Capon beamforming: Algorithm and Implementation



Algorithm descriptions

Let s(t) be the incoming waves after mixing to baseband, the sensor array signal to be processed is given by

$$X(t) = A(\theta)s(t) + n(t)$$

where $A(\theta) = (a(\theta_1), ..., a(\theta_M))$ is the steering matrix $a(\theta) = (e^{j2\pi y} e^{\sin(\theta)}, ..., e^{j2\pi y} e^{\sin(\theta)})$ is the steering vector M is number of angle bins y_n is the sensor position normalized by wavelength

Capon BF approach is $\theta_{capon} = argmin_{\theta} \{ trace(A(\theta)^*R_n^{-1} *A(\theta)^H \}$ where R_n is the spatial covariance matrix



Implementation details

- Assumptions:
 - Receive antennas are equally spaced with distance $\lambda/2$
- Assuming slow motion scenario, Rn can be constructed using multiple chirps within a frame.
- Calculation of the solution can be significantly reduced with a(θ) constructed for equally spaced antenna, combining with the fact that Rn is an Hermitian matrix that is also persymmetric.
- Current implementation of the module consists of the following sub blocks:
 - Static clutter removal is option is added before Rn matrix generation by removing DC components per range bin. Input assumes 16-bit I/Q, and output is also in 16-bit I/Q format (trade precision for memory and cycles).
 - Per range bin, construct Rn using multiple chirps within a frame. Then Rn is inverted and the upper diagonal of the R_n-1 is stored in memory for each range bin.
 - Per range bin, calculate the Capon BF solution and store the angle spectrum in memory to construct the range-azimuth heatmap.
 - Since conventional BF using covariance matrix has very similar solution $\theta_{conventional} = argmax_{\theta} \{trace(A(\theta)^*R_n^*A(\theta)^H)\}$, we have a fallback flag to use the same code to calculate conventional BF.
 - After detection, per detected point, estimate Doppler using the Capon beamweights and Doppler FFT.
 - Hardcoded for 4 and 8 antennas, and focused on 8-antenna test



Initial Benchmarks for Capon BF (C674x)

- Loop summary:
 - Range-Azimuth heatmap generation, per range bin
 - Clutter removal: mean calc: ii = 5 per 8 samples (chirp samples), remove mean: ii = 2 per 4 samples (chirp samples)

total: (5/8+2/4)*8*128 = 1152

 Rn calculation: diagonal: ii = 6 for 8 samples (chirp samples) off-diagonal: ii = 6 for 4 samples (chirp samples)

total: [8 * (6/8) + 28 * (6/4)] * 128 = 6144

- Rn inverse: total 4000 (measured by Cesar)
- Heatmap calc: ii = 15 per angle bin total 15*64 = 960
- Heatmap total: 12256 per range bin
- Doppler Est, per detected Obj:
 - beamforming output: ii = 17 per chirp samples for all antennas total: 17*128 = 2176
 - 128 size FFT : total 1400 cycles per DSPLIB formula
 - peak search: ii = 2 per Doppler bin total 2* 128 = 256
 - Doppler Est total: 3832 per detected Obj
- Measurement from 6745 Cycle-Accurate Simulator
 - Flat memory CPU cycles:
 - Heatmap: 12910 per range bin
 - DopplerEst: 4606 per detected Obj
 - Cycle accurate cycle measurement with cache misses simulated
 - Heatmap: 22122 per range bin
 - DopplerEst: 7792 per detected Obj



Memory Usage

- Data memory usage:
 - Range-Azimuth heatmap: Nrange * Nangle * 4bytes/float
 - Inv(Rn) for all range bins: Nrange * [Nant * (Nant+1)/2] * 8bytes/complex-float
 - Memory for local instances (twiddle etc): DopplerFFTsize*8bytes/complexfloat + 48 * 4bytes
 - Scratch memory: 8-antenna:max{2*DopplerFFTsize*8bytes, 360*4bytes },
 4-antenna:max{2*DopplerFFTsize*8bytes, 82*4bytes }
- Code size:
 - Signal processing blocks: ~21000 bytes
 - Module Initialization functions: 1100 bytes

