OBSERVING & EVALUATING EDA AS A FUNCTIONAL PREDICTOR FOR INTELLECTUAL TRAITS

Connor Asmus, Ja'Moya Blue, Brock Olson, Maia Pietraho



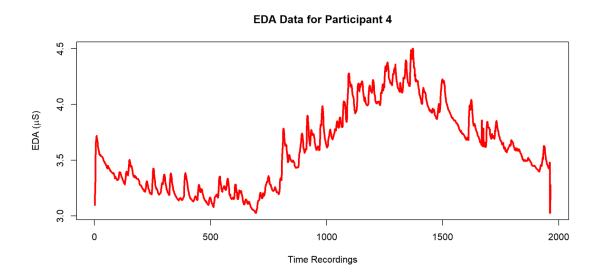






ELECTRODERMAL ACTIVITY (EDA)

- Slight changes in the electrical conductivity of a person's skin due to orienting response (activation of the sympathetic nervous system)
- Skin conductivity increases as sweat glands produce more sweat
- Changes are monitored by passing a small current through the skin and measuring the resistance (collected using a wearable device on the wrist)



time	EDA
1	3.098233
2	3.197894
3	3.302628
4	3.434590
5	3.552590
6	3.660955
7	3.683101
8	3.713844
9	3.720248
10	3.705198
11	3.688225
12	3.674455
13	3.656842
14	3.643712
15	3.620975



OBJECTIVES

- Continue to assess the use of EDA as a proxy for engagement
- Investigate EDA as a predictor for professed intellectual traits

WHAT DO YOU THINK?

Consider the sequence:

2, 4, 6









STUDY DESIGN

Initial Survey:
Provides measure of
a participant's traits

Identify the Number Sequence

Activity 2: Word Association

Activity 3: Rubik's Cube (Right Hand Algorithm)

Post Survey

I min — meditation

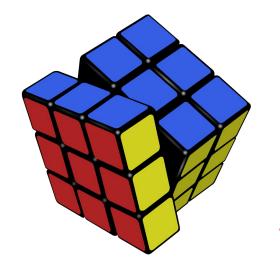
Reflection – questions

Activity I:

BREAD

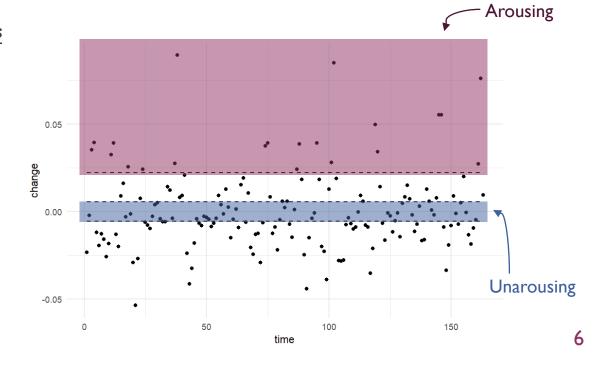
STICK WINNER BASKET

- EDA measured during each activity
- Video recorded the participant's behaviors

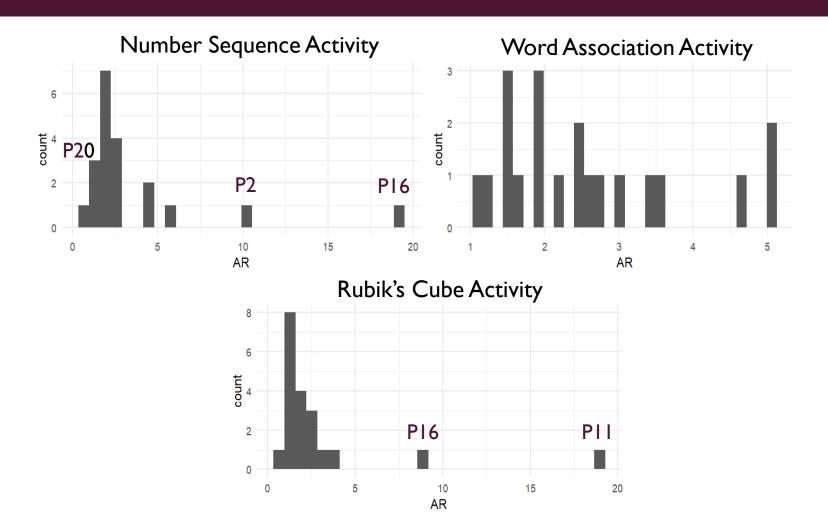


EDA & ENGAGEMENT

- EDA is generally regarded as a trusted proxy for measuring a subject's engagement in an activity
 - Orienting response \rightarrow heightened sensitivity \rightarrow increased intake and processing of information (Raskin, 1973)
 - Arousal ratio (Cain and Lee, 2022): $\frac{\text{# of unarousing moments}}{\text{# of arousing moments}}$
 - Why use EDA?
 - Determines specific times during the experiment in which the participant was seemingly engaged
 - Compare arousal ratios between subjects



AROUSAL RATIOS

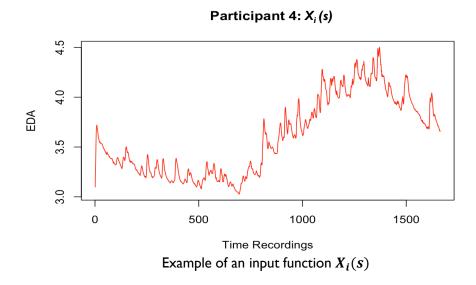


EDA & PROFESSED INTELLECTUAL / ACADEMIC TRAITS

- Used EDA to approximate professed intellectual traits
 - Negative and positive affective states associated with academic tasks
 - Test Anxiety (TA)
 - Mathematical Attitude Valuation and Enjoyment (MA_v and MA_E)
 - Cognitive process ability
 - Intellectual humility (IH)
 - Need for Cognition (NC)

MODELING EDA

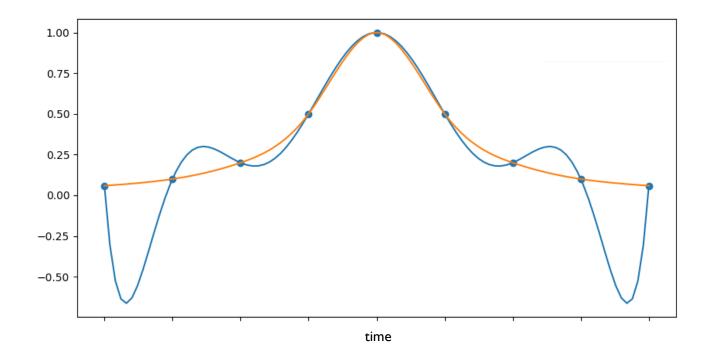
- Functional Data Analysis (FDA): analyzes data where variables are functions rather than a single scalar point
- **Scalar-on-Function Regression (SoFR):** predictor is a function
 - Simple Linear Regression model: $Y={m eta_0}+{m eta_1}x+{m arepsilon}$
 - SoFR model: $Y_i = oldsymbol{eta}_0 + \int_S \ oldsymbol{eta}_1(s) X_i(s) \ ds + oldsymbol{arepsilon}_i$



- Models full temporal pattern of physiological arousal for all subjects instead of relying on summary statistics
- Can make connections between surprising behavior in the data and video evidence
- Integral term gives the overall contribution of EDA for each specific intellectual trait

BASIS EXPANSION

- Estimate the coefficient function using a basis expansion: $\beta_1(s) = \sum_{k=1}^{n} \beta_{1k} B_k(s)$
 - Since $\beta_1(s)$ is an unknown continuous function (infinitely-dimensional), we approximate it as a *finite* linear combination of known *cubic* polynomial basis functions $B_k(s) = [1, s, s^2, s^3]$



- **Binning**: prepares the functional data for modeling by discretizing $X_i(s)$, reducing dimensionality and noise
- **Splining**: improves flexibility and smoothness of the binned data by approximating intervals of $\beta_1(s)$

OBTAINING $\beta_1(s)$

I. Plug basis equation into original equation for simplification

$$E[Y_i] = \beta_0 + \int_{\mathcal{S}} \sum_{K} \beta_{1k}(B(s)) X_i(s) ds = \beta_0 + \sum_{K} \beta_{1k} \left[\underbrace{\int_{\mathcal{S}} B_k(s) X_i(s) ds}_{C_{ik}} \right] = \beta_0 + C_{ik} \beta_{1k}$$

2. Create a final model in matrix form, $E[Y] = Z\beta$, and use ordinary least squares (Arg min_{β} $||Y - Z\beta||^2$) to estimate the best-fitting regression

$$\begin{bmatrix} y_1 \\ \vdots \\ y_{20} \end{bmatrix} = \begin{bmatrix} 1 & C_{1,1} & C_{1,2} & \cdots & C_{1,4} \\ 1 & C_{2,1} & C_{2,2} & \cdots & C_{2,4} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & C_{20,1} & C_{20,2} & \cdots & C_{20,4} \end{bmatrix} \times \begin{bmatrix} \beta_0 \\ \beta_{1,1} \\ \beta_{1,2} \\ \beta_{1,3} \\ \beta_{1,4} \end{bmatrix}$$

$$\mathbf{Y} \qquad \mathbf{Z} \qquad \mathbf{\mathcal{B}}$$

3. Use the extracted $oldsymbol{eta}$, plug back into basis expansion equation to get $oldsymbol{eta}_1(s)$

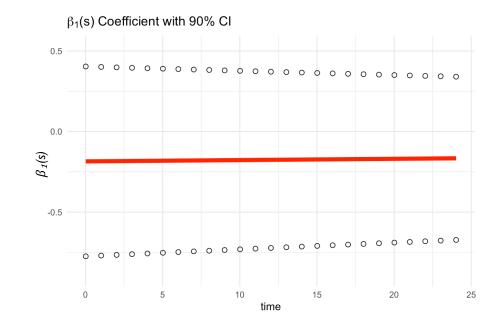
GRAPHING $\beta_1(S)$

Creating the Confidence Band

I. Calculate variance-covariance matrix of $\beta_1(s)$: $B(s)Var(\beta_1)B(s)^T$

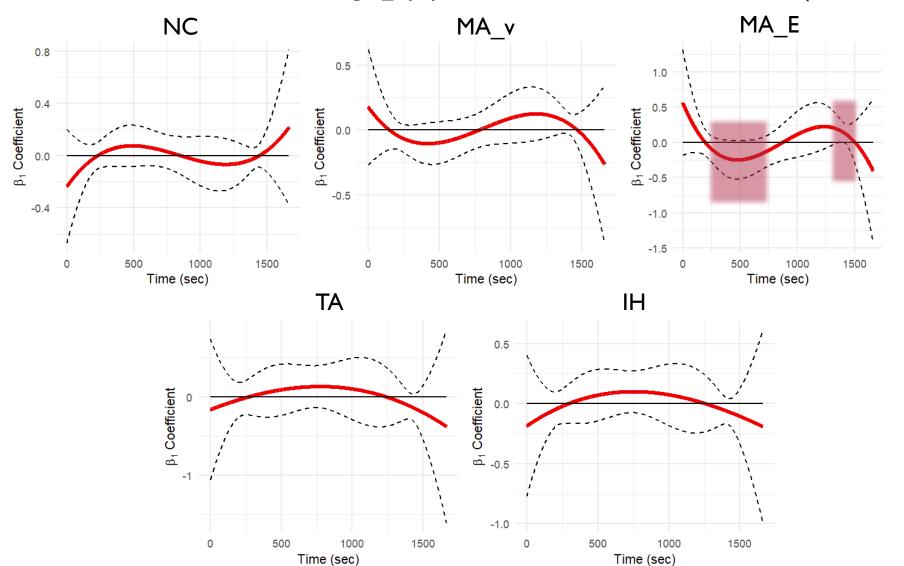
- $Var(\beta_1)$: variance-covariance of the basis coefficient vector (based on OLS)

$$\begin{bmatrix} \sigma_{1,1}^2 & \cos_{1,2} & \cdots & \cos_{1,1664} \\ \cos_{2,1} & \sigma_{2,2}^2 & \cdots & \cos_{2,1664} \\ \vdots & \vdots & \ddots & \vdots \\ \cos_{1664,1} & \cos_{1664,2} & \cdots & \sigma_{1664,1664}^2 \end{bmatrix}$$

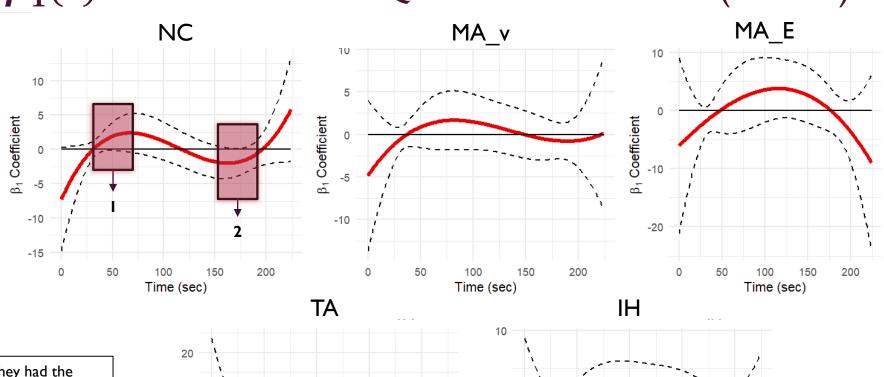


2. Use $\widehat{\beta}_1(s) \pm Z_{\alpha/2} * \sqrt{Var(\beta_1(s))}$ to get pointwise Cl's

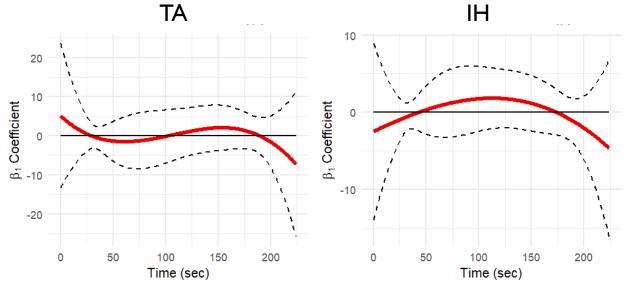
3^{RD} DEGREE MODELS OF $oldsymbol{eta_1}(s)$ FOR ALL ACTIVITIES (90% CI)



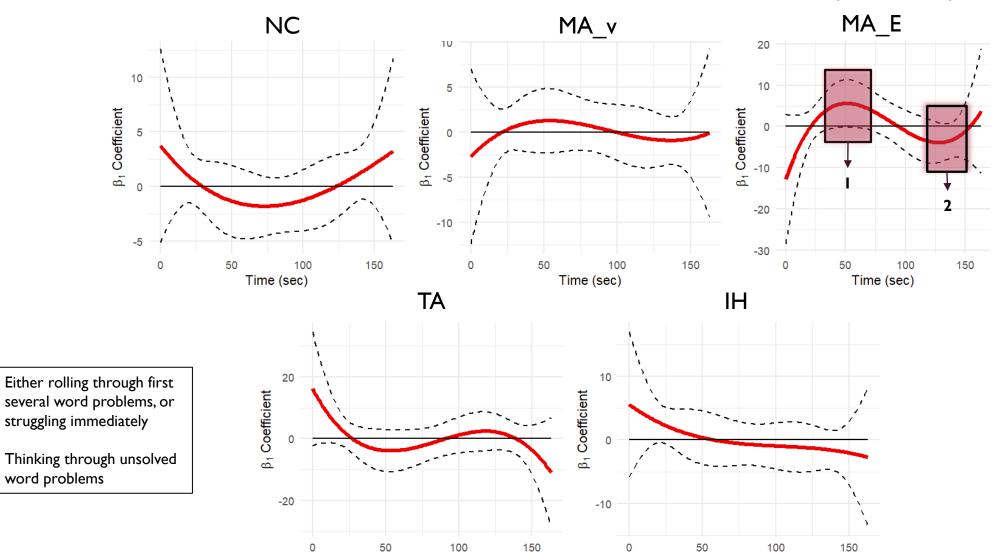
$\beta_1(s)$ FOR NUMBER SEQUENCE ACTIVITY (90% CI)



- I. Quickly believed they had the answer, but were wrong
- Either had the correct answer through trial and error, or were far from it



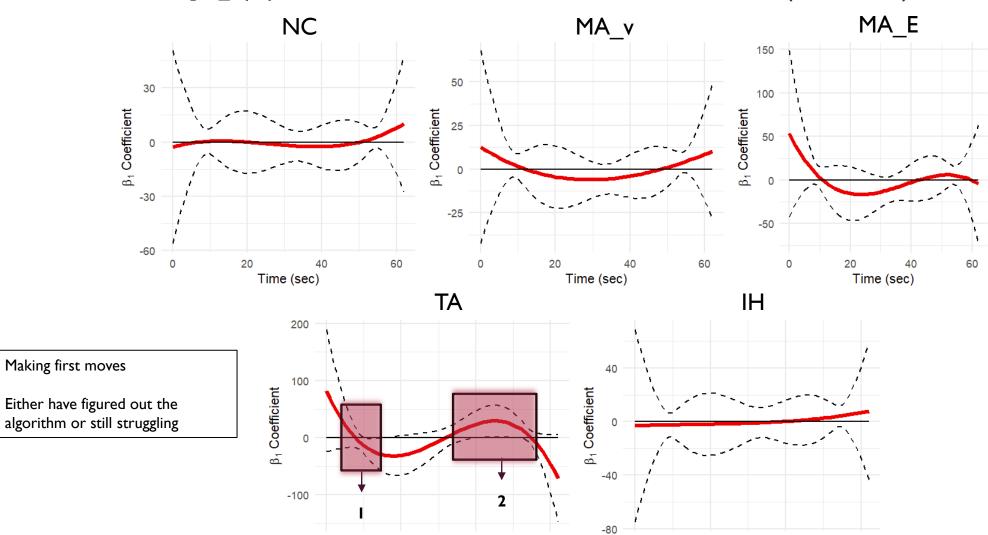
$\beta_1(s)$ FOR WORD ASSOCIATION ACTIVITY (90% CI)



Time (sec)

Time (sec)

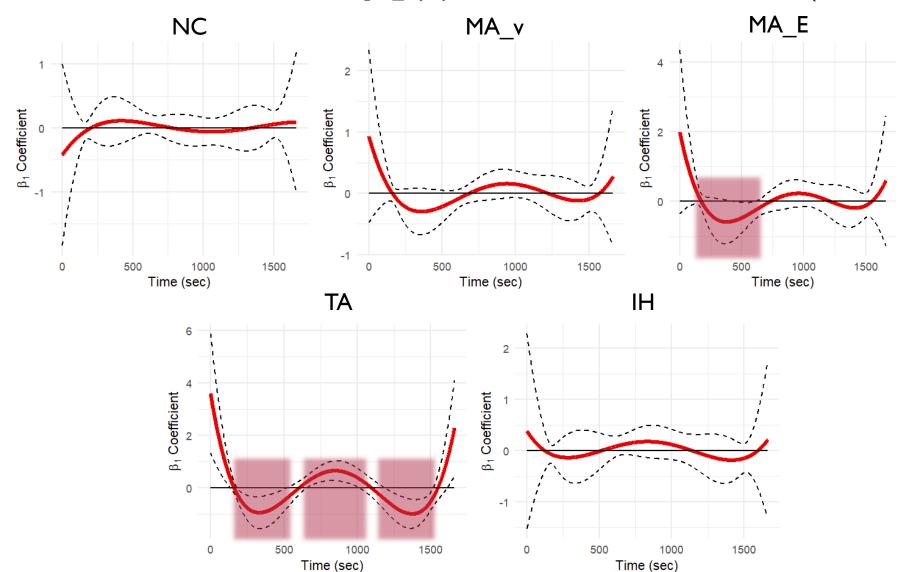
$\beta_1(s)$ FOR RUBIK'S CUBE ACTIVITY (90% CI)



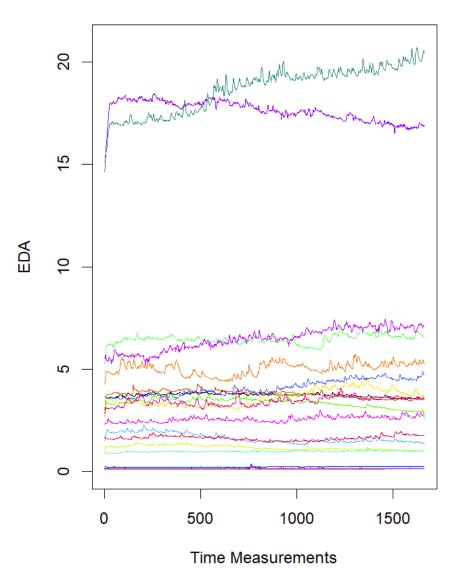
Time (sec)

Time (sec)

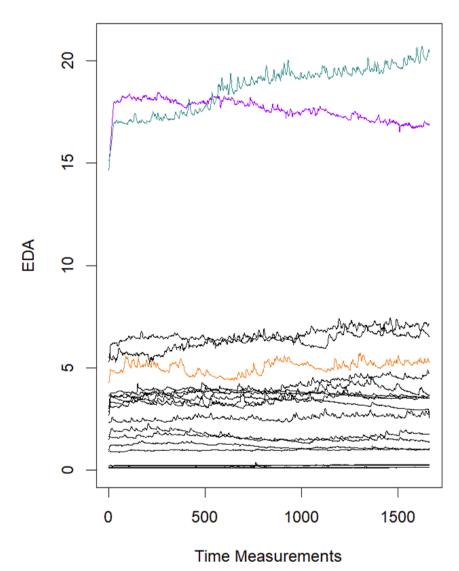
4^{TH} DEGREE MODELS OF $oldsymbol{eta}_1(s)$ FOR ALL ACTIVITIES (90% CI)



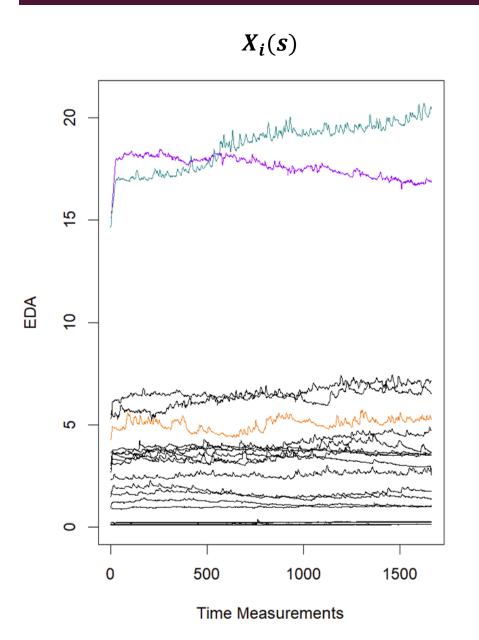


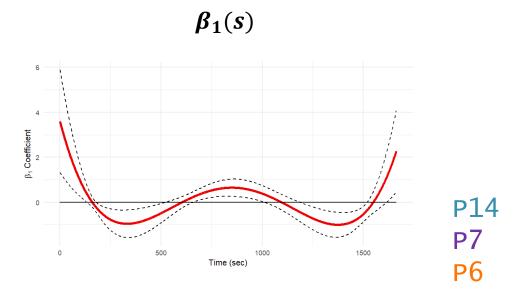


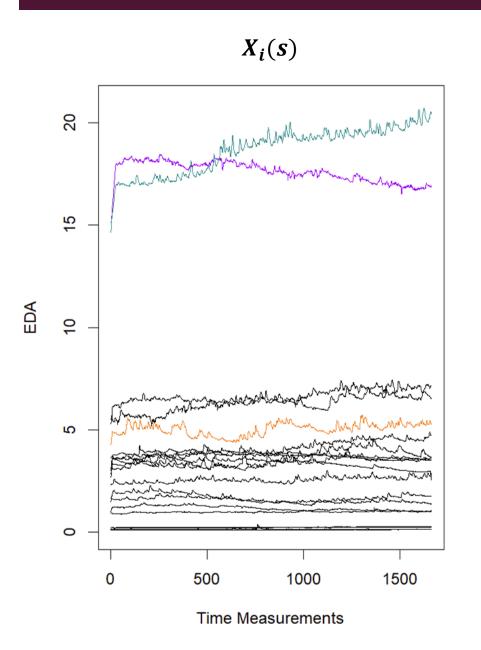


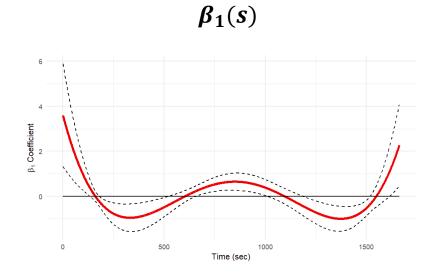


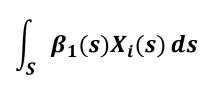
P14 P7 P6







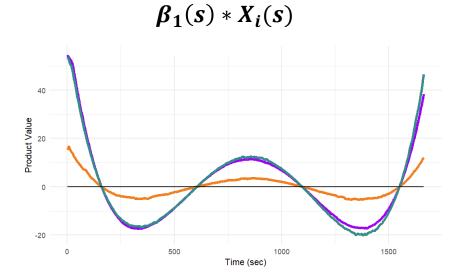




P14: -2.31745214

P7: -1.26100259

P6: -0.28121994



INFLUENCE OF EDA ON EACH OUTCOME VARIABLE

Third Degree Models:

	NC	MA_v	MA_E	TA	IH
All Activities	0.1416	0.0531	0.2212	-0.2910	0.0521
Activity I	0.1289	0.1495	0.6068	-0.0597	0.4507
Activity 2	-0.1154	0.1591	0.5003	-0.4756	0.0424
Activity 3	-0.1130	-0.7689	-1.5574	-0.8488	-0.0458

Fourth Degree Models:

	NC	MA_v	MA_E	TA	IH
All Activities	0.1735	-0.07916	-0.0297	-0.9505	0.0534
Activity I	-0.1421	-0.4622	0.0653	-0.6677	1.1958
Activity 2	-2.4397	-0.6844	0.5167	2.1038	-0.6372
Activity 3	-0.7945	-0.0952	-4.3513	-2.4603	-0.5532

CONCLUSIONS

- The Relationship between EDA and two variables showed promise, which is encouraging for further study
 - Test anxiety (TA)
 - Enjoyment of math (MA_E)
- Results were not as promising for some traits previously thought to be important
 - Intellectual Humility (IH)
- There is reason to believe that EDA data can be used as a proxy to measure affective engagement

FUTURE WORK

- Analysis of larger samples
- Further exploration of the arousal ratio as a metric
- Analyze why EDA is a better predictor for the more encouraging outcome variables
- Attempt Function-on-scalar Regression
- Additional predictors/Random effect model

REFERENCES

- Cain, R. & Lee, V. R. (2022). Measuring Electrodermal Activity in an Afterschool Maker Program to Detect Youth Engagement. In I. Management Association (Ed.), Research Anthology on Makerspaces and 3D Printing in Education (pp. 515-536). IGI Global. https://doi.org/10.4018/978-1-6684-6295-9.ch026
- Haggard, M., Rowatt, W. C., Leman, J. C., Meagher, B., Moore, C., Fergus, T., Whitcomb, D., Battaly, H., Baehr, J., & Howard-Snyder, D. (2018). Finding middle ground between intellectual arrogance and intellectual servility: Development and assessment of the limitations-owning intellectual humility scale. Personality and Individual Differences, 124, 184–193.
 https://doi.org/10.1016/j.paid.2017.12.014
- Lee, Victor R. (2021) "Youth engagement during making: using electrodermal activity data and first-person video to generate evidence-based conjectures." Information and Learning Sciences, vol. 122, no. 3/4, 2021, pp. 270- 291. https://doi.org/10.1108/ILS-08-2020-0178.
- Raskin, D. C. (1973). Attention and Arousal. Electrodermal Activity in Psychological Research. Academic Press, Inc. doi:10.1016/B978-0-12-565950-5.50007-7

ACKNOWLEDGEMENTS

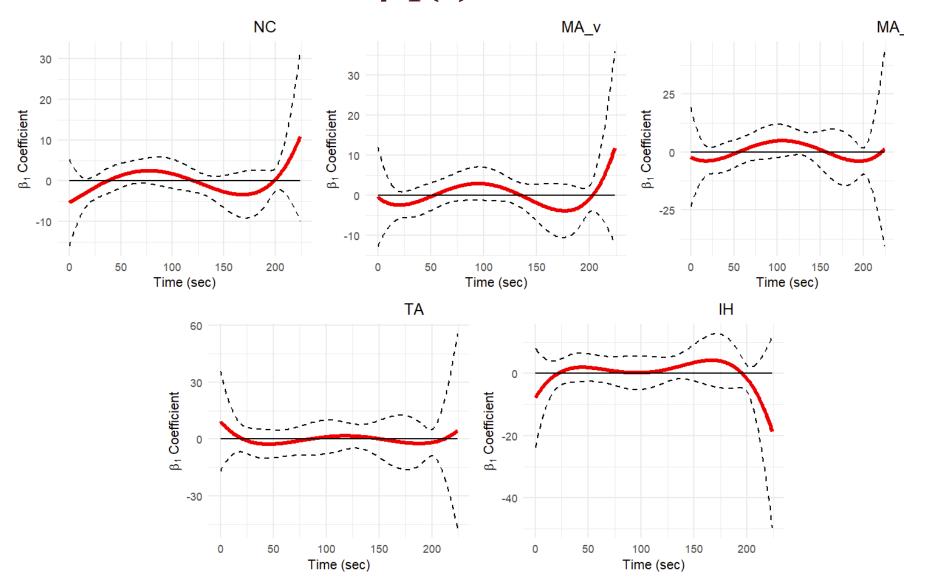
- Dr. Gideon K. D. Zamba, Professor, Dept. of Biostatistics, University of Iowa
- Dr. Matthew Lira, Assistant Professor of Learning Sciences and Educational Psychology, University of Iowa
- Dr. Stacey McElroy-Heltzel, Assistant Professor of Counseling Psychology, University of Iowa
- Minzhi Liu, Graduate RA, Dept. of Psych & Quant Foundations, University of Iowa
- University of Iowa Summer Institute in Biostatistics, College of Public Health
- ISIB Program sponsored by the National Heart Lung and Blood Institute (NHLBI), grant # HL161716-01



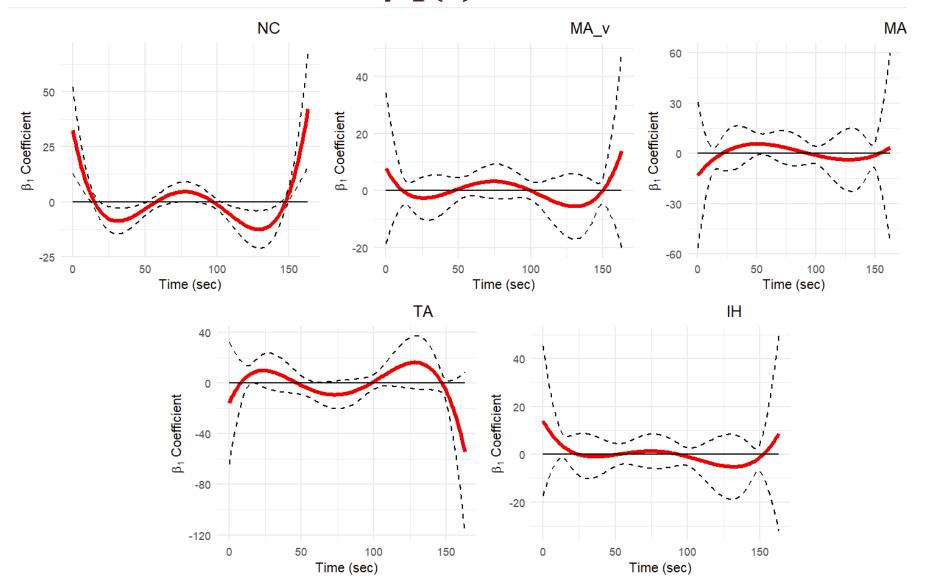


QUESTIONS?

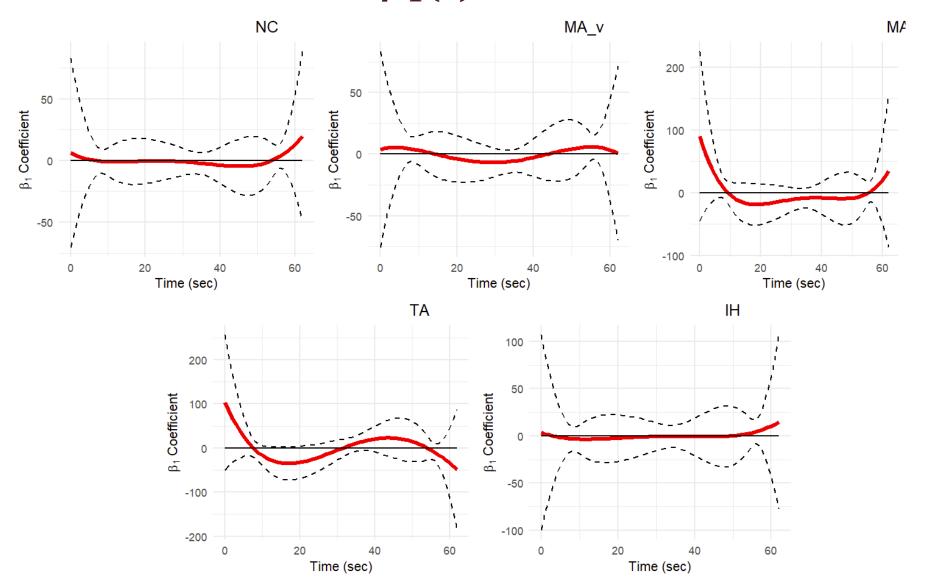
FOURTH DEGREE MODELS OF $\beta_1(s)$ FOR ACTIVITY I WITH 90% CI



FOURTH DEGREE MODELS OF $\beta_1(s)$ FOR ACTIVITY 2 WITH 90% CI



FOURTH DEGREE MODELS OF $oldsymbol{eta_1}(s)$ FOR ACTIVITY 3 WITH 90% CI



INFLUENCE OF EACH OV ON EDA

```
> int_av_table_clusters
    NC
               MA_V
                         MA E
                                               ΙH
                                    TA
    0.1446911
               0.1095379 0.5339742 -0.6788169 0.2027478
act1 0.4254922
               0.3853355 0.8043042 -0.7700489 0.2057776
act2 0.05297876 0.3005304
                          0.7559087 -0.6379394 -0.3078745
act3 0.1489526
                                              -0.0869689
               -0.2527585 -1.408443 -1.31372
> int_av_table
    NC
               MA_V
                          MA E
                                    TΑ
                                                IΗ
    0.1415942
               0.05314017 0.2212419 -0.2910357
                                               0.1520528
               0.1494517
                          0.6068284 -0.05971377 0.4506964
act1 0.1289155
act2 -0.1153542 0.159079
                          0.5003212 -0.4756384 0.04244399
act3 -0.1130072 -0.7689471 -1.557352 -0.8488349 -0.04583517
```