# **CONVOLUTIONAL NEURAL NETWORK APPROACH FOR EEG-BASED EMOTION RECOGNITION USING BRAIN CONNECTIVITY AND ITS SPATIAL INFORMATION**





## **Conventional automatic emotion recognition using EEG (electroencephalography)**



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- Only signals from a single electrode are considered
- Spatial arrangement of EEG electrodes is not considered



#### Brain connectivity of EEG signals

For two EEG signals x and y captured from different brain regions,

- PCC (Pearson correlation coefficient):
- PLV (phase locking value):
- PLI (phase lag index):

$PCC - \frac{1}{2}$	cov(x,y)
100 -	$\sigma_{\chi}\sigma_{\mathcal{Y}}$
$PLV = \frac{1}{\Lambda}$	$\frac{1}{\sqrt{2}} \sum_{n=1}^{N}$
$PLI = \frac{1}{N}$	$\sum_{n=1}^{N}$

### **Spatial information of EEG signals**

2-D input to consider spatial information through CNN (convolutional neural network)

Electrode number 2 3 ... n-1 n Η  $\sim$  $\mathbf{M}$ 

Electrode ordering methods considering hemispheric asymmetry







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We propose a new CNN-based approach for emoti considering brain connectivity and spatial informa Cross-regional relationship and spatial information pattern between hemispheres) of EEG signals impl performance

#### **CNN structures** CNN-2 $\rightarrow$ MaxPool $\rightarrow$ FC Conv CNN-5 $\rightarrow$ MaxPool $\rightarrow$ FC Conv MaxPool Conv -→ Conv CNN-10 → MaxPool Conv MaxPool MaxPool Conv Conv

### Experiment

- Classification problem low valence (negative emotion) vs. high valence (positive emotion) EEG signals from DEAP database
- Preprocessing
  - temporal segmentation (3 seconds with 2.5 seconds overlap)
  - bandpass filtering (delta, theta, low alpha, high alpha, alpha, low beta, mid beta, high beta, beta, and gamma)
- Comparison
  - PSD (power spectral density) features: max. 80.86% with CNN-5
  - SVM (support vector machine): 55.42% with PSD features

	Ordering method	CNN-2	CNN-5	CNN-10
PCC	random	93.82	94.44	91.48
	dist1	93.80	94.17	92.18
	dist2	93.57	94.30	92.68
PLV	random	96.50	97.13	89.93
	dist1	96.62	97.11	91.30
	dist2	98.58	99.72	90.92
PLI	random	85.00	74.52	59.29
	dist1	85.03	78.17	61.48
	dist2	84.98	77.48	61.24

Classification accuracies [%]

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prove emotion recognition



