24.1 Introduction: Trilateration with multiple beacons

In the previous notes we learned how to find the distance from a beacon to a receiver using cross-correlation. Then we used trilateration to combine several of these measurements and find the the location of the receiver. However, we only looked at cross-correlation for finding the distance to one beacon, but we know that we need at least 3 beacons to find our location in 2D. How can we detect our distances from many beacons simultaneously?

In this note, we’ll introduce an algorithm called **orthogonal matching pursuit (OMP)** that will allow us to unmix the signals from multiple beacons.

This algorithms isn’t only useful for estimating distances – in addition, the beacons can encode messages in the signals they send, and using OMP we can recover these messages even when multiple beacons are transmitting at the same time.

24.2 Orthogonal Matching Pursuit (OMP)

We’ll walk through the OMP algorithm using an example with three beacons. At the end of the note we’ll provide a general formulation.

Consider three beacons (in GPS, these would be satellites) which are each transmitting their own unique periodic code. We’ll denote these codes $\vec{s}_1$, $\vec{s}_2$, and $\vec{s}_3$. Each of these is length $N = 10$. For concreteness, let’s say that the codes are as follows:

$$
\vec{s}_1 = \begin{bmatrix} -4 \\ -1 \\ 1 \\ -1 \\ 1 \\ 0 \\ -3 \\ 4 \\ -4 \\ -3 \end{bmatrix}, \quad \vec{s}_2 = \begin{bmatrix} -4 \\ -5 \\ -4 \\ -4 \\ -2 \\ -2 \\ 4 \\ 2 \\ 1 \\ 3 \end{bmatrix}, \quad \vec{s}_3 = \begin{bmatrix} -2 \\ -1 \\ -1 \\ -4 \\ -4 \\ 5 \\ 2 \\ 4 \\ -4 \\ 1 \end{bmatrix}
$$

Beacon 1 transmits $\vec{s}_1$ repeatedly, beacon 2 transmits $\vec{s}_2$ repeatedly, and so on. In addition, the beacons can each multiply their code by a scalar to encode additional information. For example, each beacon could tell the receiver the temperature at its location by multiplying its code by the temperature. (This could be helpful in determining which parts of the walls or doors need more insulation).
Our receiver is a different distance away from each beacon, so it receives a circularly shifted version of each code. It can’t distinguish between the signals sent by the different beacons, so it measures the sum of the three incoming signals. We measure the first \( N = 10 \) samples at the receiver. Putting all of this together, we can model our received signal, \( \bar{r} \), as

\[
\bar{r} = a_1 \bar{s}_1(\tau_1) + a_2 \bar{s}_2(\tau_2) + a_3 \bar{s}_3(\tau_3)
\]

(1)

where \( \bar{s}_1(\tau_1) \) is \( \bar{s}_1 \) circularly shifted by \( \tau_1 \) samples. The \( a \) values are the scalars that the beacons can use to send messages.

From our known codes \( \bar{s}_1, \ldots, \bar{s}_3 \) and our measured receiver signal \( \bar{r} \), we’d like to find the coefficients \( a_1, \ldots, a_3 \) and the shifts \( \tau_1, \ldots, \tau_3 \).

Suppose our received signal is:

\[
\bar{r} = \begin{bmatrix} -11 & 2 & -3 & 12 & 4 & 0.5 & 6.5 & -11 & -12 & -12.5 \end{bmatrix}^T
\]

We will now decompose it into the scaled and shifted versions of the beacon codes with OMP. OMP is an \emph{iterative} method, which means we will repeat the same set of steps several times before reaching an answer.

**Iteration 1**

We know that our receiver signal consists of the sum of several shifted and scaled codes. First we will find the code (and it’s corresponding shift) that is most similar to the received signal. To do this, we’ll take the circular cross correlation of \( \bar{r} \) with each of the different possible codes \( (\bar{s}_1, \bar{s}_2, \bar{s}_3) \). These values are shown below graphically.

![Circular Cross Correlation Graphs](image)

The shifted code that is most similar to the received signal is the one with largest cross-correlation. Therefore, we find the code and shift with the maximum (in absolute value) cross-correlation. Why absolute value? The transmitted code could have been scaled by a negative number, which will result in a negative value of the cross-correlation when these signals are similar. In this example, the maximum is code \( \bar{s}_2 \) at shift 7.

Now we will guess that this shifted code is the only one in the received signal. In other words, we’ll create an estimate of \( \bar{r} \), denoted \( \hat{r} \), which takes the form

\[
\hat{r} = x_1 \bar{s}_2^{(7)}
\]
Now we want to find the coefficient, \( x_1 \). We can find the coefficient by setting up and solving a least squares problem. We define

\[
A = \hat{s}^{(7)}
\]

so that our least square problem and solution are as follows:

\[
A\bar{x} = \hat{r} \\
\bar{x} = (A^T A)^{-1} A^T \hat{r}
\]

Plugging in the values in our example gives

\[
\bar{x} = 2.14
\]

so our estimate of the received signal is

\[
\hat{r} = A\bar{x} = 2.14 \hat{s}_2^{(7)}
\]

However, our estimate of the received signal \( \hat{r} \), does not match our actual received signal. Next we will calculate the difference between the actual received signal and estimated received signal, call the residual. We’ll call the residual signal \( \vec{y} \).

\[
\vec{y} = \vec{r} - \hat{r} \\
\vec{y} = \vec{r} - A\bar{x}
\]

The values of \( \vec{r}, \hat{r}, \) and \( \vec{y} \) for our example are shown below:

The residual is not zero, so we know that there are more codes from different beacons in the received signal. To find them, we will repeat the same process, but instead of starting with our received signal, we’ll start with the residual. The logic is that the first shifted code we find is the strongest and will possibly block us from seeing other weaker codes. We first found this strongest code, and then removed it from the receiver signal to give us a better chance of finding the other weaker codes.

We’re now ready for the second iteration of OMP.

**Iteration 2**

In the first iteration, the first thing we did was cross-correlate the receiver signal \( \vec{r} \) with each code. We’ll do the same thing this time but use the residual \( \vec{y} \) instead of the receiver signal. Below are the circular cross correlations between the residual and each code:
Once again, we will find the shifted code that is most similar to the residual by taking the maximum (in absolute value) of the cross correlation. In this example, we find that the maximum is code $\vec{s}_1$ at shift 3 (it’s a negative value, but biggest in magnitude). Note that the cross-correlation with $\vec{s}_2$ at shift 7 (the maximum from the first iteration) is 0. This is because we completely removed this component from the residual, forcing $\vec{y}$ to be orthogonal to $\vec{s}_2^{(7)}$. This is where the “orthogonal” in “orthogonal matching pursuit” comes from.

Now we update our estimate of the received signal to include the two shifted codes that we’ve found. (Recall that in the first iteration we found $\vec{s}_2$ at shift 7, and in this iteration we found $\vec{s}_1$ at shift 3).

$$\hat{r} = x_1 \vec{s}_2^{(7)} + x_2 \vec{s}_1^{(3)}$$

We don’t assume that $x_1$ is the same value as before, since we want to account for the additional codes that we’ve found. Instead we solve for both coefficients, $x_1$ and $x_2$, using least squares. We setup our least squares problem by defining

$$A = \begin{bmatrix} \vec{s}_2^{(7)} \\ \vec{s}_2^{(3)} \end{bmatrix}, \quad \vec{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

so that

$$A\vec{x} = \hat{r}$$

$$\vec{x} = (A^T A)^{-1} A^T \hat{r}$$

If we plug in the values in our example, we get

$$\vec{x} = \begin{bmatrix} 2.0 \\ -1.12 \end{bmatrix}$$

and our estimate of our received signal is now

$$\hat{r} = 2.0 \vec{s}_2^{(7)} - 1.1 \vec{s}_1^{(3)}$$

Finally, we calculate our residual using our updates estimate $\hat{r}$.

$$\vec{y} = r - \hat{r}$$

$$\vec{y} = r - A\vec{x}$$

EECS 16A, Fall 2018, Note 24
The residual is still not zero, and we know there were three beacons, so we will do a third iteration to find another beacon code.

**Iteration 3**

First, we take the cross-correlation between the new residual and each of the beacon codes.

Next, we find the code and shift with the maximum (in absolute value) cross-correlation. In this example, the maximum is $\mathbf{s}_3$ at shift 4. Once again, notice that the codes we found in previous iterations ($\mathbf{s}_2$ at shift 7 and $\mathbf{s}_1$ at shift 3) have zeros in the cross-correlation. This is because the least squares step ensures that the residual is orthogonal to all previous codes.

Now, we update our estimate of the received signal to include all three shifted codes that we’ve found.

$$\hat{\mathbf{r}} = x_1 \mathbf{s}_2(7) + x_2 \mathbf{s}_1(3) + x_3 \mathbf{s}_3(4)$$

We setup a least squares problem by defining

$$A = \begin{bmatrix} \mathbf{s}_2(7) \\ \mathbf{s}_1(3) \\ \mathbf{s}_3(4) \end{bmatrix}, \quad \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$$

so that

$$A\mathbf{x} = \hat{\mathbf{r}}$$

$$\mathbf{x} = (A^T A)^{-1} A^T \hat{\mathbf{r}}$$
If we plug in the values in our example, we get
\[
\vec{x} = \begin{bmatrix} 2.0 \\ -1 \\ 0.5 \end{bmatrix}
\]
and our estimate of our received signal is now
\[
\hat{\vec{r}} = 2.0\vec{s}_2(7) - 1.0\vec{s}_1(3) + 0.5\vec{s}_3(4)
\]
Finally, we calculate our residual using our updated estimate \( \hat{\vec{r}} \).
\[
\vec{y} = \vec{r} - \hat{\vec{r}}
\]
Our residual is \( \vec{0} \)! We've found all of the codes:
\[
\vec{r} = \hat{\vec{r}} = -1\vec{s}_1(3) + 2.0\vec{s}_2(7) + 0.5\vec{s}_3(4)
\]

### 24.3 Stopping Condition: How many iterations?

In the example above we did three iterations of OMP and at the end the residual was \( \vec{0} \). However, in most examples there will be some noise in the measurement, so the residual will never completely vanish. How do we know how many iterations to run? We have two options:

1. We can run iterations until the residual is small. We set a threshold value \( th \) and when the norm of the residual is less than \( th \), we will assume that the signal contains only noise and we’ll stop iterating.
2. From the problem statement, we may already know how many beacon codes are contained in the received signal, \( k \). Each iteration provides one new code, so the number of codes present will determine the number of iterations. In practice, we may not know the exact number, but we can guess and then do a few more iterations just in case.

We will use both of these criteria and we’ll stop iterating when either is met. In other words, we’ll set a number of iterations \( k \), but we’ll stop before \( k \) iterations if \( \|\vec{r}\| < th \).
24.4 OMP with many beacons

In the previous example, we looked at using OMP with three beacons, each with their own code. However, we can do OMP with many more beacons! These could things like temperature sensors, light-meters, humidity sensors, or something that measures how loudly your cat is purring. As in the previous example, each beacon has its own unique code that it transmits, and the beacons send information by multiplying their codes with a scalar, $a_i$. The value of $a_i$ is the message.

For OMP to work, the each code should be “almost” orthogonal to all of the other codes and to all the shifted versions of itself. (Orthogonal means that the inner product of two codes is exactly zero. “Almost” orthogonal means that the magnitude of the inner product is close to zero.) The “gold codes” that we looked at in the homework are good examples of codes that fulfill these properties and would work well for OMP.

Now imagine we have 2,000 beacons, each with its own code: $\vec{s}_0, \vec{s}_1, ..., \vec{s}_{1999}$. Each message is a real number that scales the song: $a_0, a_1, ..., a_{1999}$, for each of the 2,000 sensors. Each of the messages from the transmitters is delayed by a different amount, given by: $\tau_0, \tau_1, ..., \tau_{1999}$. Each code is periodic with period $N = 400$.

The signal received ($\vec{r} \in \mathbb{R}^{400}$) is a linear combination of shifted codes from all transmitting users given by:

$$\vec{r} = a_0 \vec{s}_0(\tau_0) + a_1 \vec{s}_1(\tau_1) + ... + a_{1999} \vec{s}_{1999}(\tau_{1999})$$

Here, the subscript of $\vec{s}$ indicates which beacon, and the superscript in parenthesis represents the how many samples the code is shifted by.

We have 400 measurements ($\vec{r} \in \mathbb{R}^{400}$) and 2000 unknown values of $a$. How can we possibly determine the messages? We need more information! The additional information is that only a small number (say 10) beacons are transmitting at the same time. This means that only 10 of the beacons have non-zero $a_i$ coefficients, but we don’t know which beacons (and we don’t know their shifts). If we put all of the $a_i$’s in a vector, most of the vector would contain 0’s. We call this a sparse vector and the sparsity level $k$ is the number of non-zero elements.

In the simplest case, there will be only one beacon transmitting data to the receiver, for example. If beacon 130 is sending a message of value 1 with a delay of 10, the received signal will be:

$$\vec{r} = 1 \cdot \vec{s}_{130}^{(10)}$$

Now if we perform cross-correlation of the received signal with the code from beacon 130, $\vec{s}_{130}$, the result will have a peak at shift index 10, with a normalized amplitude of 1, shown below:
If we correlate with all codes and plot the maximum from each one, we get the results shown below. There is a single peak at the beacon ID 130, telling us that beacon 130 was “on” and sending a value of 1.

Imagine beacons 40, 100, and 312 are simultaneously transmitting with delays of 13, 20, and 45 (respectively) and message values of 10, 10, and 8 (respectively). Our received signal is now:

\[
\bar{r} = 10 s_{40}^{(13)} + 10 s_{100}^{(20)} + 8 s_{312}^{(45)}
\]
If we cross-correlate the received signal with every code, find the maximum peak of each cross-correlation, and plot it for each beacon ID (normalizing by the length of the song), we get the plot below.

We can see above, that beacons 40, 100, 312 are all “on”.

Let us look at one more example in which 4 beacons (beacon ID 40, 100, 312 and 350) transmit simultaneously and their corresponding messages are 100, 10, 8 and 0.02. The result of correlating the received signal with all of the different codes is shown below:
The peaks corresponding to sensor 100, 312 and 350 don’t seem to appear at all! Even if we zoom in, we can’t distinguish peaks corresponding to these sensors. What could possibly cause this? Perhaps the very high value of user 40’s made it so that we cannot see the songs from users 100, 312 and 350.

However, as discussed in the previous example, OMP identifies the dominant code in the received signal, then removes it! We can use OMP in this example to recover the weaker codes as well as the dominant ones.

Here is a description of OMP with an example received signal of: 
\[ \vec{r} = a_{40} \vec{s}_{40}^{(13)} + a_{100} \vec{s}_{100}^{(8)} \] where \( a_{40} = 100 \) and \( a_{100} = 10 \).

1. Find the song with the highest correlation with \( \vec{r} \), and find the shift that maximizes the correlation. We’ll denote this shifted song as vector \( \vec{s}_*^{(s)} \), where * indicates the song ID and shift that maximize the cross-correlation. What does it mean for a vector \( \vec{s}_*^{(s)} \) to have the highest correlation with \( \vec{r} \)? The highest correlation between two vectors means the error vector between them is the lowest. We find this vector via a cross-correlation with all the circular shifts of all the songs, the same first step we have always done. This gives us \( \vec{s}_{40}^{(13)} \).

2. Use least squares to solve for \( \vec{x} \) in the equation \( A \vec{x} = \vec{b} \) where \( \vec{b} \) is the received signal \( \vec{r} \) and \( A \) is 
\[
\begin{bmatrix}
\vec{s}_{40}^{(13)} \\
\vec{s}_{100}^{(8)}
\end{bmatrix}.
\]
\( \vec{x} = (A^T A)^{-1} A^T \vec{r} \). In this case, \( \vec{x} = 100 = a_{40} \). The orthogonal projection of the least squares solution onto the subspace spanned by \( A \) (aka our ‘ideal message’) is given by: \( A \vec{x} \).

3. Now find the residual of \( \vec{r} \), denoted \( \vec{y} \), left over by subtracting our ideal message from the received signal: 
\[ \vec{y} = \vec{r} - A \vec{x}. \]

4. Repeat the above steps now using the residual \( \vec{y} \) instead of the received signal \( \vec{r} \). A new correlation between \( \vec{y} \) and all of possible songs would find that the next strongest song is: \( \vec{s}_{100}^{(8)} \). We then update our matrix \( A \) to be 
\[
\begin{bmatrix}
\vec{s}_{40}^{(13)} \\
\vec{s}_{100}^{(8)}
\end{bmatrix}.
\] Finally, we solve \( \vec{x} \) via least squares. For this step, we use the received signal \( \vec{r} \) (not the residual), \( \vec{x} = (A^T A)^{-1} A^T \vec{r} \). We find that \( \vec{x} = \begin{bmatrix} 100 \\ 10 \end{bmatrix} \). We have now recovered both of our message values!

5. We stop iterating when we have gone \( k = 10 \) steps, the known the sparsity of the signal, OR until the norm of the residual is below some threshold value, meaning the residual contains only noise.

### 24.5 OMP Summary

**Setup:**
Let the number of unique codes be \( m \). Each code is length \( n \). We represent each of the \( m \) codes with a vector \( \vec{s}_i \) where \( i \) is an integer between 0 and \( m - 1 \). Each code can potentially carry a message \( a_i \) along with it. Then the measurement at the receiver is
\[ \vec{y} = a_0 \vec{s}_0^{(\tau_0)} + a_1 \vec{s}_1^{(\tau_1)} + ... + a_{m-1} \vec{s}_{m-1}^{(\tau_{m-1})} \]
We will assume that the number of codes that are being broadcast at the same time is very small (for example, 10). If a code is not being broadcast, its coefficient $a_i$ will be zero. Therefore, there are at most $k$ non-zero $a_i$'s.

**Inputs:**
- A set of $m$ codes, each of length $n$: $S = \{\vec{s}_0, \vec{s}_1, \ldots, \vec{s}_{m-1}\}$
- An $n$-dimensional received signal vector: $\vec{r}$
- The sparsity level $k$ of the signal. This is the number of codes with non-zero coefficients.
- Some threshold, $th$. When the norm of the signal is below this value, the signal is assumed to contain only noise.

**Outputs:**
- A set of codes that were identified, $F$, which will contain at most $k$ elements.
- A vector, $\vec{x}$ containing the coefficients of the codes ($a_1$, etc.), which will be of length $k$ or less.
- An $n$-dimensional residual $\vec{y}$

**Procedure:**
- Initialize the following values: $\vec{y} = \vec{r}$, $j = 1$, $A = [\ ]$, $F = \{\emptyset\}$
- while ($(j \leq k)$ & ($\| \vec{r} \| \geq th)$):
  1. Cross correlate $\vec{y}$ with each of the codes. Find the code index, $i$, and the shifted version of the code, $\vec{s}_i(\tau)$, with which the received signal has the highest correlation value.
  2. Add $i$ to the set of code indices, $F$.
  3. Column concatenate matrix $A$ with the correct shifted version of the song: $A = [A \| \vec{s}_i(\tau)]$
  4. Use least squares to obtain the code coefficients: $\vec{x} = (A^T A)^{-1} A^T \vec{r}$
  5. Update the residual value $\vec{y}$ by subtracting: $\vec{y} = \vec{r} - A\vec{x}$
  6. Update the counter: $j = j + 1$