Brain Computer Interface for Communication and Control

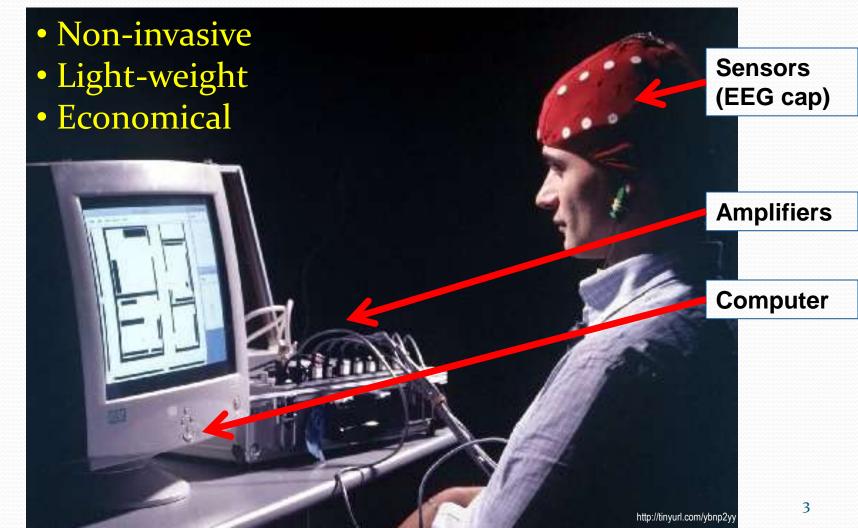


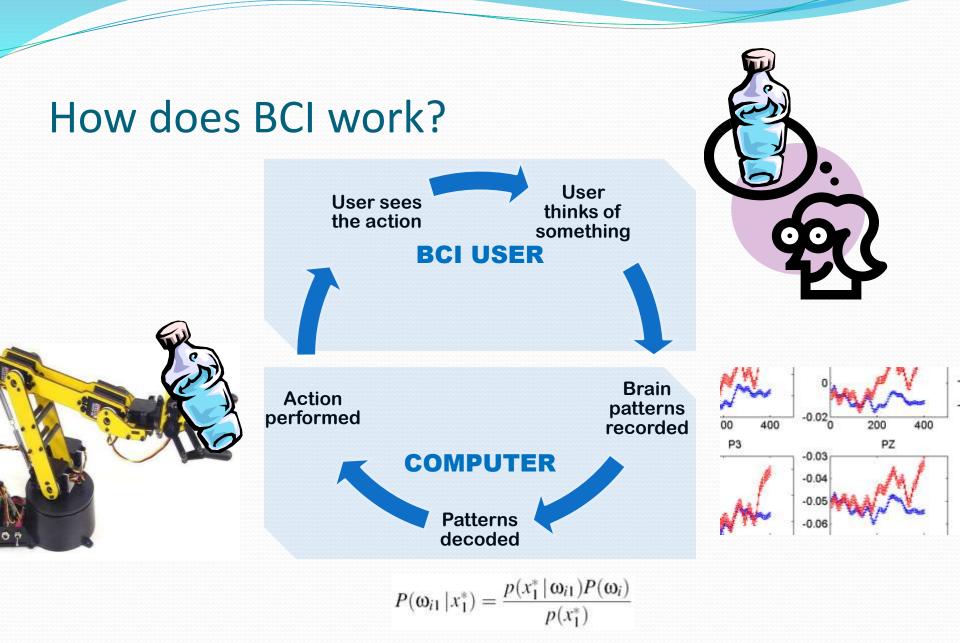
Po T Wang Nenadić Lab Oct 3, 2009

What is BCI and what can it do?

- Brain computer interface (BCI) is a direct communication pathway between human brain and a computer device.
- Often used for restoring lost bodily functions
 - Amyotrophic lateral sclerosis (ALS) patients
 - Spinal cord injury (SCI) patients
- Can help
 - Communicate with people
 - Control assistive/prosthetic devices

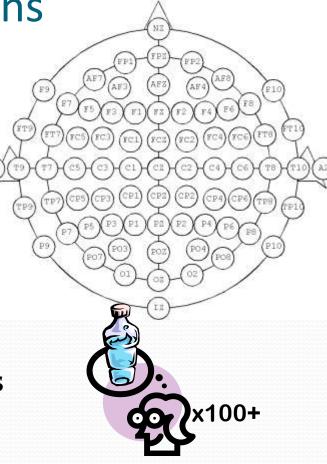
Electroencephalography (EEG)-based BCI





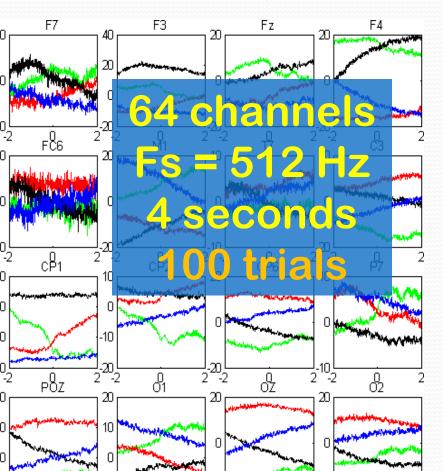
Discovering new useful patterns

- BCI relies on consistent mental patterns to work
- Where are the useful patterns, in which sensors, at what time?
 - Record all sensors for as long as feasible.
 - Repeat the mental tasks many times
 - Data can be over 100,000 dimensions per trial for just a few seconds of recording



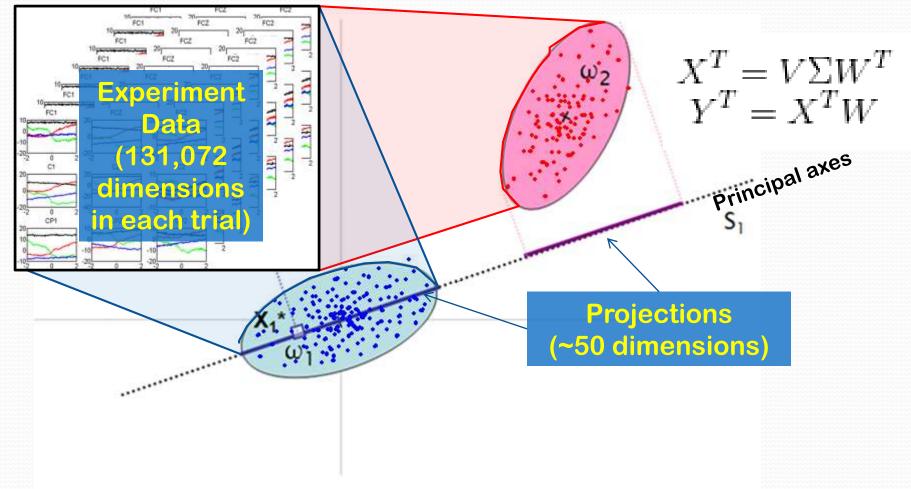
Problem of high dimensionality

• Recorded signals have extremely high dimensions



	J					
	through 7					
0.0028	-0.0079	-0.0392	-0.0377	-0.0514	0.0623	0.0363
0.0036	-0.0074	-0.0383	-0.0370	-0.0507	0.0637	0.0370
0.0046	-0.0062	-0.0363	-0.0359	-0.0495	0.0655	0.0378
0.0049	-0.0055	-0.0348	-0.0354	-0.0488	0.0664	0.0381
0.0035	-0.0067	-0.0352	-0.0364	-0.0499	0.0651	0.0370
0.0014	-0.0089	-0.0360	-0.0381	-0.0515	0.0631	0.0351
-0.0001	-0.0104	-0.0381	-0.0394	-0.0525	0.0620	0.0338
-0.000 <mark>6</mark>	-0.0108	-0.0385	-0.0398	-0.0522	0.0623	0.0334
0.0001	-0.0101	-0.0300	-0.0393	-0.0511	0.0634	0.0336
0.0016	-0.0087	A-0 (259)	-0.03		0.0649	0.0344
0.0031	-0.0072	-0-02 6	-0.0/15	-//0_89	0.0658	0.0346
0.0044	-0.0059	-0,03 3	-0.0.71	- 1.047	0.0662	0.0344
0.0051	-0.0052	-0 /1		b. 0 🦛 👘	0.0662	0.0345
0.0052	-0.0053	-0.0328	-0.0372	-0.0493	0.0656	0.0339
0.0047	-0.0065	-0.0338	-0.0378	-0.0505	0.0644	0.0336
0.0040	-0 0 79	-0.0350	-0.0303	-0.0517	0.06 9	0.0333
0.0039	C POPC				16 2	0.0330
0.0047	-0 DI 71		-1 0	0.50	6 8	0.0338
0.0051				4 7	6 8	0.034
0.0044	-0.0065	-0.0367	-0.0341	-0.0491	0.0645	0.0354
0.0024	-0.0079	-0.0363	-0.0354	-0.0498	0,0643	0.0350
0.0001	-0.0102	-0.0367	- 0374	-0.0513	0.0636	0.034
0.0010	-0.0127	18 00 77		12007	0.0620	0.0330
0.0030	-0.0144		-1 0 99	-0.22.7	0.0623	0.033
0.0037	-0.0152	0-0-290	259		0.0618	0.0330
0.0037	-0.0140	-0.0390	-0.0404	-0.0542	0.0616	0.033
0.0029	-0.0135	-0.0384	-0.0399	-0.0544	0.0621	0.033
0.0016	-0.011	0.020	- (0390	-0.05 5	0,0633	0.032
0.0004	-0.00.01	-0): 50		0.0	0.0652	0.032
0.0023	-0.0075	0 0 11	-1 0 57		0.0673	0.032
0.0040	-0.0062				0,0692	0.031
0.0047	-0.0054	-0.0304	-0.0337	-0.0494	0.0699	0.031
0.0040	-0.0050	-0.0305	-0.0340	-0.0506	0.0692	0.030
0.0026	-0.0069	-0.0315	-0.0369	-0.0524	0.0675	0.027
0.0016	-0.0078	-0.0325	-0.0390	-0.0536	0.0656	0.024
0.0015	-0.0077	-0.0332	-0.0405	-0.0537	0.0646	0.021
0.0028	-0.0063	-0.0331	-0.0411	-0.0524	0.0648	0.019
0.0046	-0.0040	-0.0324	-0.0408	-0.0504	0.0659	0.018
0.0064	-0.0016	-0.0313	-0.0400	-0.0479	0.0675	0.018
0.0070	0.0004	-0.0301	-0.0390	-0.0450	0.0690	0.010
0.0081	0.0012	-0.0292	-0.0386	-0.0447	0.0697	0.019
0.0078	0.0012	-0.0292	-0.0388	-0.0447	0.0697	0.019
0.0078	0.0013	-0.0283	-0.0388	-0.0443	0.0693	0.019
0.0073	0.0017	-0.0200	-0.0395	-0.0448	0.0693	0.019
0.0062	0.0018	-0.0279	-0.0417	-0.0453	0.0683	0.020
0.0061 0.0070	0.0022	-0.0279 -0.0272	-0.0422 -0.0417	-0.0455 -0.0448	0.0682	0.020

Classwise principal component analysis (cPCA)



Discriminant analysis

- Finds combinations of features that best separate the classes
- Brings down to 1 3 dimensions

Sample covariance of class means: $\Sigma_b = \frac{1}{C} \sum_{i=1}^{C} (\mu_i - \mu) (\mu_i - \mu)^T$

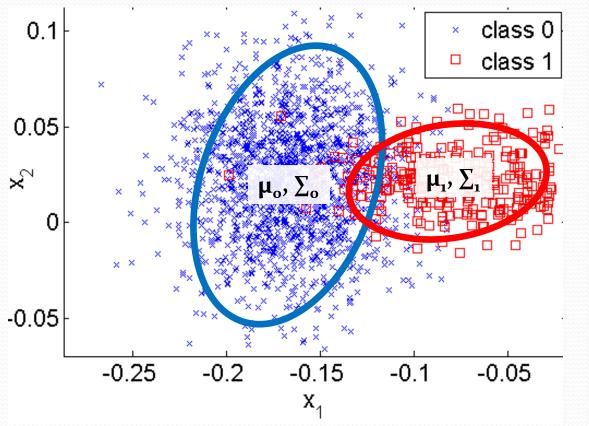
Class separation:

$$S = \frac{\vec{w}^T \Sigma_b \vec{w}}{\vec{w}^T \Sigma \vec{w}}$$

Optimal separation:

 $(\Sigma^{-1}\Sigma_b)\vec{\omega} = \Lambda \vec{\omega}$

Classification: Where do new data belong?



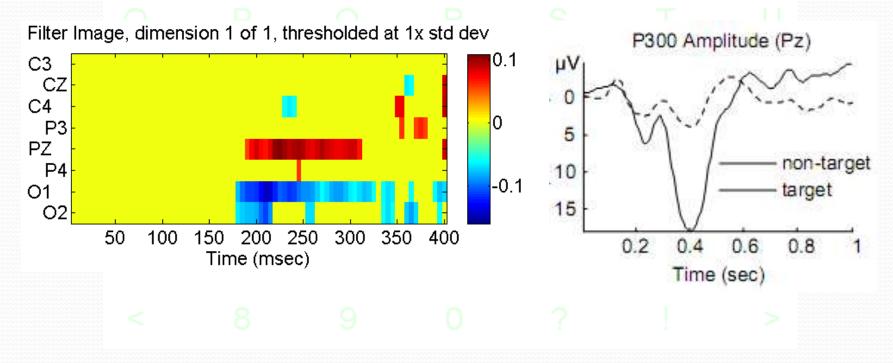
$$p(\omega_i|D) = \frac{p(D|\omega_i)p(\omega_i)}{p(D)}$$
$$= \frac{N(\mu_i, \Sigma_i, D)p(\omega_i)}{\sum_i N(\mu_i, \Sigma_i, D)p(\omega_i)}$$

Class decision = $max_i(p(\omega_i|D))$

Nenadic Lab, unpublished work (2009)

P300 spelling system

 P300 is a positive-deflecting brain wave pattern occurring 300 ms after an oddball object is presented



Sellers & Donchin, Clinical Neurophysiology 117 (2006) 538-548

A B C D E F G P300 spelling system

- Spelling software flashes groups of alphabets and waits for the P300 signal
- Each P300 signal narrows down the choices of alphabets, until one is chosen

P300 speller 2009-06-21

ONLINE STAGE

tosterior=0.99647, thebomb=1 fou entered: E likelyLetters = 1.TES Testerior=0.99676, thebomb=1 You entered: . Glapsed time is 291.113005 seconds. >> p300spelier1((1:0) (1) TrainDB1) RIGEAC APT Library HEDEV.DLL loaded DR Connected to BIOPAC HELSO device. Righest Letter: 7 source=0.004675 mean=0.0135 std=0.0335

N.

С

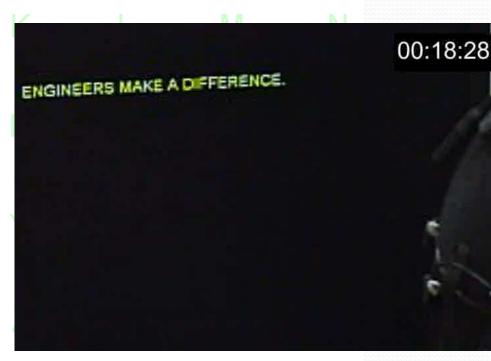
н



00:19:23

A B C D E F G P300 spelling is easy to use

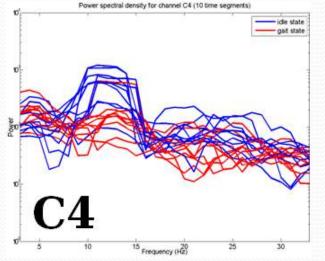
- P300 signal is natural: 97% of population^[1] responds to the oddball.
- Users in our lab can achieve 80-100% accuracy with less than 15 minutes training.



[1] C Guger, et al. Neuroscience Letters 462 (2009) 94-98

Future works

- Computer mouse system
- Frequency based decoding
- Combining multiple events
- Continuous trajectory decoding



	i	4.0%(2.0%)	1.3%(1.2%)	91.1%(0.0%)	4.0%(3.4%)
	l	11.6%(4.6%)	15.3%(4.2%)	2.0%(0.0%)	63.7%(4.1%)
ž	·+- 	aida(3) line,	81.0%(1.7%)	(#1)	
	I	85.8%(5.7%)	1.3%(1.2%)	5.9%(0.0%)	16.8%(2.0%)
	Ι	0.0%(0.0%)	82.7%(2.3%)	1.0%(0.0%)	15.8%(2.0%)
	Ι	4.0%(2.0%)	1.3%(1.2%)	91.4%(1.1%)	3.3%(2.3%)
	I	10.2%(4.1%)	14.7%(3.1%)	1.7%(1.1%)	64.0%(5.0%)
ž-	+-				
	I	aida(2) line,	79.9%(2.6%)	(#4)	
	I	83.8%(1.1%)	0.3%(1.2%)	5.6%(1.1%)	15.8%(3.4%)
	Ι	1.7%(1.1%)	81.3%(4.6%)	0.0%(0.0%)	18.2%(2.3%)
	da.	F 00 / 0 00 V	4 190 / 0 005	00 40 / 4 405	4 00 / 0 00 5

Acknowledgements

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