time before which satellite has emitted this data, so if we compare, it can be (1023-delay) which we take positive many times