

Finding Neural Codes Using Random Projections

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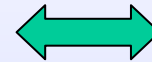
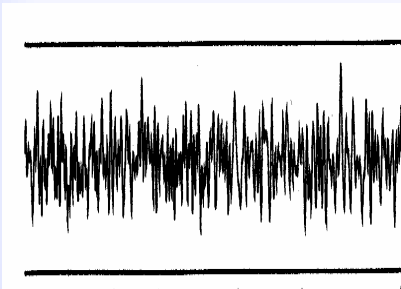
Outline

- **Coding Problem:** find the neural code that maximizes mutual information between input and output
- **Previous Work:** local search methods for nonlinear temporal codes, single channel data
- **A New Approach:** Random Projections
- **Preliminary Experimental Results**

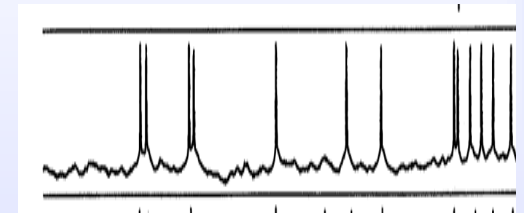
Background: The Coding Problem

- How does neural activity represent information about environmental stimuli?

stimulus
 $X(t)$



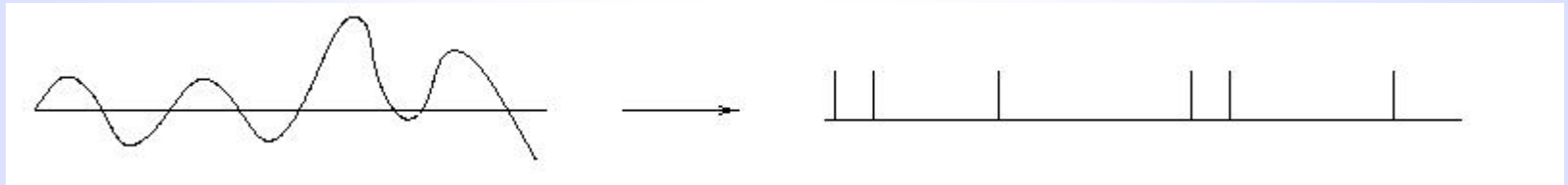
response
 $Y(t)$



Stochastic Framework

Random Variables:

- X stimulus (waveform)
- Y neural response (single channel for now)



stimulus $X=x$

neural response $Y=y$

... (X, Y) forms a joint distribution

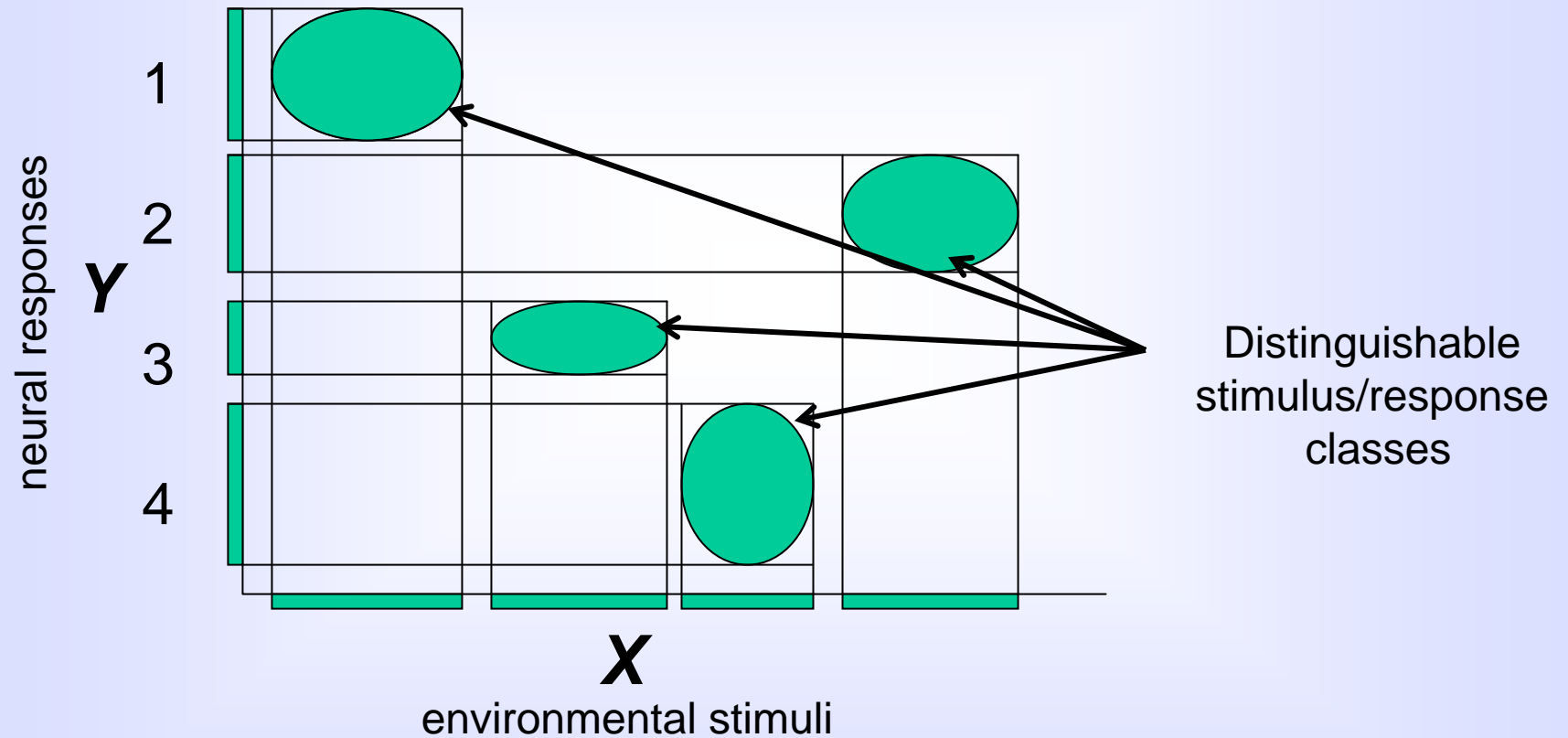
Partitioning the Joint Space

- **Problem:** partition (X, Y) into discrete classes (X_N, Y_M) so as to maximally preserve *mutual information*:

$$I(X_N, Y_M) = \sum_{i,j} \Pr(X_i, Y_j) \log \frac{\Pr(X_i, Y_j)}{\Pr(X_i) \Pr(Y_j)}$$

- NB: this is an *NP-complete* problem
- *Assumption*: there are a small number of distinguishable clusters in the joint space

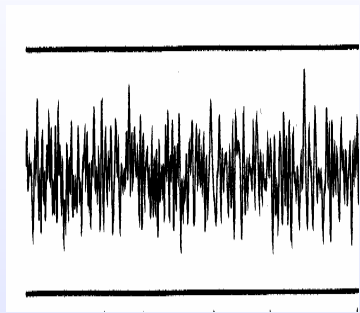
Example Partition



The Curse of Dimensionality

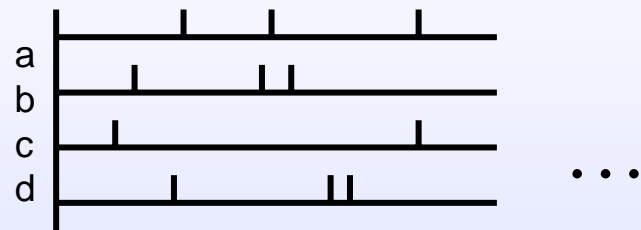
- Both X (input stimuli) and Y (neural response) can be very high dimensional...

X



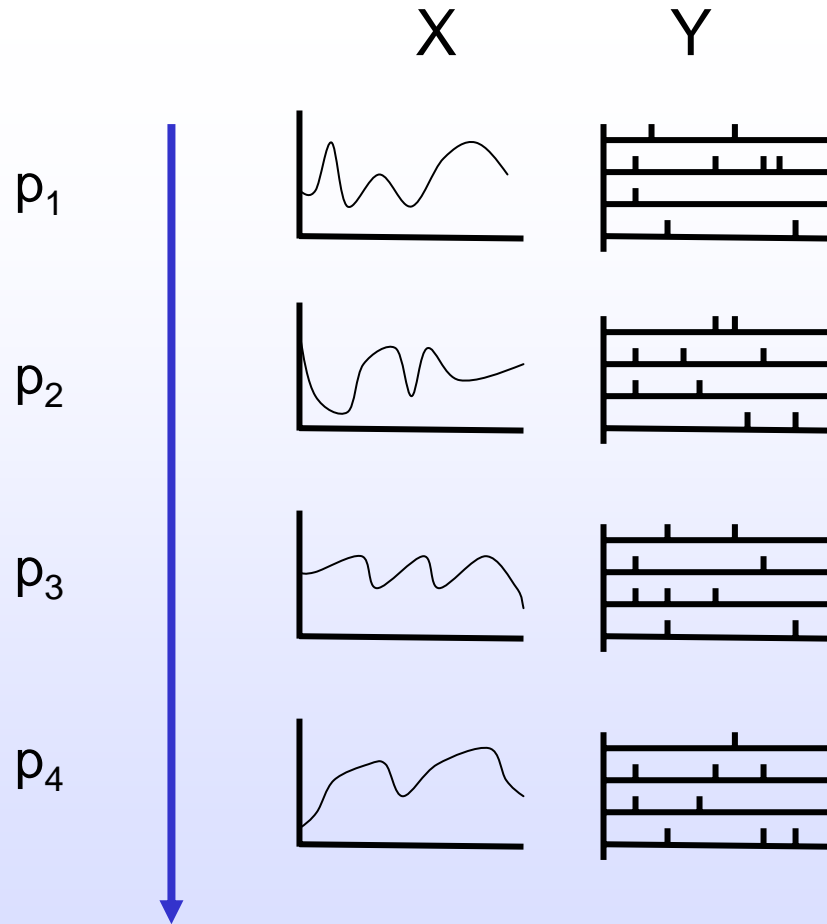
stimulus waveform
represented as a vector
in $R^{>100}$

Y



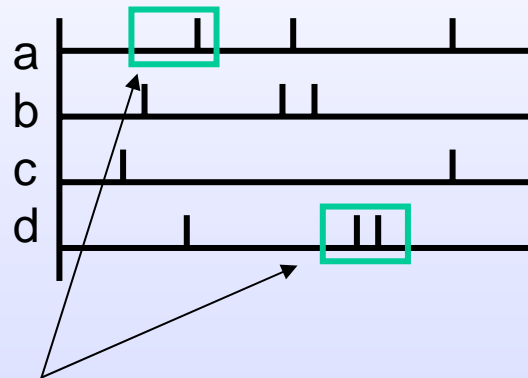
response may be multiplexed
across several channels...
may require several hundred
bits if encoded in binary

Observed Data



Hidden Codes

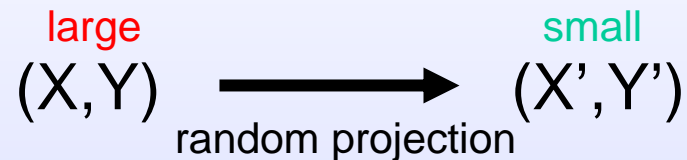
- The optimal partitioning may be based on subtle signals in the data.
- It is difficult to know what features of the data are relevant beforehand...



e.g. perhaps this pattern represents
an important response codeword

Random Projections

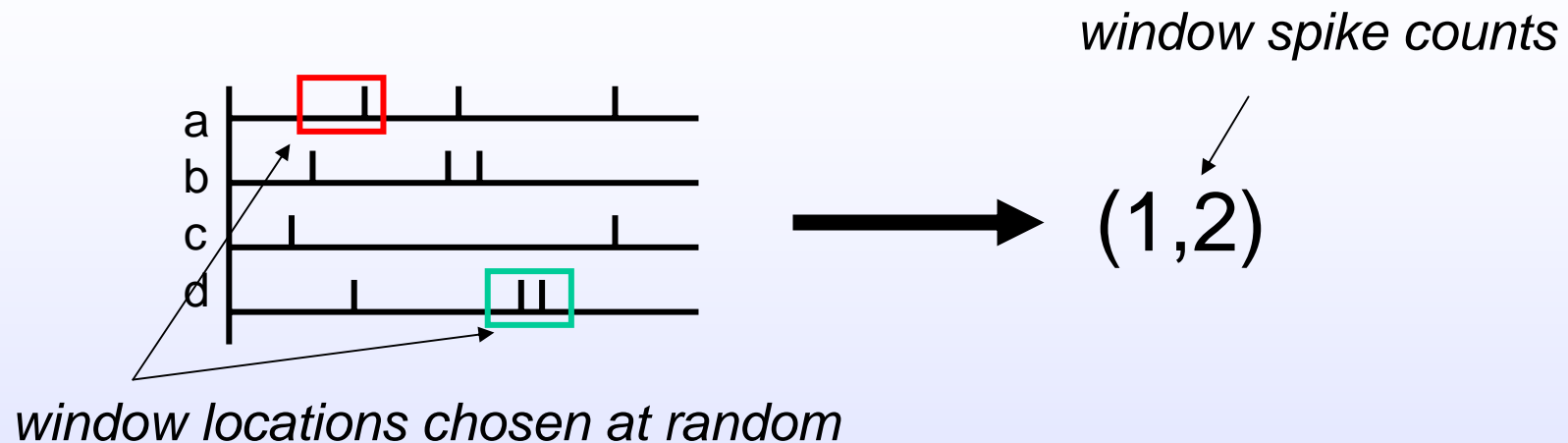
- **Idea:** use random projections to measure “similarity” between observed data points...
(originally proposed by Indyk et al. for computational geometry problems)
- **Outline:** project to a much smaller space; check for patterns in the projected space...



- ...repeat many times

Choosing Projections

- Projections are drawn at random from projection function families:

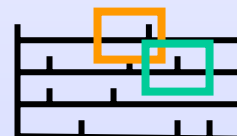
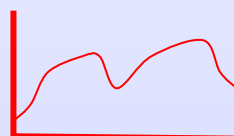
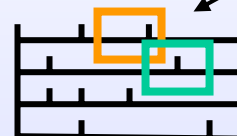
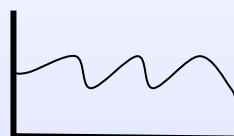
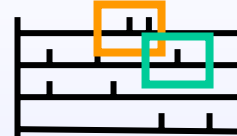
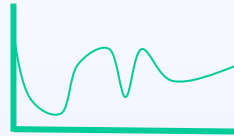
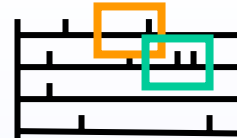
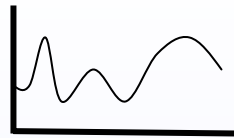


- Less restrictive than pre-imposed metric based methods

Example

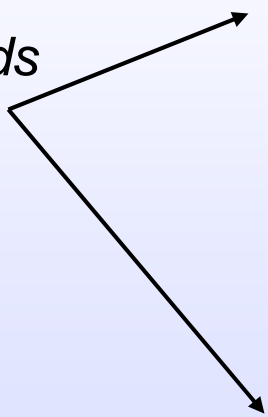
X

Y

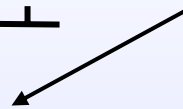


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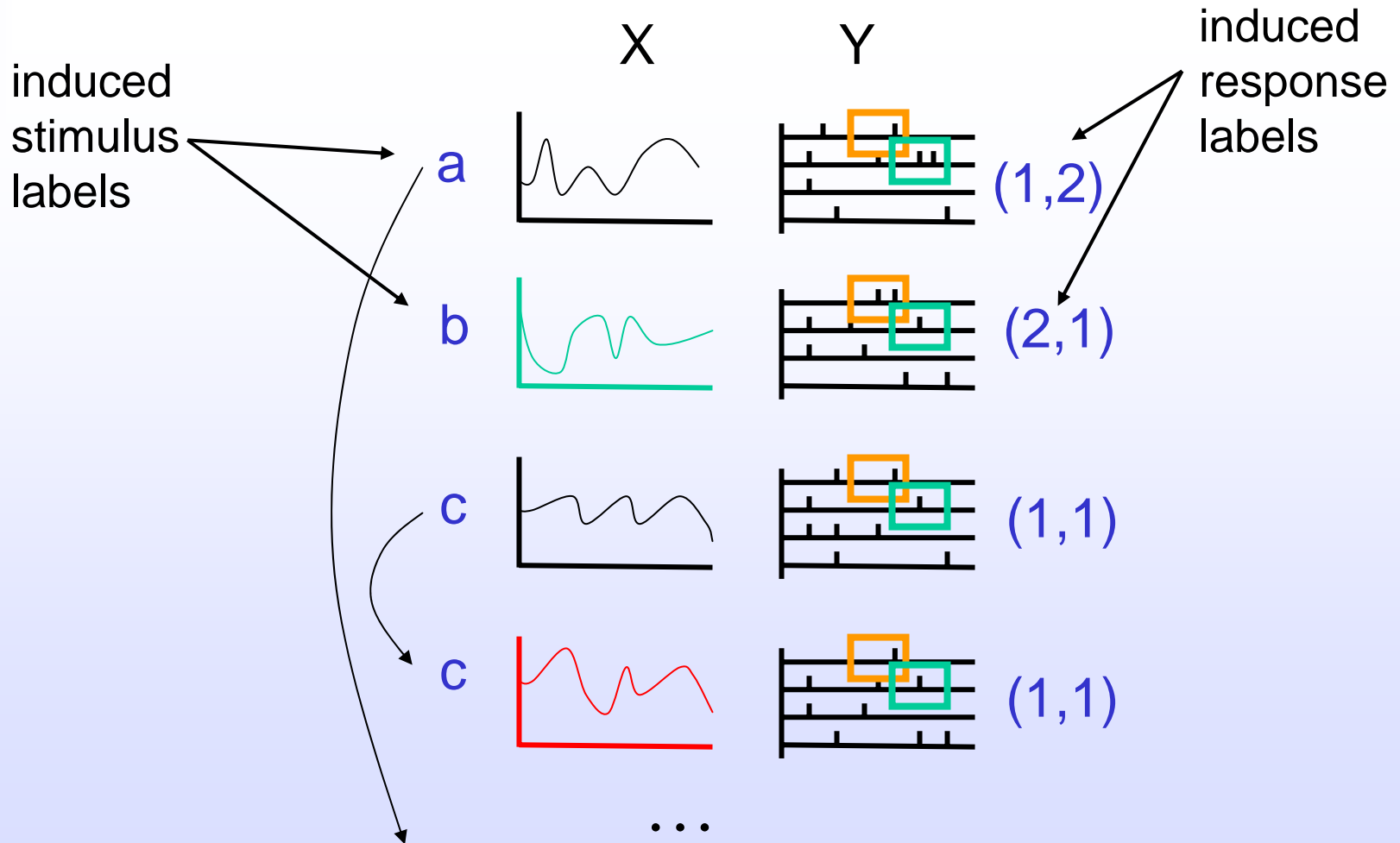
m random
cluster seeds
chosen



random
projection
chosen



Label Assignment



Incorporate Mutual Information

Labeled joint points:

p_1	a	(1,2)	→	Calculate mutual information, decide how <i>interesting</i> this random projection was...
p_2	b	(2,1)		
p_3	c	(1,1)		
p_4	c	(1,1)		

...

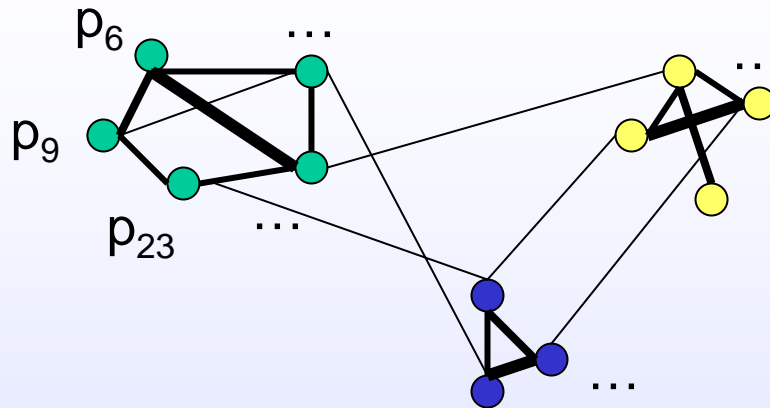
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If *interesting*, this provides some evidence that p_3 and p_4 are in the same joint cluster (distinct from p_1 and p_2)

Idea: repeat this many times...

Joint Similarity Graph

- Construct a weighted graph summarizing the similarity evidence found between each pair of data points over random trials:



- Look for “near-cliques” in this graph - these represent the clusters of the underlying joint distribution and so define the neural code... (we use the *CLICK* software for this)

Some Theoretical Analysis

- **Assumptions:**

- joint space is “clustered” into distinct clusters C_1, \dots, C_k
- cluster C_i sampled with probability p_i , $\sum p_i = c \leq 1$.
- “Good projection family”:

$$x \in C_i, y \in C_j$$

$$i = j \Rightarrow \Pr[\text{proj}(x) = \text{proj}(y)] \geq q$$

$$i \neq j \Rightarrow \Pr[\text{proj}(x) = \text{proj}(y)] \leq r$$

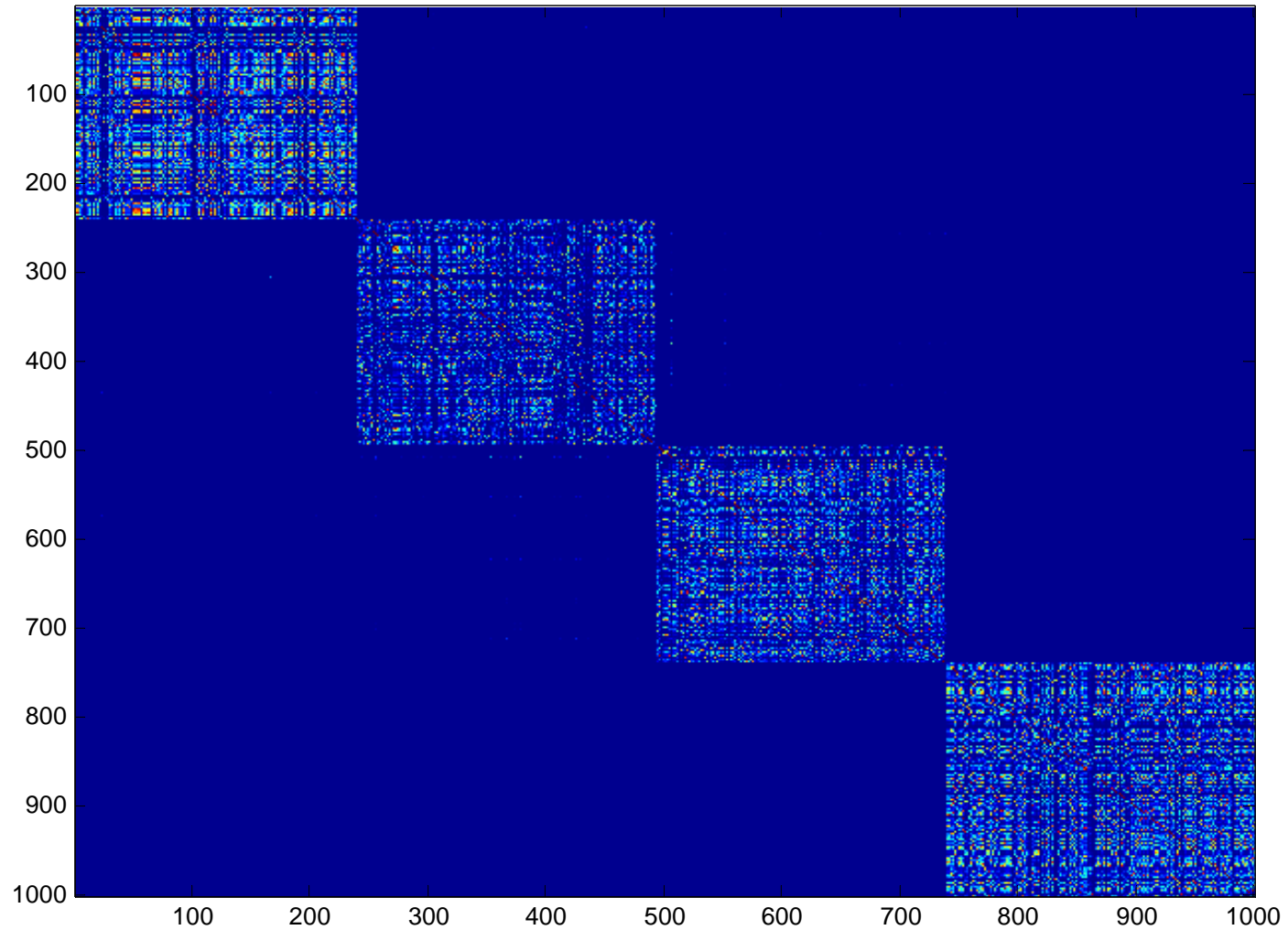
$$q > r$$

- **One Result:** If $q-r = \Omega(1/n)$, can recover true clusters with high probability with $O(n)$ random projections (using some recent random graph clustering results)

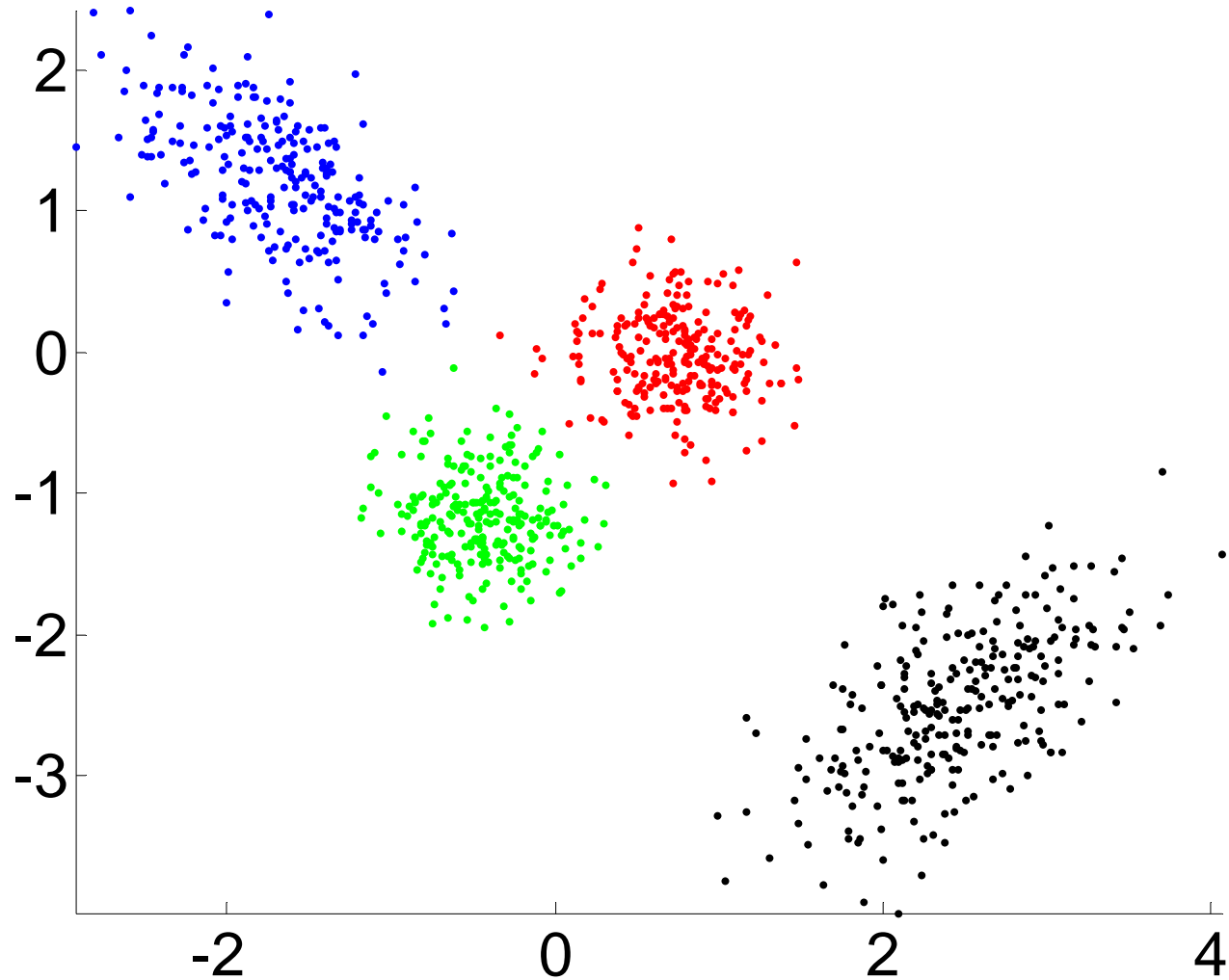
Simple Test Case

- Four Gaussian joint clusters (mixture)
- Recovers clusters perfectly...

Permuted Similarity Matrix



Cluster Assignments



Test case: The cricket cercal system

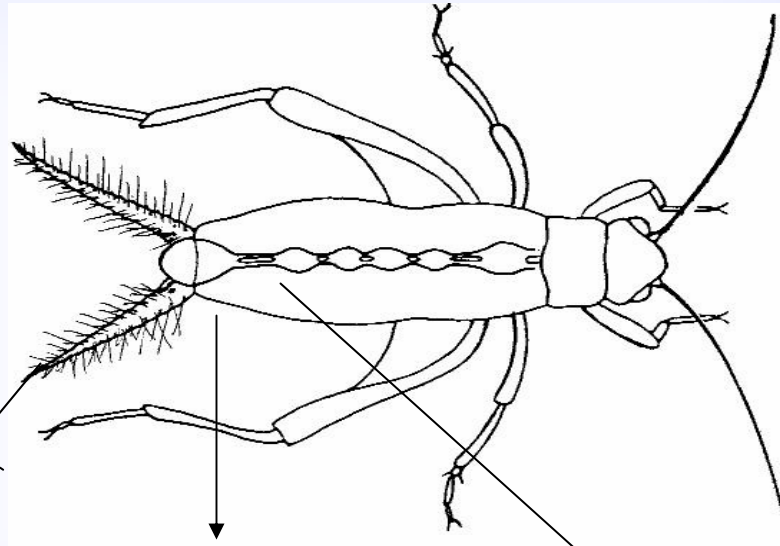
(a low-frequency, near field extension of the auditory system)





Cricket cerci sensory system

Signal from speakers



Terminal
ganglion

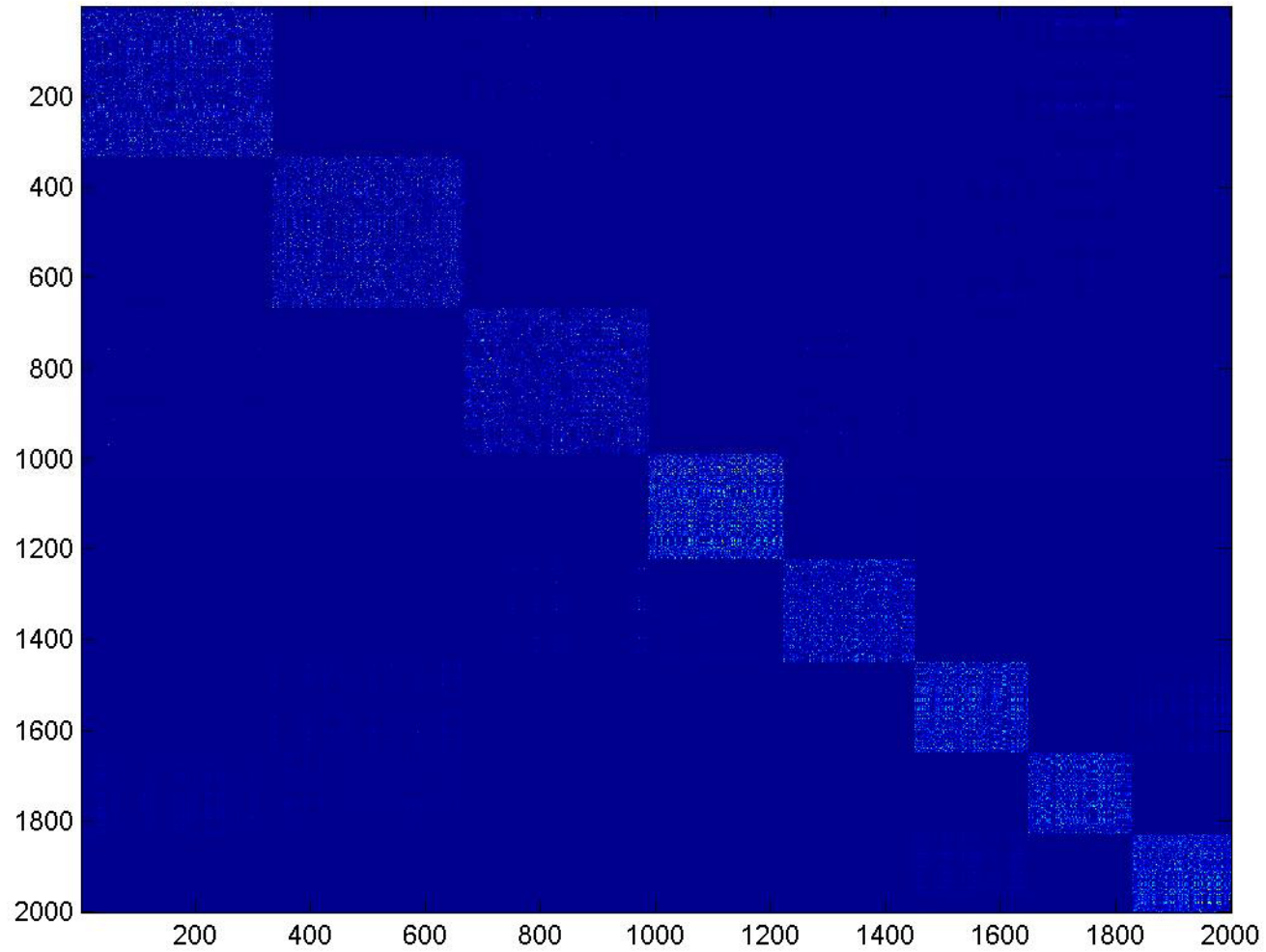
Cerci with
hair

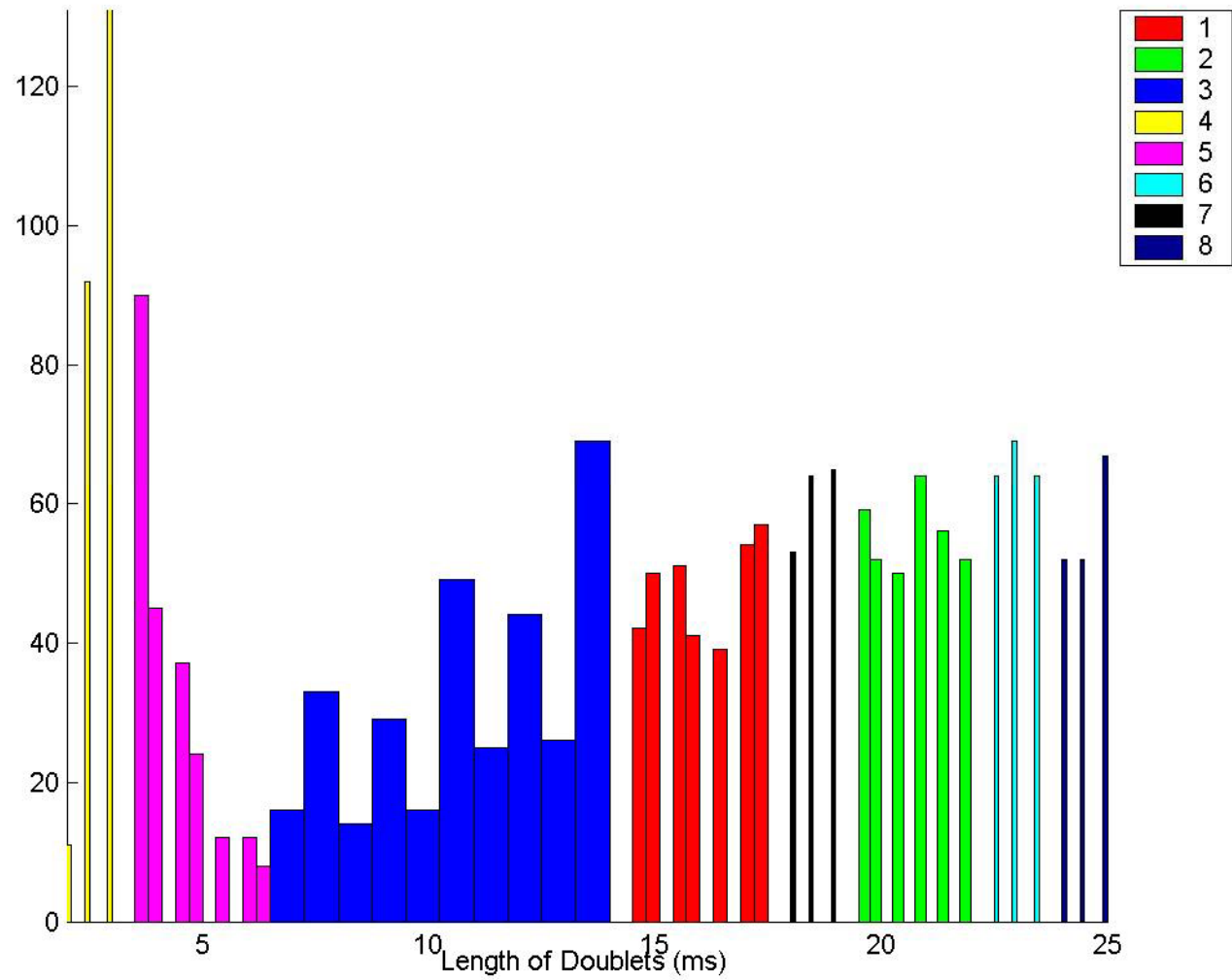
Measurements
of spike trains

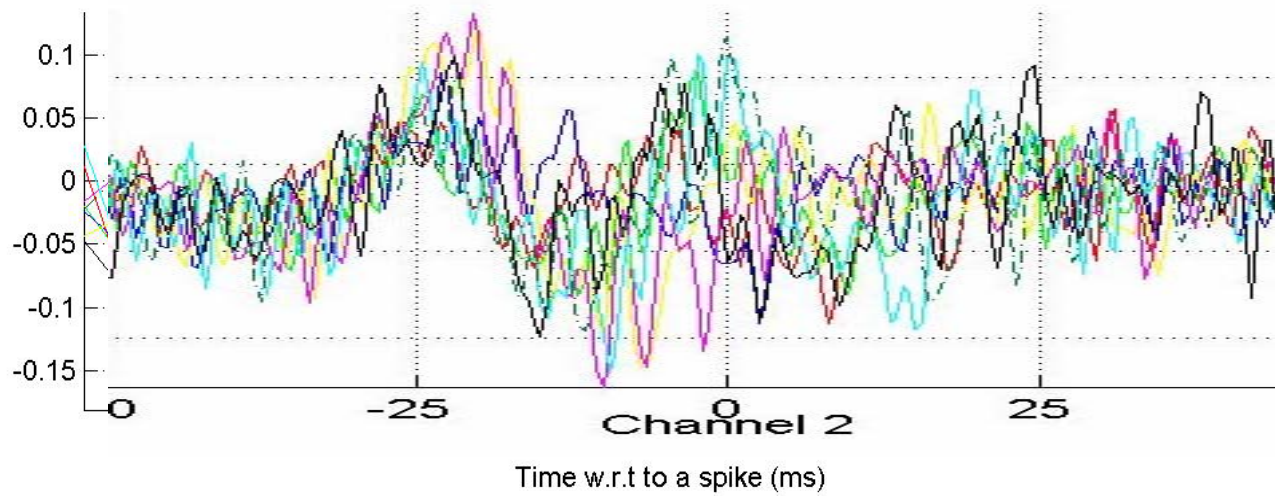
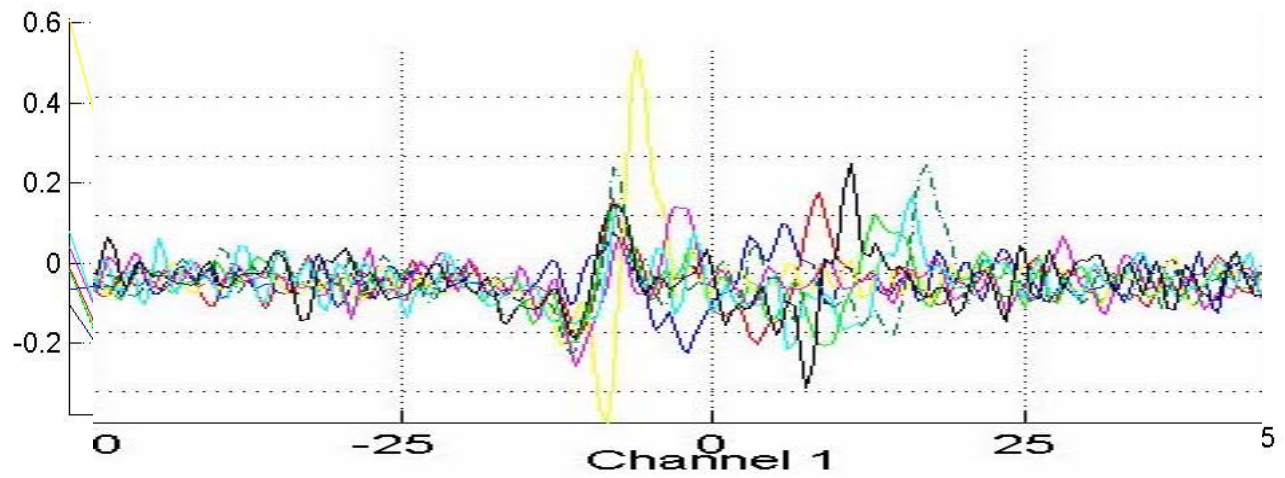
Sample Neural Data Set

- 1000 input/output pairs
- two channel input
- single channel output
- 8 joint clusters found

Data Point Similarity Matrix







Current Work

- Optimizing the class of projection functions used
- Improving bounds on the number of random trials required
- More experiments...
 - multiple channel data
 - synthetic test data