

FACULTÉ DES SCIENCES Département d'informatique



Estimating emotions and tracking interest during movie watching, based on multimedia content and physiological responses

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- Introduction
- Dataset
- Affective model
- Multimedia and physiological signals correlation
- Affective characterization
- Classification for interest detection
- Conclusion and future work



- Affect and interest detection from movie scenes is useful for indexing, highlighting and user profiling.
- Find correlates features
- Assessment of user behaviour and emotions
 - User's emotion prediction by arousal/valence levels from audio-video content analysis
 - User's emotional characterization by arousal/valence levels from physiological signals
- Assess interest levels

Video dataset



- In the experimental data set we have 8 movies from 4 different genres:
 - Horror: The Ring, 28 days later
 - Action: Kill Bill VOL I, Saving private Ryan
 - Drama: Hotel Rwanda, The Pianist
 - Comedy: Mr. Bean's Holiday, Love actually
- 8 short video clips were extracted from these movies.
- Neutral clip between each two video clips to record baseline and let the participant to return to neutral state.
- Total duration ~2 hours.

| | Neutral clip | Movie scene | Self assessment | | |
|--------------|--------------|-------------|-----------------|-------------|---|
| | ~30 seconds | 1~2 minutes | time | → | |
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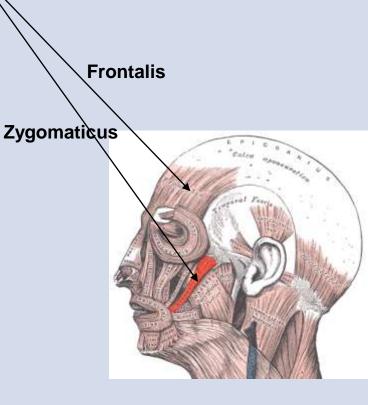


- All video content features are extracted with the help of OVAL and OMT software packages (*Viper*)
- Features extracted from multimedia content, namely:
 - Average shot duration
 - Shot change rate and variation
 - Color variance
 - Key lighting
 - Zero crossing rate (ZCR)
 - Audio energy
 - Audio type vector (music, speech, environment sound, and silence ratio over time)

Physiological feature extraction



| EMG | Contraction Power | |
|-------------|------------------------|-----|
| ECG | Heart Rate |] ` |
| | Heart Rate Variability | |
| | Inter-Beat- Interval | |
| Blinks | Blink Amplitude | |
| | Blink Duration | |
| | Blink Frequency |] |
| Plethysmo- | Heart Rate | |
| graph | Heart Rate Variability | |
| 9 | Inter-Beat- Interval | |
| Respiration | Respiration Depth | |
| | Respiration Rate | |
| GSR | GSR | |
| | Number of peaks | |
| | Peak amplitude | 1 |
| | dGSR/dt |] |
| Temperature | Temperature | |
| | dTemp/dt | |

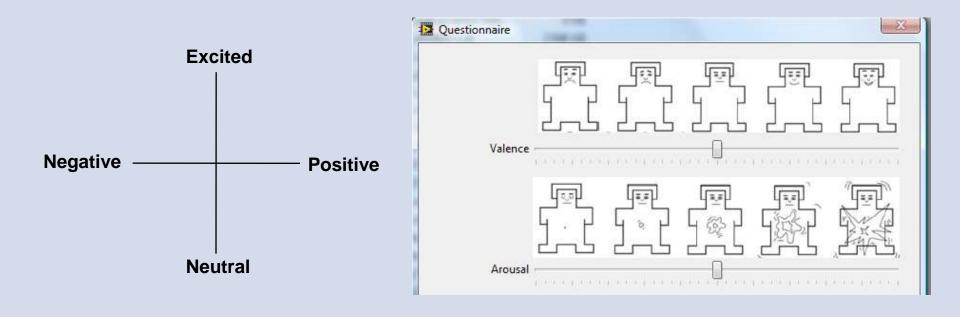


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Self Assessment



• Introducing the arousal valence space



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Feature correlation



| C | Correlation of selected multimedia features with physiological features of 8 participants | | | |
|---|---|--|--|---|
| | EMG Zygomatic. Energy/Key lighting | Skin temp.standard deviation /5 th autocorrelation of MFCC coefficient | Skin temperature range/Shot length variation | EMG Zygomatic. energy/ 15 th /20 bin of hue histogram |
| 1 | 0.24 | - | - | -0.41 |
| 2 | 0.62 | 0.44 | 0.42 | -0.41 |
| 3 | 0.46 | 0.40 | 0.56 | -0.34 |
| 4 | 0.40 | 0.32 | 0.43 | -0.30 |
| 5 | 0.36 | 0.39 | 0.58 | - |
| 6 | 0.44 | 0.31 | 0.51 | -0.32 |
| 7 | 0.47 | 0.34 | 0.27 | -0.43 |
| 8 | 0.54 | 0.34 | 0.42 | -0.45 |

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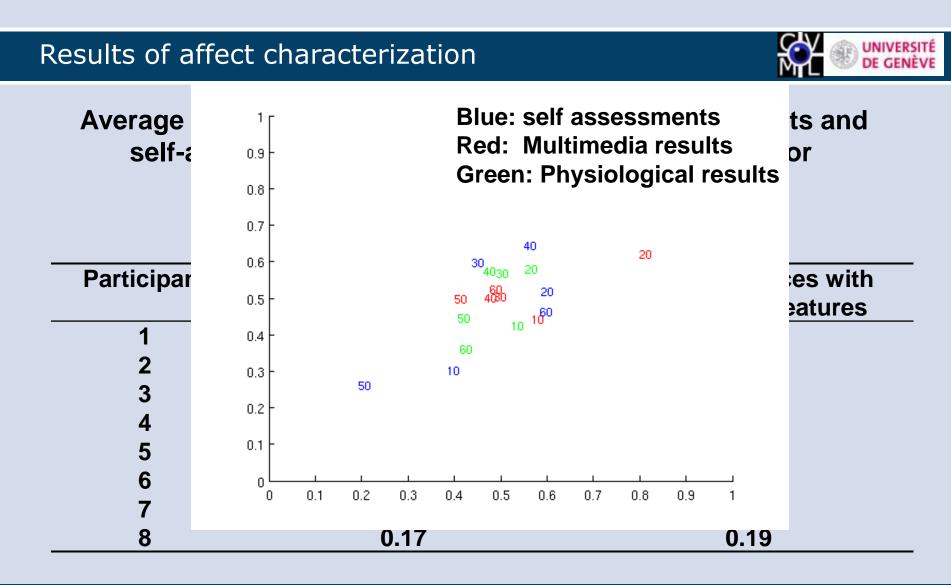
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- Relevance vector machine is used to compute linear regression weights (*w_i*) to estimate arousal/valence scores. (Tipping toolbox)
- y_j is the estimation from arousal or valence value for j-th observation (video clip), x_i is the i-th feature

$$\hat{y}(j) = \sum_{i} w_i x_i + w_0$$

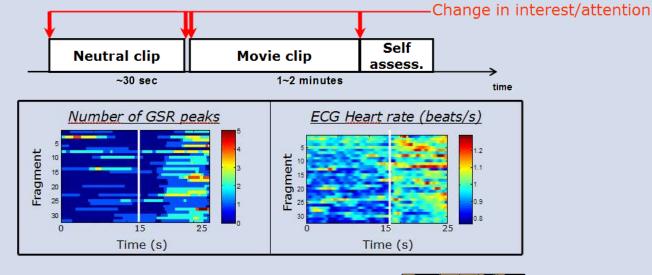
Leave one out cross validation



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Need for training data which contains relevant features for at least two levels of interest.



LDA Classifier output: 72% accuracy

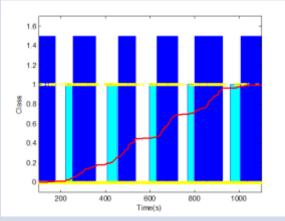


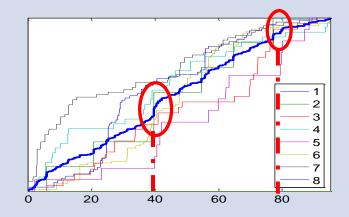
Interest detection during movies

- Analyze full experiment.
- Compare participants.
- Evaluate.













Conclusion



- The correlation between stimuli (multimedia) and response (physiological signals) was shown.
- Results show the ability of the physiological signals for affective characterization
- Temporal detection of interesting movie fragments is feasible.



- Implementation of head-movement and facial expressions as features.
- Obtaining objective measure for detecting interesting episodes instead of visual inspection.
- Using prior information for affect estimation (genre, users rating, etc)
- Modeling and estimating dynamics of affect in movie watching by multimedia features.