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ASSESSING STUDENTS' EMOTIONAL STATES: AN APPROACH TO IDENTIFY LECTURES THAT PROVIDE AN ENHANCED LEARNING EXPERIENCE

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Introduction



PRESENTATION OVERVIEW



- Background
- Motivation
- Methodology
- Case Study
- Results
- Conclusions
- Future Work





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EMOTIONS IN THE CLASSROOM















Research Motivation

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LECTURES THAT ENHANCE LEARNING



need to understand levels of engagement, delight, frustration and boredom

relationship between student attitude and academic achievements Shultz and Lanehart (2002) Shultz and Pekrun (2007) Sungh et al. (2002)



How can we identify lectures that offer an enhanced learning experience?



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EMOTIONS IN THE CLASSROOM

Emotional	Learning	References
State	Gains Impact	
Engagement/	Positive	[21,7]
Interest		
Frustration	Negative	[25,9]
Boredom	Negative	[<u>26</u>]
Confusion	Positive	[<u>9,25,27</u>]
Delight	Positive	[9]

students experience positive mental states while minimizing those mental states associated with negative connotations.







STUDENTS' UNDERSTANDING OF LECTURE MATERIAL

Existing Assessment Techniques

Data Mining Lectures





Limitations in understanding the root causes of poor students' performance





Engrand STATE EMOTIONAL STATES AND LECTURES

Engagement



Frustration

Hypothesis: Students' emotional states are correlated with the lecture characteristics (lecture style and lecture content)

-towards individually customized learning



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Research Hypothesis

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METHODOLOGY: QUANTIFYING STUDENTS EMOTIONAL STATES TOWARDS LECTURES

Experimental Setting	Data Capturing	Data Analysis	Optimal Lectures
Experimental Setting	Data Capturing	Data Analysis	Optimal Lectures
	Background Attitude	$r = \frac{\sum_{i=1}^{n} (X_i - \bar{X})(Y_i - \bar{Y})}{\sqrt{\sum_{i=1}^{n} (X_i - \bar{X})^2} \sqrt{\sum_{i=1}^{n} (Y_i - \bar{Y})^2}}$ $\frac{DF}{\sqrt{\sum_{i=1}^{n} (X_i - \bar{X})^2} \sqrt{\sum_{i=1}^{n} (Y_i - \bar{Y})^2}$ $\frac{DF}{2AC} = \frac{205}{0.44} - \frac{205}{0.51} - \frac{211}{0.51} - \frac{0.12}{0.44} - \frac{0.25}{0.54} - \frac{0.25}{0.51} - \frac{0.55}{0.55} - 0$	



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Methodology

EXPERIMENTAL SETTING



- Students from various fields (degree, habits of study, class participation, interest and area of expertise)
- A Likert scale (1 to 5)

Methodology

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- Lecture composed by various lessons.
- Survey asking students about emotional states during lecture





Methodology

Munoz, Tucker 2014 http://www.engr.psu.edu/datalab/

DATA ANALYSIS



r	Type of relationship	Color Code
$\pm [0.0 \text{ to } 0.2]$	Weak or no relationship	
$\pm [0.2 \text{ to } 0.4]$	Weak relationship	
± [0.4 to 0.6]	Moderate relationship	
$\pm [0.6 \text{ to } 0.8]$	Strong relationship	
± [0.8 to 1.0]	Very strong relationship	

$$r = \frac{\sum_{i=1}^{n} (X_i - \bar{X})(Y_i - \bar{Y})}{\sqrt{\sum_{i=1}^{n} (X_i - \bar{X})^2} \sqrt{\sum_{i=1}^{n} (Y_i - \bar{Y})^2}}$$

n: Sample size

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 X_i : Value of i - th observation from sample X, i: 1 to n \overline{X} : Average value of all observations from sample X Y_i : Value of i - th observation from sample Y, i: 1 to n \overline{Y} : Average value of all observations from sample Y



RELATION TO LECTURE MATERIAL



Value path graphs

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• Value paths range from 1 to 5

Methodology

- Emotions categorized as negative will be normalized to have 5 as more desirable and 1 as less desirable
- Positive emotions and negative emotions are directly compared



CASE STUDY: EMOTIONAL STATES IN THE CLASSROOM





Case Study

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CASE STUDY





RESULTS: CORRELATIONS AMONG EMOTIONAL STATES

Correlation Matrix

	ENG	BOR	INT	FRU	DEL	CON
ENG		-0.74	0.74	-0.45	0.61	-0.36
BOR	-0.74		-0.71	0.42	-0.54	0.54
INT	0.74	-0.71		-0.43	0.65	-0.38
FRU	-0.45	0.42	-0.43		-0.29	0.43
DEL	0.61	-0.54	0.65	-0.29		-0.27
CON	-0.36	0.54	-0.38	0.43	-0.27	

Emotional states included

Engagement (ENG) Boredom (BOR) Interest (INT) Frustration (FRU) Delight (DEL) Confusion (CON)

Interesting insights

ENG is strongly positively associated to INT (0.74) and DEL (0.61), which are usually defined as positive mental states.

ENG, is strongly negatively associated to BOR (-0.74),

moderately negatively related to FRU (-0.45), and weakly negatively associated to CON (-0.36).

In addition, CON is weakly (-0.36, -0.38, and -0.27) or moderately associated (0.54 and 0.43) to all other mental states.



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RESULTS: EMOTIONAL STATES AND PERCEPTUAL FACTORS

Correlation Matrix

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	ENG	BOR	INT	FRU	DEL	CON
DIF	-0.07	0.01	-0.11	0.13	0.05	0.35
BAC	0.44	-0.45	0.50	-0.16	0.36	-0.21
UND	0.51	-0.54	0.61	-0.40	0.36	-0.53
STY	0.77	-0.77	0.68	-0.51	0.55	-0.50

Perceptual factors included

- Perceived difficulty (DIF)
- Background (BAC)
- Understanding (UND)
- Teaching style (STY)

Interesting insights

- The level of perceived difficulty was not significantly correlated with the reported emotional states except for confusion in which a weak relationship was found (0.35).
- Background of the student is not significantly associated to the frustration reported. However, this statement cannot be generalizable for other settings in which more field-specific lectures are presented.
- Teaching style to engage students is correlated to all the emotions reported.





PENNSTATE **RESULTS: CORRELATIONS** AMONG PERCEPTUAL FACTORS

Correlation Matrix

	DIF	BAC	UND	STY
DIF		0.06	-0.24	-0.01
BAC	0.06		0.27	0.32
UND	-0.24	0.27		0.53
STY	-0.01	0.32	0.53	

Perceptual factors included

- Perceived difficulty (DIF)
- Background (BAC)

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- Understanding (UND)
- Teaching style (STY)

Interesting insights

- The reported understanding level (UND) is moderately correlated with the perception of the ability of the speaker to engage the audience, STY (0.53).
- Perceived difficulty (DIF) was not significantly correlated to neither background (BAC) nor speaker style (STY).



Results and Discussion

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RESULTS: VALUE PATH GRAPHS

Value Path Graph (Set of lectures A)

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Lectures A4 and A5 are nondominated video-lectures.

A1, A2, and A3, are considered to be dominated video-lectures.

Value Path Graph (Set of lectures B)



Lectures **B1 and B5 are non**dominated video-lectures.

B2, B3, and B4, are considered to be dominated video-lectures.



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Results and Discussion

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DATA MINING LECTURE MATERIAI

Lecture Analysis

Results and Insights

- **Lecture Related Metrics**
- Like/View (L/V)
- Like/Dislike (L/D)
- Subscriptions/View (S/V)
- Share/View (H/V)

 The ratio S/V was moderately correlated to ENG, BOR, INT, and FRU, and weakly correlated to DEL and CON.

	ENG	BOR	INT	FRU	DEL	CON
L/V	-0.10	0.18	-0.06	-0.29	0.10	-0.09
L/D	0.35	-0.33	0.36	-0.25	0.32	-0.14
S/V	0.53	-0.53	0.55	-0.43	0.33	-0.25
H/V	0.08	0.00	0.13	-0.48	0.05	-0.04
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Results ar	nd Discussior	Muno	z, Tucker 2014	http://www.er	ngr.psu.edu/datala	ıb/ 19



CONCLUSIONS



Engagement

Interest Delight

Positively correlated (r > 0.6)



- **Frustration**
- Confusion
- Positively correlated (r > 0.4)



Engagement and boredom are strongly negatively correlated (r = -0.74)



Confusion has the weakest correlations



Conclusions

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FUTURE WORK

Current and future work

 Capture emotional states automatically (non-invasive sensors)



Potential applications

Identify factors influencing student's emotional states and early advice



Team matching

Evaluate teaching methods and lecture structure



Determine optimum length of a lecture or topic



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Future Work

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Acknowledgement & References

Contributors:

• D.A.T.A. Lab: David Munoz , Conrad Tucker

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QUESTIONS





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