# A Similarity Retrieval System for Multimodal Functional Brain Images



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# **Functional Brain Imaging**



- Study how the brain works
- Imaging while subject performs a task
- Image represents some aspect of the brain e.g.
  - fMRI: brain blood oxygen level
  - **ERP**: scalp electric activity

### Motivation

Given a database of functional brain images from various subjects, cognitive tasks, and image modality.

> Database users need to retrieve similar images

A system that can automatically perform this retrieval will reduce amount of time and effort users spend during this task

### Contributions

- 1. Created a similarity retrieval system for multimodal brain images
  - I. fMRI, ERP, and combined fMRI-ERP
  - II. User interface
- 2. Developed feature extraction methods for fMRI and ERP data
- 3. Developed pair-wise similarity metrics
- 4. Simulated human expert similarity scores

# Outline

- Background
  - ≻ fMRI
  - > ERP

Existing Similarity Retrieval Systems for these modalities

- Feature Extraction Process
- Similarity Metric
- User Interface
- Retrieval Performance
- Simulate Human Expert

# Functional Magnetic Resonance Imaging (fMRI)



- A non-invasive brain imaging technique
- Records blood oxygen level in brain
- > While imaging, subject performs a task

### fMRI Statistical Images



Statistical Analysis

Voxel Thresholding



### **Event-Related Potentials (ERP)**



@ 2004 by Nucleus Communications, Inc.

### A non-invasive brain imaging technique

- Records electric activity along scalp
- > While imaging, subject performs a task

### **ERP Source Localization**

Researchers want to identify the electric activity and its source for each electrode

But, multiple sources for each electrode



LORETA approximates anatomic locations of sources

### Comparison of fMRI and ERP Data

	fMRI	ERP
Spatial resolution	Good (in mm)	undefined/poor
Temporal resolution	Poor (in sec)	Excellent (in msec)

# Similarity Retrieval Systems for fMRI Images

Retain "Most Important" Voxels Whole Brain Similarity Region of Interest Similarity Feature Selection

Our System	Codebook	Wavelet	Bipartite	RV-Coefficient	Correspondence
Yes	No	No	Yes	No	Yes
Yes	Yes	Yes	Yes	Yes	No
Yes	No	No	No	No	Yes
Yes	No	No	No	No	No

# Similarity Retrieval Systems for ERP Images

### No relevant literature found

# Similarity Retrieval Systems for Combined fMRI-ERP Images

### No relevant literature found

## Outline

### Background

# Feature Extraction Process FMRI features

- TIVIRI features
- ERP features
- Similarity Metric
- User Interface
- Retrieval Performance
- Simulate Human Expert

### fMRI Feature Extraction



### **ERP Feature Extraction**



## Outline

- Background
- Feature Extraction Process

# Similarity Metric Summed Minimum Distance Similarity Score for Combined fMRI-ERP Images

- User Interface
- Retrieval Performance
- Simulate Human Expert

### Summed Minimum Distance (SMD) for fMRI and ERP Images

### Subject Q

### Subject T



**Q2T =** 
$$\frac{\sum_{r \in Q} \min_{s \in T} d_E(r, s)}{N_Q}$$

SMD = (Q2T+T2Q) / 2

### Sample SMD Scores

#### similarityRetrievalGUI\_2

1		0.00	
ו ר		11.00	· · · · · ·
2		15.02	
о Л		15.33	
ч 5	HealthyAOD_13	16.29	
0 6		16.30	
7	HealthyAOD 12	10.47	
, 8	HealthyAOD_12 HealthyAOD_4	21 21	
g	HealthyAOD 3	21.31	
10	HealthyAOD 5	21.52	
11	HealthyAODMean con	21.00	
12	HealthyAOD 15	24 43	
13	FaceUnVsEixation 14	25.91	
14	HealthyAOD 7	26.83	
15	HealthvAOD 10	27.44	
16	FaceUpVsFixation 4	27.98	
17	FaceUpVsFixation 9	28.18	
18	FaceUpVsFixation 20	28.21	
19	FaceUpVsFixation 3	28.50	
20	FaceUpVsFixation_19	28.59	
			*
	D	one	

## Similarity Score for Combined fMRI-ERP Images

### $SIM(i,j) = \alpha SMD_{fMRI}(i,j) + (1-\alpha)SMD_{ERP}(i,j)$

### Outline

- Background
- Feature Extraction Process
- Similarity Metric

### User Interface

- Retrieval Performance
- Simulate Human Expert

### **GUI: Front Page**

#### Similarity Retrieval Tool for Multimodal Brain Images

	Choose Modality 💿 fMRI	○ ERP ○ Both	
		1	
IMRI Infeshola	0.01	fMRI Feature Weights	
ERP Threshold	10	Cluster Centroid	< b 1
Scope	⊚ Global ○ ROI	Cluster Area	4
ERP Timeframe	TF1 101 TF2 121	Voxel Mean Distance to Centroid	4
Alpha	< > 0	Voxel Mean Activation Value	4 <b>F</b>
	Upload Database	Variance of Voxel Activation Values	4
Q		Variance of Voxel Distances to Centroid	x >
Query Brain	HealthyAODMean_con		
Query Brain Viewer		Detun Ten 15	
Slices	-26:6:26 Or All Slices		matches
Axial	Coronal Sagittal	Get Matches	

### **GUI:** Retrievals with SMD Scores

#### 🔸 similarityRetrievalGUI\_2

1	HealthyAOD 11	0.00	^
2	HealthyAOD 8	11.82	
3	HealthyAOD_1	15.33	
4	HealthyAOD_13	15.47	
5	HealthyAOD_6	16.38	
6	HealthyAOD_9	16.47	
7	HealthyAOD_12	19.01	
8	HealthyAOD_4	21.31	
9	HealthyAOD_3	21.32	
10	HealthyAOD_5	21.53	
11	HealthyAODMean_con	22.98	
12	HealthyAOD_15	24.43	
13	FaceUpVsFixation_14	25.91	
14	HealthyAOD_7	26.83	
15	HealthyAOD_10	27.44	
16	FaceUpVsFixation_4	27.98	
17	FaceUpVsFixation_9	28.18	
18	FaceUpVsFixation_20	28.21	
19	FaceUpVsFixation_3	28.50	
20	FaceUpVsFixation_19	28.59	
			~
	D.		

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### GUI: Query-Target Activations (fMRI)



#### Target Image



### GUI: Query-Target Activations (ERP)



Charles

Target Image

# Outline

- Background
- Feature Extraction Process
- Similarity Metric
- User Interface
- Retrieval Performance
  - Data Sets
  - FMRI Retrieval Performance
  - ERP Retrieval Performance
  - Combined fMRI-ERP Retrieval Performance
- Simulate Human Expert

### Data Sets for fMRI Retrievals



# Checkerboard -- 12 subjects (Face Recognition)



### Central-Cross -- 24 subjects (Face Recognition)





SB -- 15 subjects (Memorization)



AOD -- 15 subjects (Sound Recognition)

### Data Set for ERP Retrievals









View Human Faces (Face Up) -- 15 subjects View Houses (House Up) -- 15 subjects

# Data Set for Combined fMRI-ERP Retrievals

> ERP: same data set as used in ERP retrieval

- ≻ fMRI:
  - Task: Face recognition using a house up background
  - Same subjects and images as data set for ERP retrieval

### fMRI Retrieval Performance

1. RFX Retrievals

2. Individual Brain Retrieval

3. Testing Group Homogeneity

4. Feature Selection

### fMRI Retrieval Score

Retrieval Score = 
$$\frac{1}{N \times N_{rel}} \left( \sum_{i=1}^{N_{rel}} R_i - \frac{N_{rel}(N_{rel}+1)}{2} \right)$$

Perfect score :Retrieval Score = 0Random score:Retrieval Score ~ 0.5Worst score:Retrieval Score = 1

### fMRI Individual Brain Retrievals

### Use individual brain as query

### Mean Retrieval Scores (Top 6% activated voxels)

Checkerboard SB Central-Cross

AOD

(1000) activated voxels)	
0.09	
0.16	
0.21	
0.26	

### Testing Group Homogeneity for fMRI



### **ERP Retrieval Performance**



### Subject #8 Retrievals

Top Retrievals

Bottom Retrievals

Subject 8	Subject 12	Subject 14	Subject 13	Subject 10
0.00	0.11	0.23	0.25	0.26

Subject 9	Subject 4	Subject 5	Subject 15	Subject 3
0.70	0.75	0.78	0.91	0.95

### **Combined fMRI-ERP Retrieval**



 $SIM(i,j) = \alpha SMD_{fMRI}(i,j) + (1-\alpha)SMD_{ERP}(i,j)$ 

### Outline

- Background
- Feature Extraction Process
- Similarity Metric
- User Interface
- Retrieval Performance
- Simulate Human Expert
  - Simulation Method
  - Data Set
  - Testing Function Performance

### Simulate Human Expert

### Current retrieval system requires some expert knowledge

	Centroid Only	Centroid and	Average Activation
		Average Activation	Value Only
		Value Only	
Correlation	0.60	0.64	0.52
Coefficients			

Estimate a function to generate similarity scores with high correlation to expert scores

### Simulation Method

- 1. Uniform feature representation: create codebook and encode each subject
- 2. Concatenate the codebook features for each pair of subjects
- 3. Create eigenfeatures
- 4. Estimate a function
- 5. Test function performance

### 1. Uniform Feature Representation



### 2. Concatenate Codebook Features



### 3. Create Eigenfeatures



Use PCA to project each feature onto N principal components. Here N=1.



### 4. Estimate a Function



Linear function using linear regression

Non-linear function using generalized regression neural networks (GRNN)

### 5. Test Function Performance

The Pearson Correlation Coefficient (CC)

$$\frac{\sum_{i=1}^{n} \left(\widehat{y}_{i} - \mu_{\widehat{y}}\right) \left(y_{i} - \mu_{y}\right)}{\left(n-1\right) s_{\widehat{y}} s_{y}}$$

The Average Absolute Error (A-ABSE)

$$\sum_{i=1}^{n} \left[ y_i - \hat{y}_i \right]$$

n

The Root Mean Square Error (RMSE)

$$\sqrt{\frac{\sum_{i=1}^{n} \left[y_i - \widehat{y}_i\right]^2}{n}}$$

### Data Set



fMRI data (Central-Cross)

- -- 23 subjects
- -- Face Recognition task

Human Expert Generated Pairwise Similarity Matrix

### **Overall Function Performance**

		Original Codebook Features		Eigenfeatures	
		Linear	Non-Linear	Linear	Non-Linear
		Function	Function	Function	Function
Training	A-ABSE	1.82	0	2.11	0.58
	RMSE	2.25	0	2.57	0.82
	CC	0.52	1	0.35	0.96
Testing	A-ABSE	2.26	1.74	2.18	1.36
	RMSE	2.83	2.32	2.67	1.77
	CC	0.23	0.59	0.25	0.76
	-				

### **Feature Selection**



### Contributions

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### **Future Direction**

- > Add more modalities to the system
- Obtain more expert scores for function estimation
- Obtain more data

Develop similarity metrics other than SMD

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