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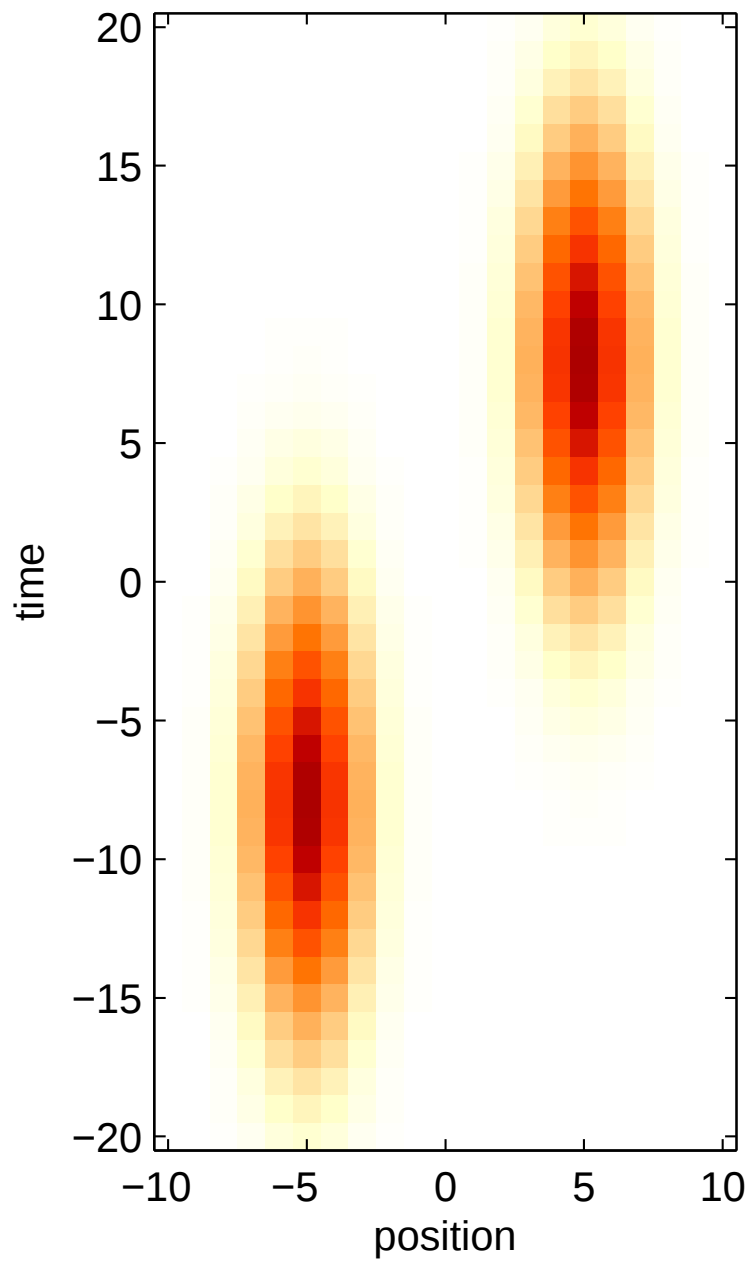
Receptive field example

Shows an example of the kind of receptive fields that we're trying to estimate. Here, we're looking at peri-saccadic receptive fields. A stimulus is flashed right before a saccade. If it's in the LFP's receptive field, there's a corresponding change in the LFP amplitude after the visual lag. However, and very interestingly, if the stimulus is in the spot where the receptive field will be **after** the saccade, there's a later activation: this is the so-called future field. It's not a purely visual signal; it's mostly an attentional thing, as it goes away as soon as you show more than a couple of targets. Our objective: tease out the two (or more) components on every electrode, and characterize them.

```
clf
% Create a "remapped receptive field"
[xi,yi] = meshgrid(-10:10,-20:20);

% early visual component
c1 = exp(-((xi+5).^2/2+(yi+8).^2/32)/2));
% late remapped attentional component
c2 = exp(-((xi-5).^2/2+(yi-8).^2/32)/2));

rf = c1+c2;
imagesc(xi(:),yi(:),rf,[-1,1]);
xlabel('position');
ylabel('time');
axis xy
axis equal;
axis tight
colormap(margecolor);
```



SVD

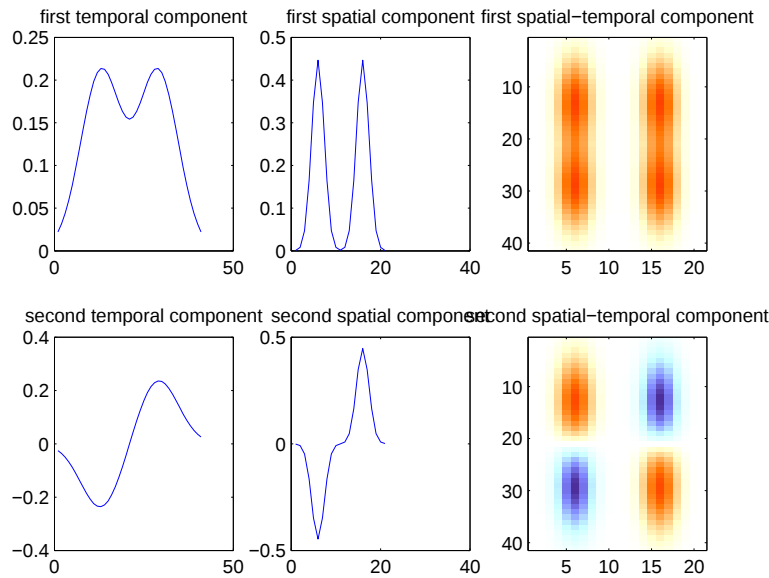
Try and isolate with the SVD Look at the SVD: It's finding a mean and a difference component; we'd like to find two localized components

```

clf
[U,S,V] = svd(rf);

subplot(2,3,1);plot(U(:,1)*sign(mean(U(:,1)))));title('first temporal component');
subplot(2,3,2);plot(V(:,1)*sign(mean(U(:,1)))));title('first spatial component');
subplot(2,3,3);imagesc(S(1,1)*U(:,1)*V(:,1)',.7*[-1,1]);title('first spatial-temporal component');
subplot(2,3,4);plot(U(:,2)*sign(mean(U(:,2)))));title('second temporal component');
subplot(2,3,5);plot(V(:,2)*sign(mean(U(:,2)))));title('second spatial component');
subplot(2,3,6);imagesc(S(2,2)*U(:,2)*V(:,2)',.7*[-1,1]);title('second spatial-temporal component');

```



NMF

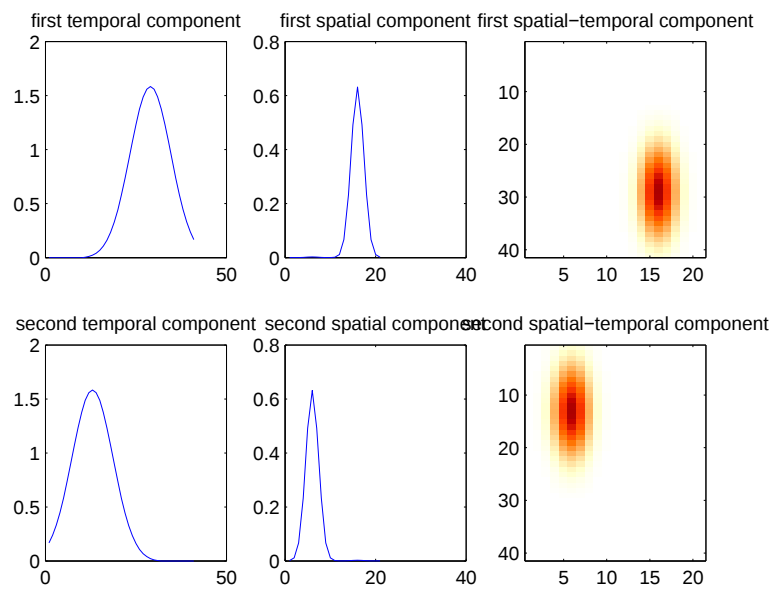
The total RF is low-rank, $U \cdot S \cdot V'$, but $U > 0$ and $V > 0$. Therefore use non-negative matrix factorization

```

[U,V] = mynnmf(rf,2);
V = V';

subplot(2,3,1);plot(U(:,1)*sign(mean(U(:,1)))));title('first temporal component');
subplot(2,3,2);plot(V(:,1)*sign(mean(U(:,1)))));title('first spatial component');
subplot(2,3,3);imagesc(U(:,1)*V(:,1)',[-1,1]);title('first spatial-temporal component');
subplot(2,3,4);plot(U(:,2)*sign(mean(U(:,2)))));title('second temporal component');
subplot(2,3,5);plot(V(:,2)*sign(mean(U(:,2)))));title('second spatial component');
subplot(2,3,6);imagesc(U(:,2)*V(:,2)',[-1,1]);title('second spatial-temporal component');

```



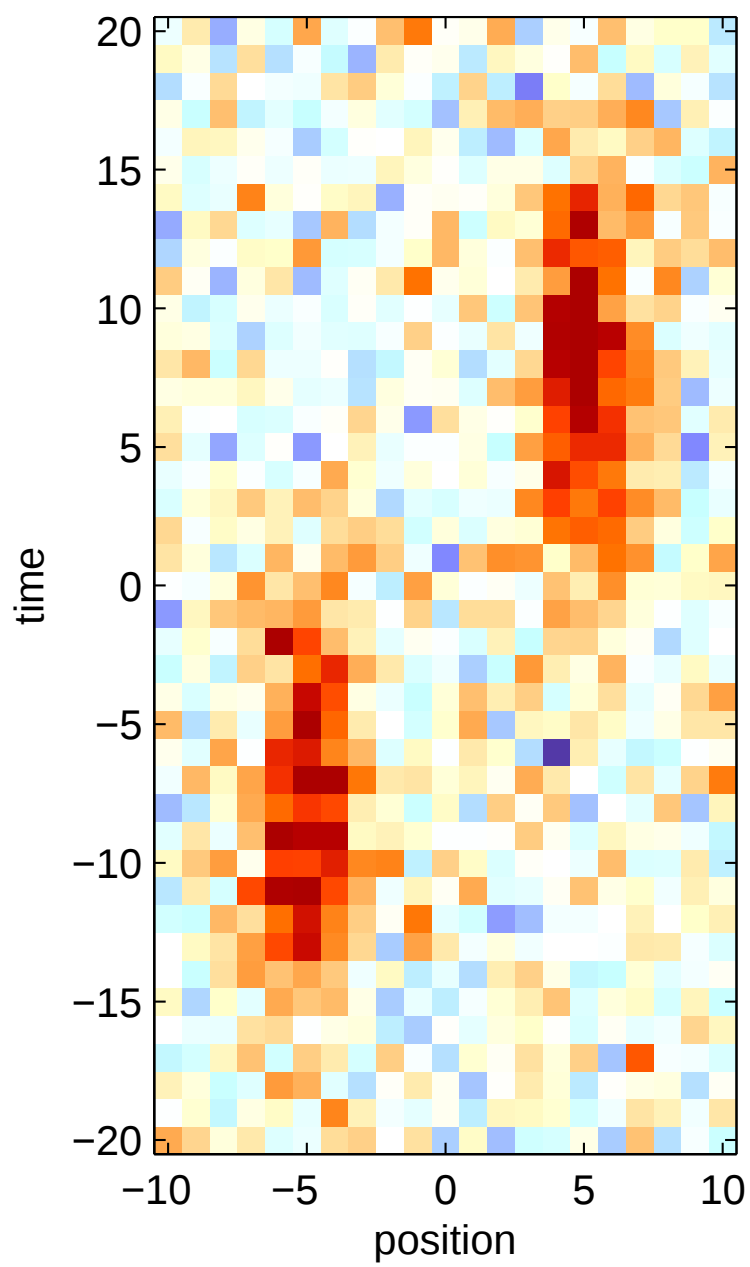
Bingo!

With noise

Try again, this time with noise

```
rf = c1+c2+randn(size(c1)).*0.2;

clf
imagesc(xi(:),yi(:),rf,[-1,1]);
xlabel('position');
ylabel('time');
axis xy
axis equal
axis tight
```

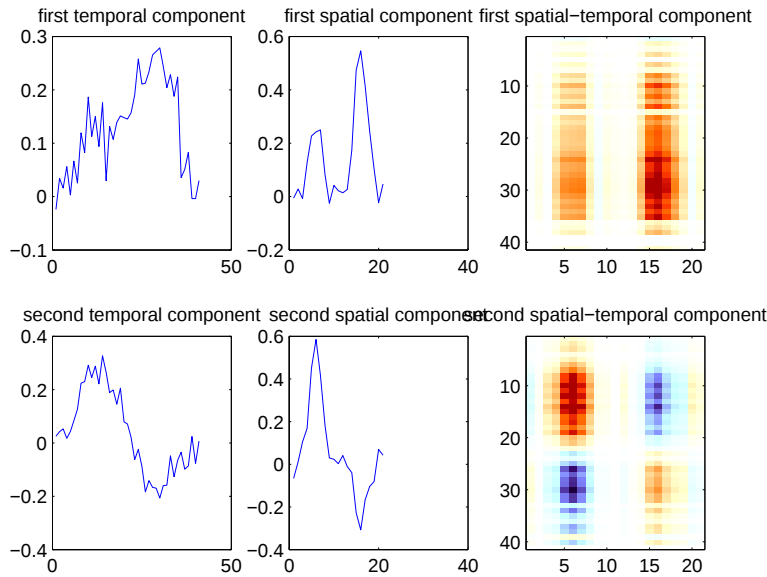


Noise + SVD

Look at the SVD:

```
[U,S,V] = svd(rf);
```

```
subplot(2,3,1);plot(U(:,1)*sign(mean(U(:,1))));title('first temporal component');
subplot(2,3,2);plot(V(:,1)*sign(mean(U(:,1))));title('first spatial component');
subplot(2,3,3);imagesc(S(1,1)*U(:,1)*V(:,1)',.7*[-1,1]);title('first spatial-temporal component');
subplot(2,3,4);plot(U(:,2)*sign(mean(U(:,2))));title('second temporal component');
subplot(2,3,5);plot(V(:,2)*sign(mean(U(:,2))));title('second spatial component');
subplot(2,3,6);imagesc(S(2,2)*U(:,2)*V(:,2)',.7*[-1,1]);title('second spatial-temporal component');
```

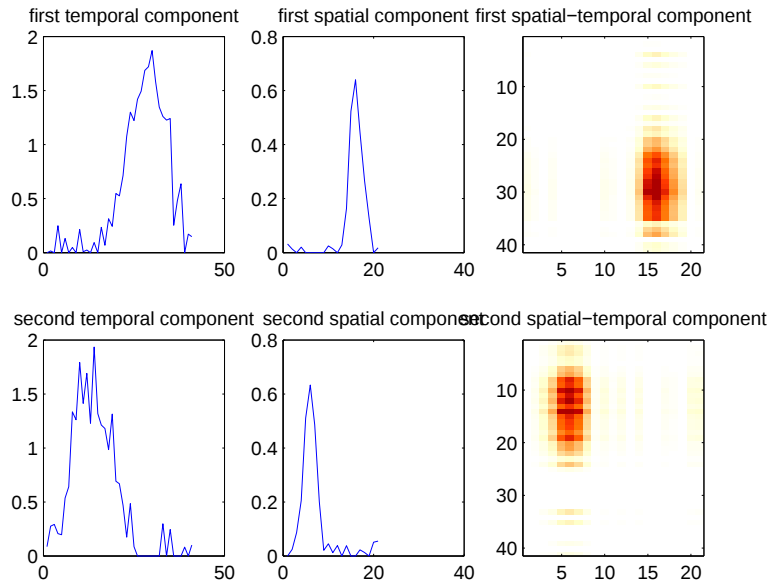


This is not the decomposition we're looking for.

Noise + NMF

```
[U,V] = mynnmf(rf,2,'replicates',5);
V = V';
```

```
subplot(2,3,1);plot(U(:,1)*sign(mean(U(:,1))));title('first temporal component');
subplot(2,3,2);plot(V(:,1)*sign(mean(U(:,1))));title('first spatial component');
subplot(2,3,3);imagesc(U(:,1)*V(:,1)',[-1,1]);title('first spatial-temporal component');
subplot(2,3,4);plot(U(:,2)*sign(mean(U(:,2))));title('second temporal component');
subplot(2,3,5);plot(V(:,2)*sign(mean(U(:,2))));title('second spatial component');
subplot(2,3,6);imagesc(U(:,2)*V(:,2)',[-1,1]);title('second spatial-temporal component');
```



Not half bad, but suffers from lack of regularization on spatiotemporal components

ALD proof-of-concept

Instead of saying that $\text{rf} = U \cdot S \cdot V'$, $U > 0$, $V > 0$, say that $U(:,1), U(:,2), V(:,1), V(:,2)$ are localized. Right now, does this via algo:

1. $\text{res} \leftarrow \text{rf} - U(:,1) \cdot V(:,1)'$
2. $\text{design_mat} \leftarrow \text{blkdiag}(U(:,2), U(:,2), U(:,2), \dots)$
3. $V(:,2) \leftarrow \text{ald}(\text{res}(:,2), \text{design_mat})$
4. $\text{design_mat} \leftarrow \text{blkdiag}(V(:,2), V(:,2), V(:,2), \dots)$
5. $U(:,2) \leftarrow \text{ald}(\text{res}'(:,2), \text{design_mat})$
6. lather, rinse, repeat for other component until convergence

```
niter = 3;
ncomps = size(V,2);

% Start with U,V from NMF
sserror = zeros(niter,1);

for ii = 1:niter
    for kk = 1:ncomps
        res = rf - U*V' + U(:,kk)*V(:,kk)';
        Xmat = [];
```

```

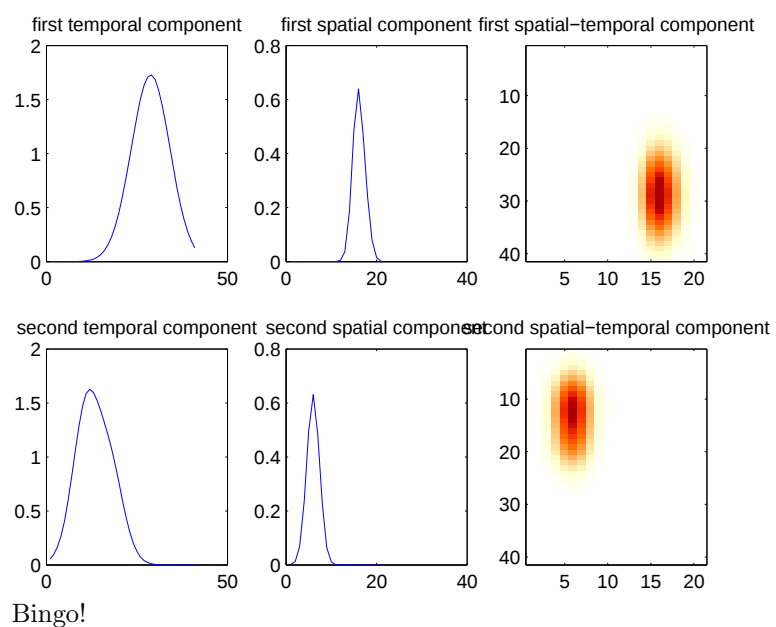
        for nn = 1:size(V,1)
            Xmat = blkdiag(Xmat,U(:,kk));
        end
        %reestimate spatial filters
        y = res(:);
        results = runALD(Xmat,y,size(V,1),1);
        V(:,kk) = results.khatSF;

        Xmat = [];
        for nn = 1:size(U,1)
            Xmat = blkdiag(Xmat,V(:,kk));
        end
        %reestimate spatial filters

        y = res';
        y = y(:);
        results = runALD(Xmat,y,size(U,1),1);
        U(:,kk) = results.khatSF;
    end
    yp = U*V';
    sserror(ii) = .5*sum((rf(:)-yp(:)).^2);
end

subplot(2,3,1);plot(U(:,1)*sign(mean(U(:,1))));title('first temporal component');
subplot(2,3,2);plot(V(:,1)*sign(mean(U(:,1))));title('first spatial component');
subplot(2,3,3);imagesc(U(:,1)*V(:,1)',[-1,1]);title('first spatial-temporal component');
subplot(2,3,4);plot(U(:,2)*sign(mean(U(:,2))));title('second temporal component');
subplot(2,3,5);plot(V(:,2)*sign(mean(U(:,2))));title('second spatial component');
subplot(2,3,6);imagesc(U(:,2)*V(:,2)',[-1,1]);title('second spatial-temporal component');

```



Bingo!