

Direction of Arrival Analysis on a Mobile Platform

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Objectives

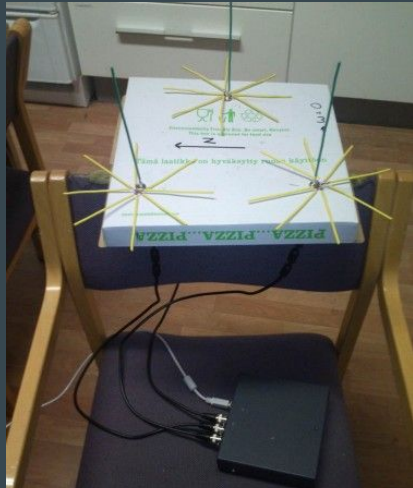
- Find a transmitter
- Be mobile



Previous Work

Tatu Peltola - 3 RTL
dongles

<https://www.youtube.com/watch?v=8WzblmgZ0EE>

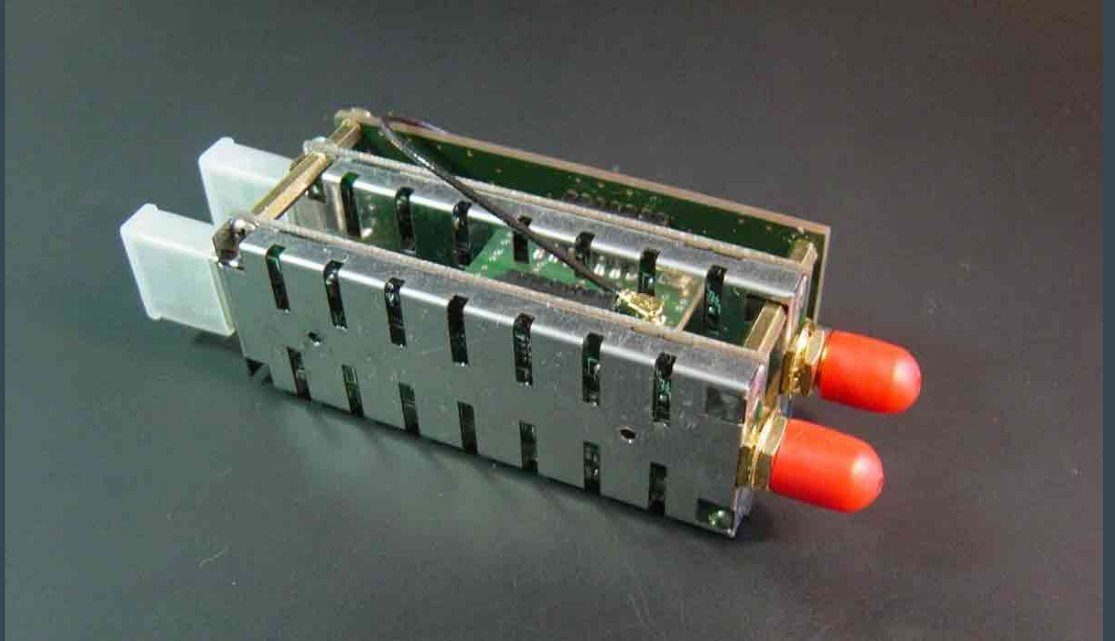


Our Hardware

- 2 RTL-SDR

<http://coherent-receiver.com>

- Moto X Pure



Previous Work - Methods

- MUSIC IEEE Trans. Antennas Propagation, Vol. AP-34 (March 1986), pp.276-280.
- MN (Min Norm) <http://ieeexplore.ieee.org/document/4350746/>
- MVDR (Capon) J. Capon “High-Resolution Frequency-Wavenumber Spectrum Analysis”
- Cross Correlation
- TDOA (time difference of arrival)

Cross Correlation

Didn't give us the DOA results we wanted, but played an important role in synchronizing our receivers.

$$r_p(T) = \sum_{t=-\infty}^{\infty} x_1(t)x_2(t+T)$$

$$\tau = \operatorname{argmax}(r_p(T))$$

x_1, x_2 : signals from receivers

r : cross correlation

τ : the sample difference

Cross Correlation

We can reduce the needed number of computations significantly by completing this problem in the frequency domain.

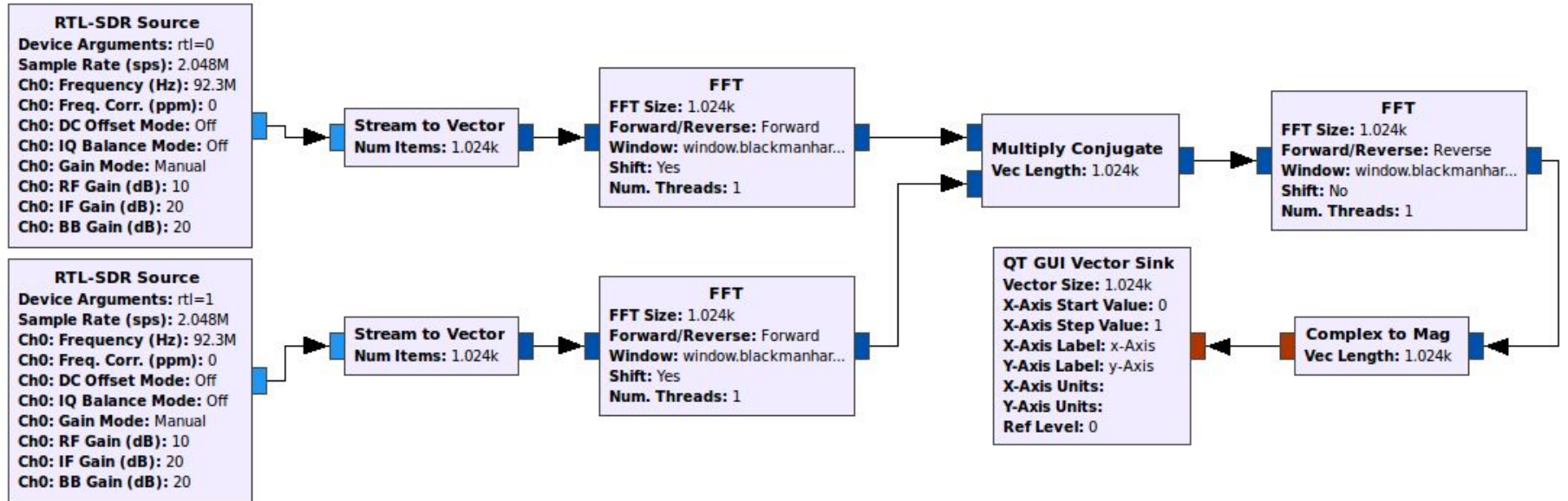
$$x_1(t) \Leftrightarrow X_1(f)$$

$$x_2(t) \Leftrightarrow X_2(f)$$

$$r_p(t) \Leftrightarrow R_p(f)$$

$$R_p(f) = X_1(f) \times X_2(f)^*$$

Cross Correlation



Cross Correlation

For electromagnetic waves traveling at the speed of light and a sample rate of 2 MHz, this solution is able to resolve differences of..

$$\begin{aligned} \textit{Resolution} &= c \times \textit{SamplePeriod} \\ &= \frac{c}{\textit{SampleRate}} \\ &= \frac{3e8}{2e6} = 150 \text{ meters} \end{aligned}$$

Sub sample resolution can be achieved through interpolation

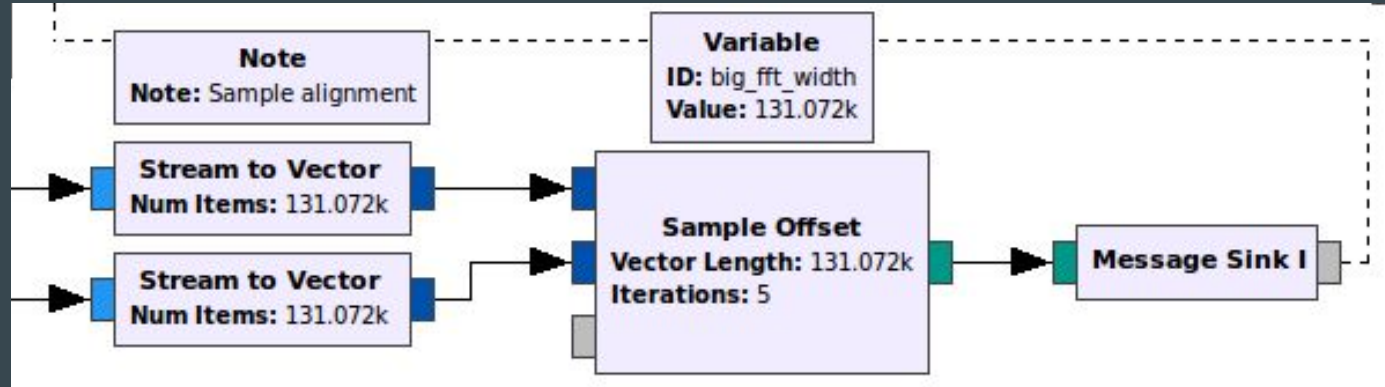
Why bring up Cross Correlation?

The RTL-SDR receivers don't start at the same time (a sample offset).

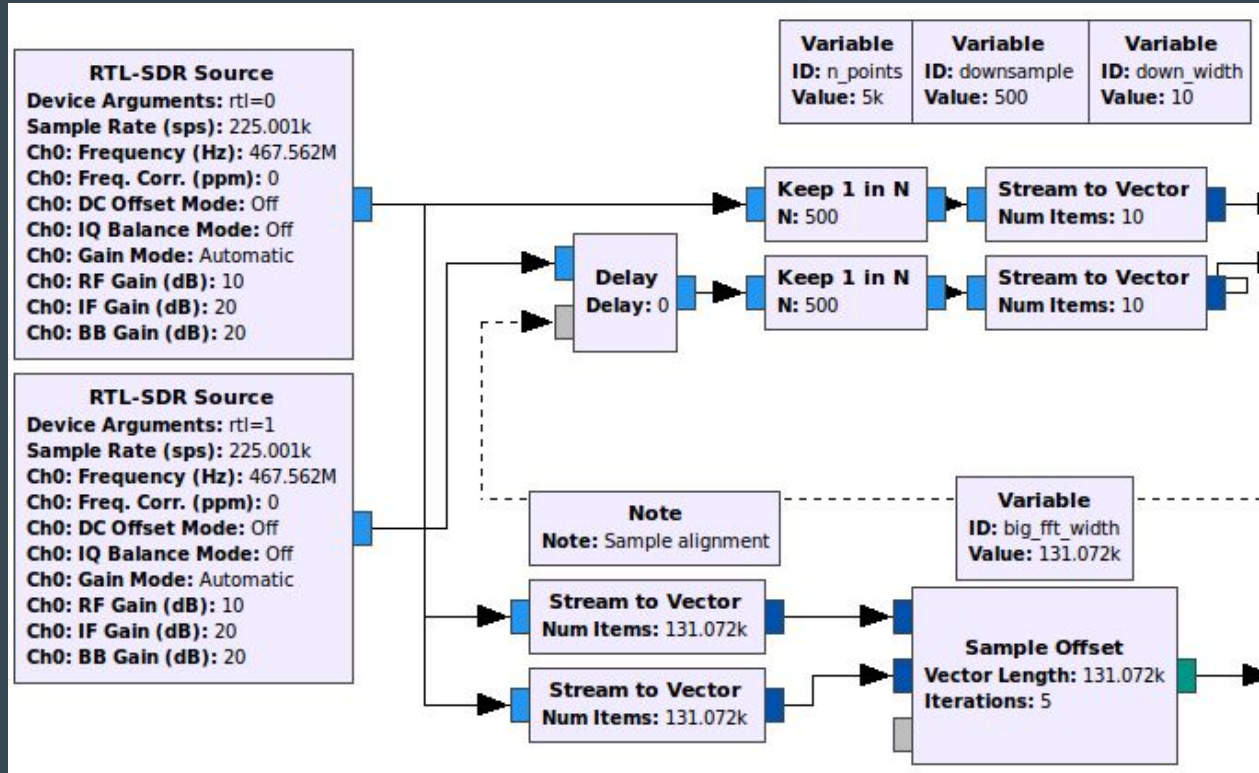
We can correct this by delaying one of the signals until they are roughly aligned in time. The delay we need to apply is the value we get from cross correlation.

Sample Offset Block

- Median sample offset from several cross-correlations
- Stops after n iterations
- Can be reset via message



Sample Offset Block



Eigenvector method

- Phase difference method
- Still works after downsampling to less than 1 kS/sec
- Steer our phased array with a vector until we find the most power

$$\max w^H R w$$

$$\text{s.t. } w^H w = 1$$

Eigenvector method

1. Estimate covariance matrix
2. Find maximum eigenvalue
3. Return the argument of the eigenvector

Eigenvector method (Not Complicated)

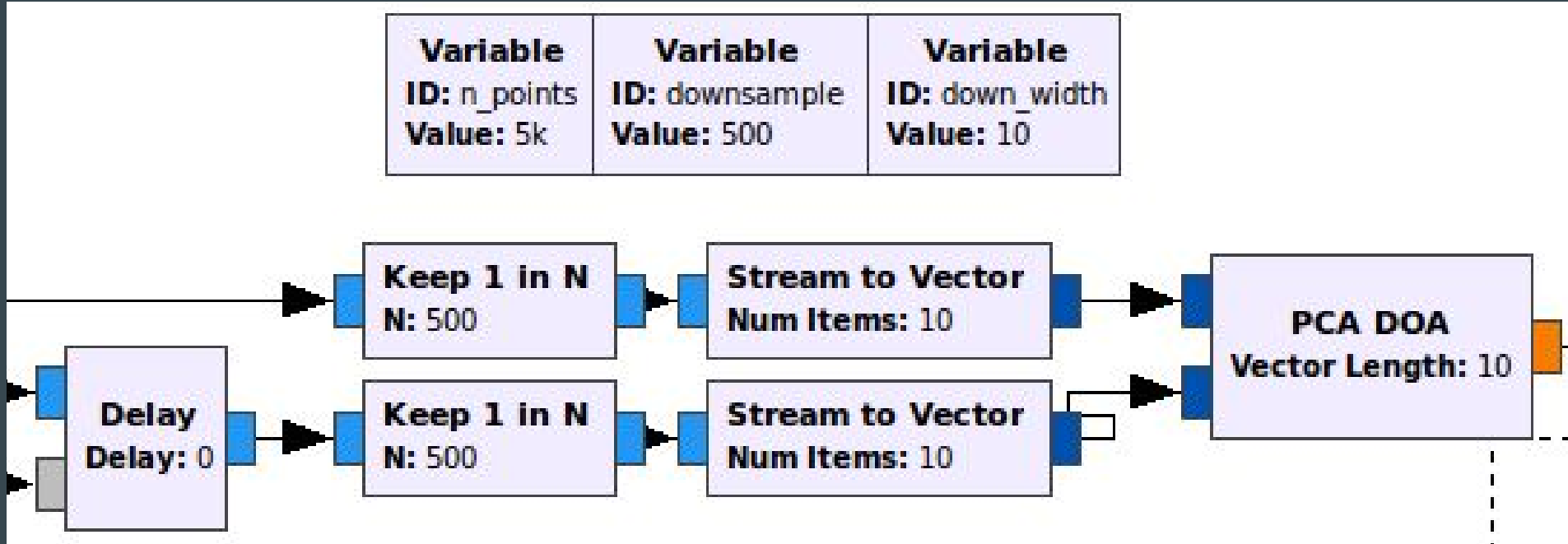
Analytical solution for the eigenvalue problem.

```
// find eigenvalues
gr_complex lambda0 = (covar[0][0]+covar[1][1]+sqrt(pow((covar[0][0]+covar[1][1]),2)-gr_complex(4,0)*(covar[0][0]*covar[1][1]-covar[0][1]*covar[1][0])))/gr_complex(2,0);
gr_complex lambda1 = (covar[0][0]+covar[1][1]-sqrt(pow((covar[0][0]+covar[1][1]),2)-gr_complex(4,0)*(covar[0][0]*covar[1][1]-covar[0][1]*covar[1][0])))/gr_complex(2,0);
gr_complex max_lambda = (abs(lambda0) > abs(lambda1)) ? abs(lambda0) : abs(lambda1);
```

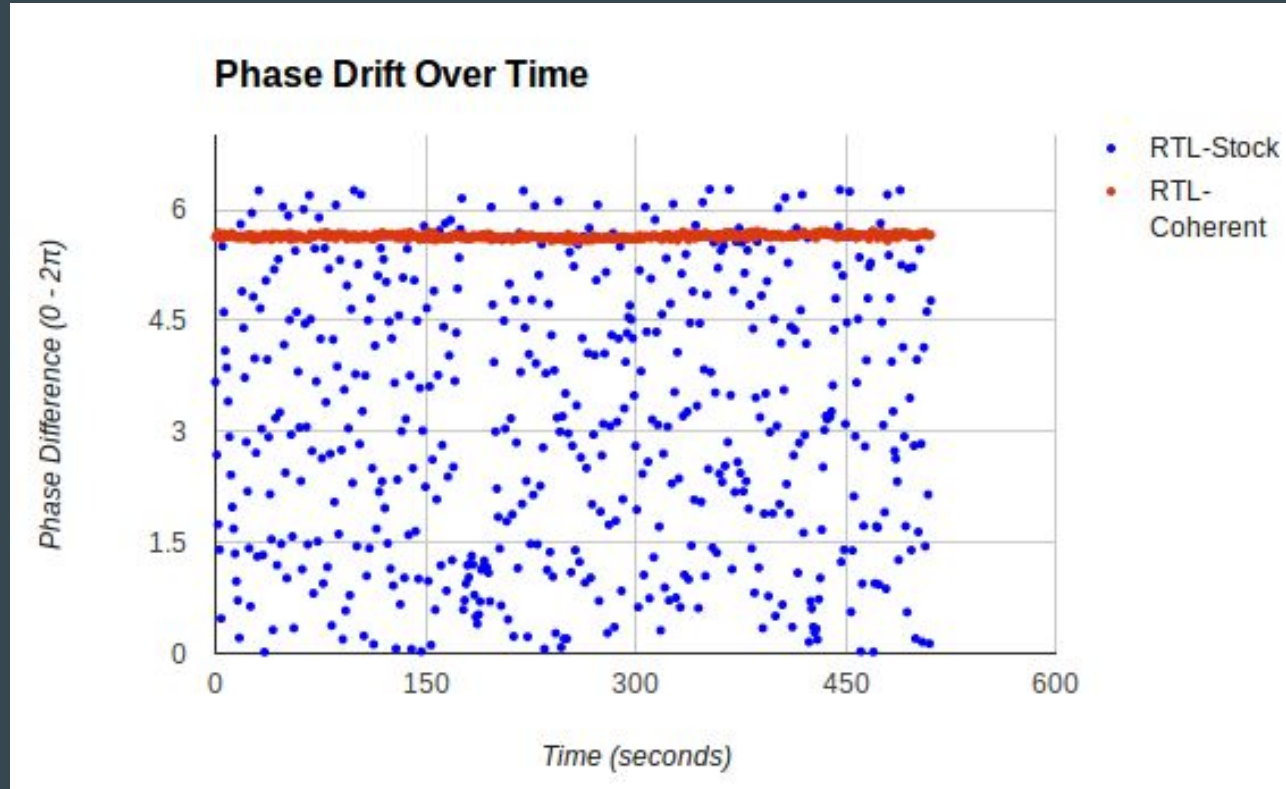
Return the argument of the eigenvector

```
94  return std::arg((max_lambda - covar[0][0])/covar[0][1]);
95 }
```

Eigenvector method

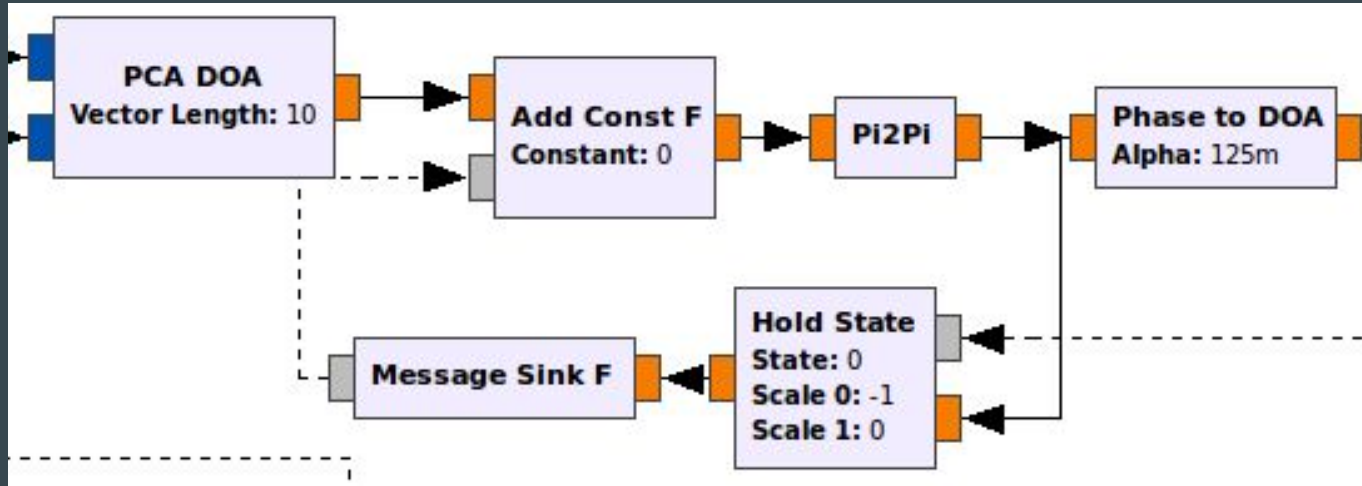


Phase Coherency



Phase Coherency

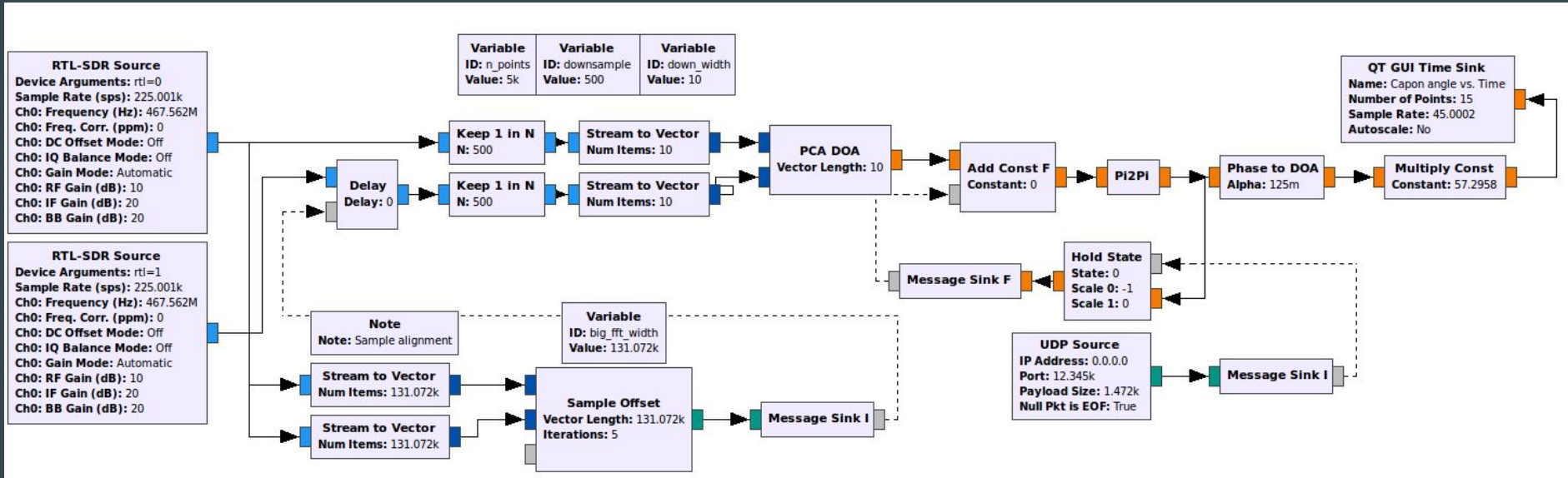
- A constant phase offset can be corrected with addition



Which blocks are custom?

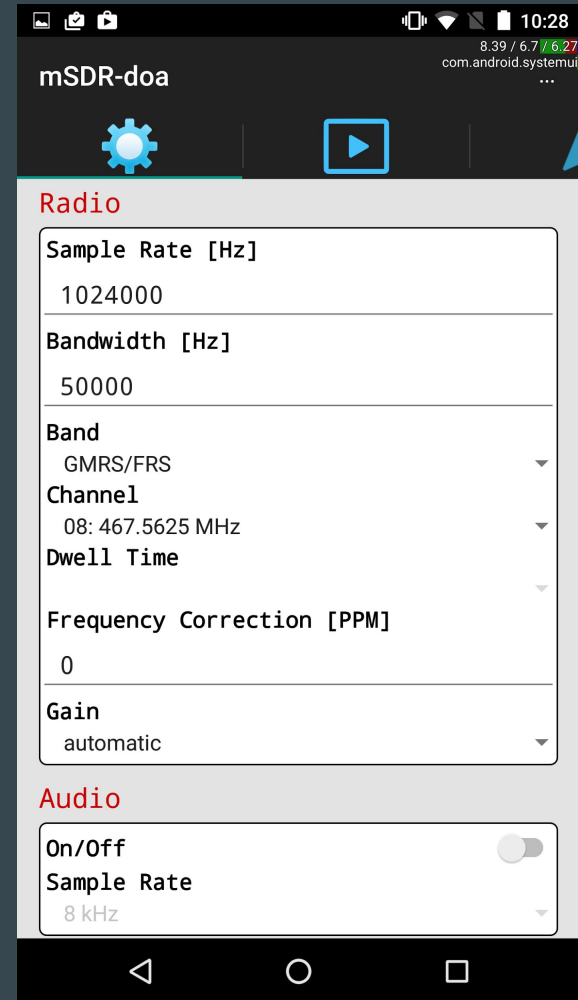
- *Sample Offset* - cross correlation once (or n times)
- *PCA DOA* - returns phase difference
- *Pi2Pi* - wraps a float into the range from $-\pi$ to π
- *Phase to DOA* - map from phase to DOA with arc cosine
- *Hold State* - holds/updates a calibration value
- *Message sinks* - turns a value into a message
- *Delay* - added message control
- *Add Const F* - message control for adding a float constant

Complete Flowgraph



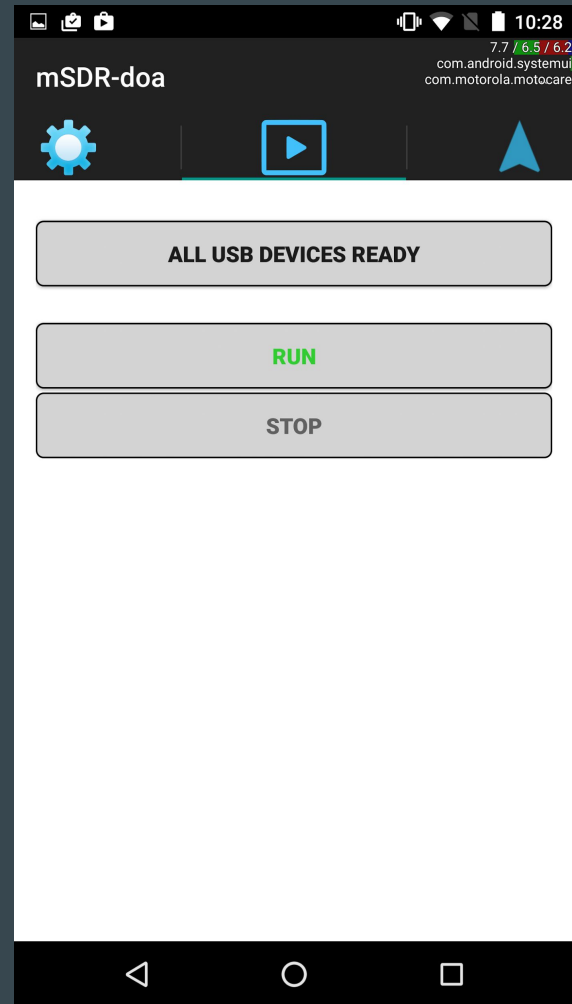
Making it Mobile

- Gather parameters in Java UI



Making it Mobile

- Cross-compile C++ flowgraph to statically linked executable
- Call it using Android Java Native Interface wrapper



Making it Mobile

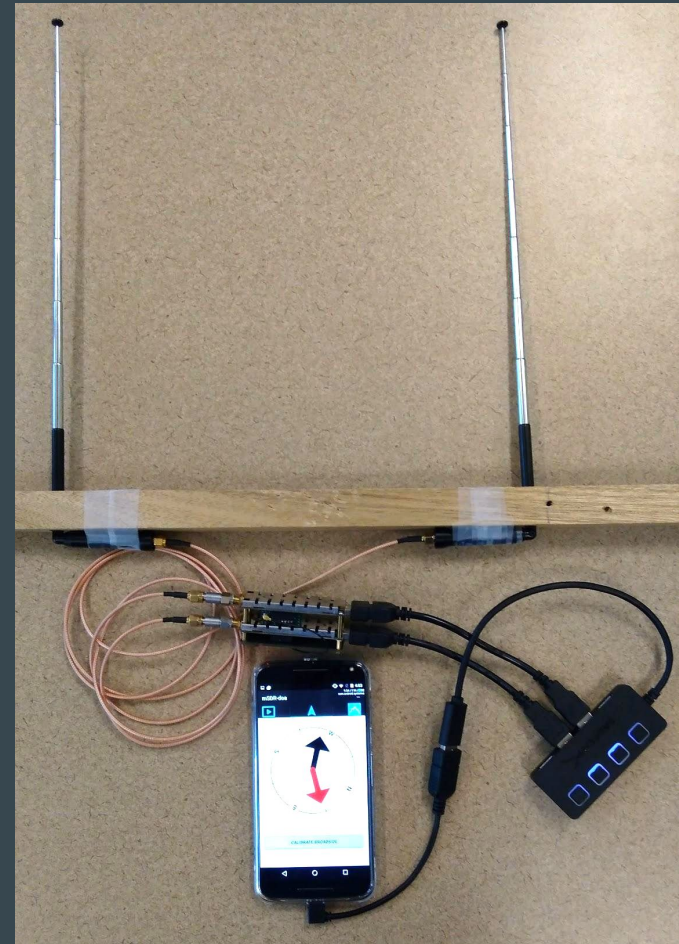
- UDP packets for data, control between Java and Gnuradio
- Now we can add maps, headings, arrows etc.

Red arrow represents the direction ambiguity due to using a phase difference method



The Prototype

- Finest-quality plywood and tape
- Half-wavelength spacing (FRS band)
- USB hub for RTL-SDR dongles



Testing

- Walkie Talkies (FRS Band)
- FM Radio Stations

<http://a.co/goIlb5u>



Videos

- <https://www.youtube.com/watch?v=aUubSNUxX8g>



Videos

- <https://www.youtube.com/watch?v=RKDnkXq2mVM&t=8s>



Code

- <https://github.com/samwhiting/gnuradio-doa>

Thanks

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Template app by Dave Mahoney

Auxiliary Slides

Principal Component Analysis

Our incoming signal $s(t)$ is received by two different antennas as $x_1(t)$ and $x_2(t)$. We'll call $x_1(t)$ our reference, and let $x_2(t)$ have a delay represented as τ .

$a(t)$ = baseband signal

$s(t)$ = modulated signal

$$s(t) = a(t)\cos(\omega t)$$

$$x_1(t) = s(t) = a(t)\cos(\omega t)$$

$$x_2(t) = s(t - \tau) = a(t - \tau)\cos(\omega(t - \tau))$$

Principal Component Analysis

Use the narrow-band assumption to say that the primary difference between the two signals can be represented by a phase shift in the carrier frequency.

$$x_1(t) = s(t) = a(t)\cos(\omega t)$$

$$x_2(t) = s(t - \tau) = a(t - \tau)\cos(\omega(t - \tau))$$

$$x_2(t) = a(t - \tau)\cos(\omega t - \phi)$$

$$\text{Where } \phi = \omega\tau$$

$$\tilde{x}_1(t) = a(t)$$

$$\tilde{x}_2(t) = a(t)e^{-j\phi}$$

Principal Component Analysis

Stack our signals into a vector

$$\bar{x}(t) = \begin{bmatrix} \tilde{x}_1(t) \\ \tilde{x}_2(t) \end{bmatrix}$$
$$\bar{x}(t) = a(t) \begin{bmatrix} 1 \\ e^{-j\phi} \end{bmatrix}$$

Principal Component Analysis

Set up a beam steering vector, and apply it to our signal.

$$y(t, \hat{\phi}) = \begin{bmatrix} 1 \\ e^{-j\hat{\phi}} \end{bmatrix}^H \bar{x}(t)$$

$$y(t, \hat{\phi}) = \begin{bmatrix} 1 \\ e^{-j\hat{\phi}} \end{bmatrix}^H a(t) \begin{bmatrix} 1 \\ e^{-j\phi} \end{bmatrix}$$

$$y(t, \hat{\phi}) = a(t)(1 + e^{j(\hat{\phi}-\phi)})$$

$y(t, \hat{\phi})$ has a maximum when $\hat{\phi} = \phi$

Principal Component Analysis

Take the variance of y in order to look at power of the filter output.

$$\text{let } w = \begin{bmatrix} 1 \\ e^{-j\hat{\phi}} \end{bmatrix}$$

with a zero-mean signal, $\sigma_y^2 = E|y|^2$

$$E|y|^2 = E|w^H \bar{x}(t)|^2 = E[(w^H \bar{x})(w^H \bar{x})^*]$$

$$= E[w^H \bar{x} \bar{x}^H w] = w^H E[\bar{x} \bar{x}^H] w = w^H R_{xx} w$$

Principal Component Analysis

Maximize the power by changing the steering vector with a constraint to limit the gain.

$$\max w^H R w$$

$$\text{s.t. } w^H w = 1$$

Principal Component Analysis

Setup the constrained optimization problem with Lagrange multipliers.

$$\nabla f = \lambda \nabla g$$

$$\frac{d}{dw} w^H R w = \lambda \frac{d}{dw} (w^H w - 1)$$

$$R w = \lambda w$$

Principal Component Analysis

We are left with an eigenvalue problem.

$$Rw = \lambda w$$

$$w(R - \lambda I) = 0$$

Eigenvalue/vector pairs

$$\lambda_1 \leftrightarrow E_1$$

$$\lambda_2 \leftrightarrow E_2$$

Principal Component Analysis

The maximum eigenvalue has an eigenvector that provides us with our optimal steering vector. This can be normalized, and the phase angle extracted.

$$\lambda_1 \leftrightarrow E_1$$

$$\lambda_2 \leftrightarrow E_2$$

$$\max [\lambda_1, \lambda_2] = \lambda \leftrightarrow E$$

$$E = \begin{bmatrix} u \\ v \end{bmatrix} = c \begin{bmatrix} 1 \\ e^{j\hat{\phi}} \end{bmatrix}$$

```

57 float capon_ccf_impl::direction(const gr_complex* in0, const gr_complex* in1) {
58     // put in0 and in1 into vectors x1 and x2
59     std::vector<gr_complex> x1(in0, in0+d_vector_size);
60     std::vector<gr_complex> x2(in1, in1+d_vector_size);
61
62     // find average value of each one
63     gr_complex x1_avg = accumulate(x1.begin(), x1.end(), gr_complex(0,0)) / gr_complex(d_vector_size,0);
64     gr_complex x2_avg = accumulate(x2.begin(), x2.end(), gr_complex(0,0)) / gr_complex(d_vector_size,0);
65     // subtract the average value from each element
66     for (int i=0; i<d_vector_size; ++i) { x1[i] -= x1_avg; }
67     for (int i=0; i<d_vector_size; ++i) { x2[i] -= x2_avg; }
68     // find conjugates of x1 and x2
69     std::vector<gr_complex> x1_c, x2_c;
70     x1_c.reserve(d_vector_size);
71     x2_c.reserve(d_vector_size);
72     for (int i=0; i<d_vector_size; ++i) { x1_c.push_back(conj(x1[i])); }
73     for (int i=0; i<d_vector_size; ++i) { x2_c.push_back(conj(x2[i])); }
74     // multiply vectors together
75     std::vector<gr_complex> c11, c12, c22;
76     c11.reserve(d_vector_size);
77     c12.reserve(d_vector_size);
78     c22.reserve(d_vector_size);
79     transform(x1.begin(), x1.end(), x1_c.begin(), back_inserter(c11), std::multiplies<gr_complex>());
80     transform(x1.begin(), x1.end(), x2_c.begin(), back_inserter(c12), std::multiplies<gr_complex>());
81     transform(x2.begin(), x2.end(), x2_c.begin(), back_inserter(c22), std::multiplies<gr_complex>());
82     // make covariance matrix
83     std::vector<std::vector<gr_complex> > covar(2, std::vector<gr_complex>(2, 0));
84     covar[0][0] = accumulate(c11.begin(), c11.end(), gr_complex(0,0)) / gr_complex(d_vector_size,0);
85     covar[0][1] = accumulate(c12.begin(), c12.end(), gr_complex(0,0)) / gr_complex(d_vector_size,0);
86     covar[1][1] = accumulate(c22.begin(), c22.end(), gr_complex(0,0)) / gr_complex(d_vector_size,0);
87     covar[1][0] = conj(covar[0][1]);

```


Capon Method

Analytical solution for the eigenvalue problem.

```
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gr_complex lambda1 = (covar[0][0]+covar[1][1]-sqrt(pow((covar[0][0]+covar[1][1]),2)-gr_complex(4,0)*(covar[0][0]*covar[1][1]-covar[0][1]*covar[1][0])))/gr_complex(2,0);
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Return the argument of the eigenvector

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94         return std::arg((max_lambda - covar[0][0])/covar[0][1]);
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