

# Simple procedures to help mitigate the influence of Hair Type and Texture on ERP Data Quality

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# Background

Despite the wide-spread perspective that certain hair types and hair styles may compromise EEG/ERP data quality, no empirical evidence regarding the impact of participant hair has been examined. As researchers work to expand the inclusiveness of participants in EEG/ERP research, guidance on specific practices and techniques that ensure data quality is needed.

## Aims

We aimed to evaluate these assumptions and determine if and how:

- 1. Differences in hair type impact various features of raw recorded and pre-processed EEG data, as well as analysis-ready ERP data; and,
- 2. Those impacts are alleviated by accounting for individual differences in hair volume.

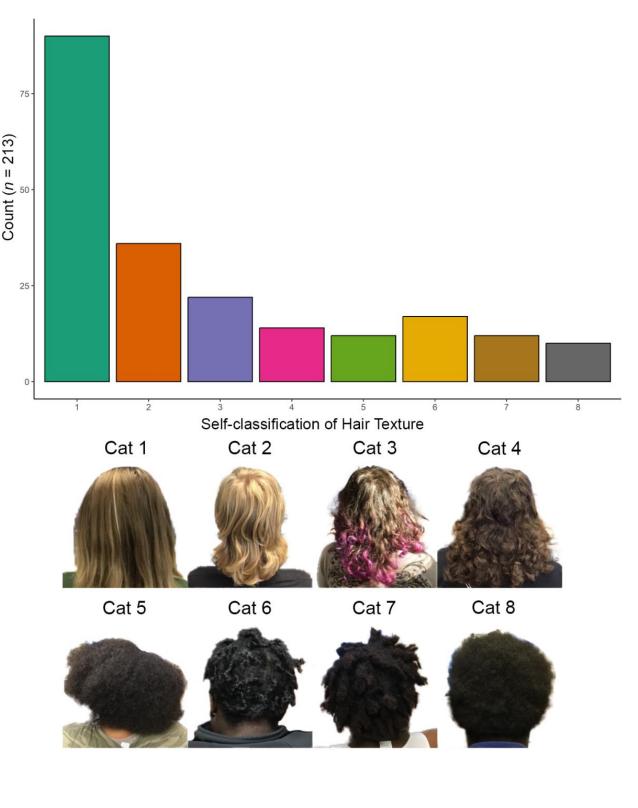
# Methods

#### **Participants**

- 213 young adults  $(M_{age} = 18.34, SD_{age} = 0.42; 48.36\%$  female) from economically disadvantaged rural regions of PA and NC.
- 69% White or Caucasian, 24% Black or African American, 7% Multi-racial.

#### Difference in Hair Type

- Participants self-reported their hair-type using the 8 categories developed by *de la Mettrie, et al., 2006* and *Loussouarn et al., 2007*.
- These categories were developed based on a physical assessment of:
  - Curve Diameter i.e., the arc of hair.
  - Curl Meter i.e., if the hair fits into a circle of a specific size.
  - The number of wave crests within a 4 cm measure of hair.
- Research assistants used graduated syringes to quantify the amount of electrode gel in milliliters used in the EEG cap set-up (i.e., gel volume).



■ Participants were divided into two groups (Group 1 = Categories 1-4,
 n = 161; Group 2 = Categories 5 - 8, n = 52) prior to analysis.

## EEG Data Collection & Metrics

- EEG data were recorded using a 34- Channel Biosemi ActiveTwo System.
- Collected during a resting baseline and the Monetary Incentive Delay Task.

#### **Data Quality Metrics**

- 1. Raw recorded data were evaluated using the DC Offset calculated as the mean voltage across the full recording.
- 2. After passing through a fully automated pre-processing pipeline (bad channels removed, bandpass filter at 0.1 30 Hz, ICA to remove eye and muscle activity) pre-processed data were evaluated using a Signal-to-

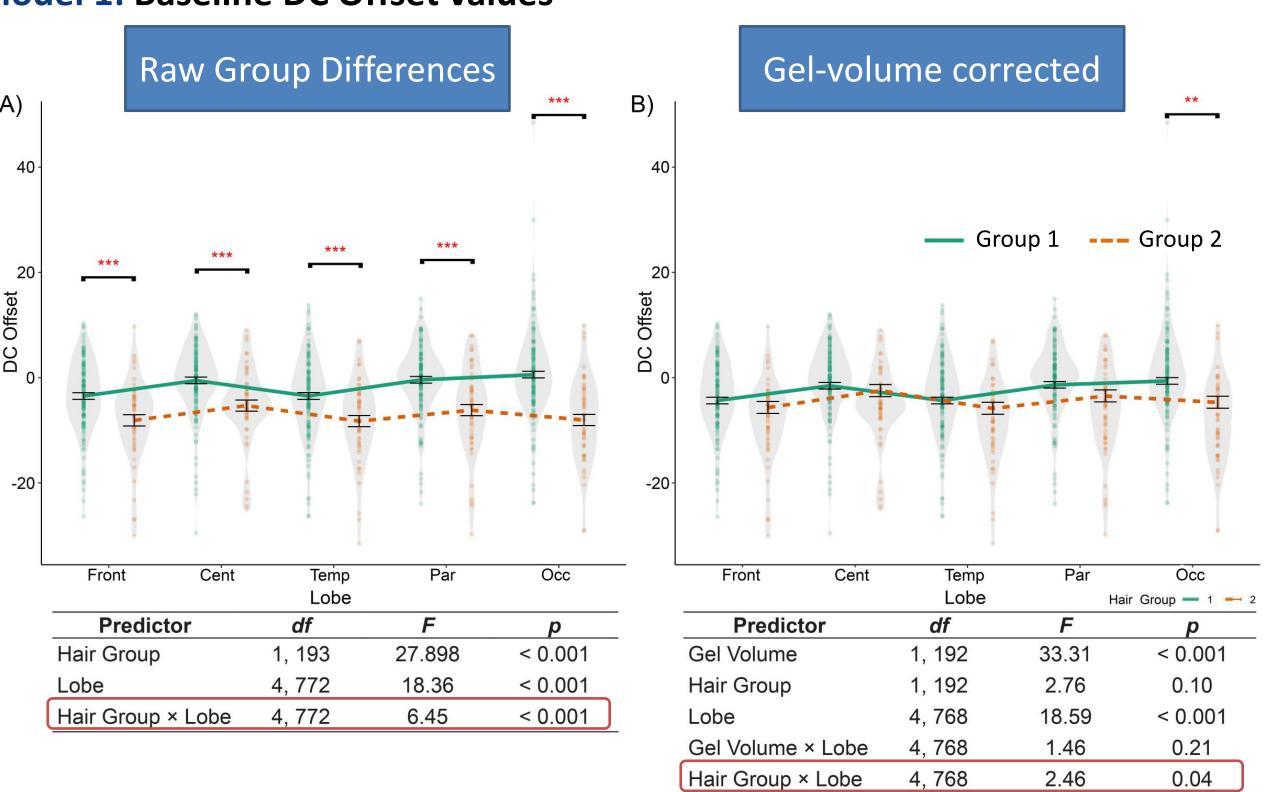
Noise Ratio calculated as: SNR = 10

ted as:  $SNR = 10 \cdot log_{10} \frac{\sum_{(i=1)}^{N} x_i^2}{\sum_{(i=1)}^{N} (s_i - x_i)^2}$ 

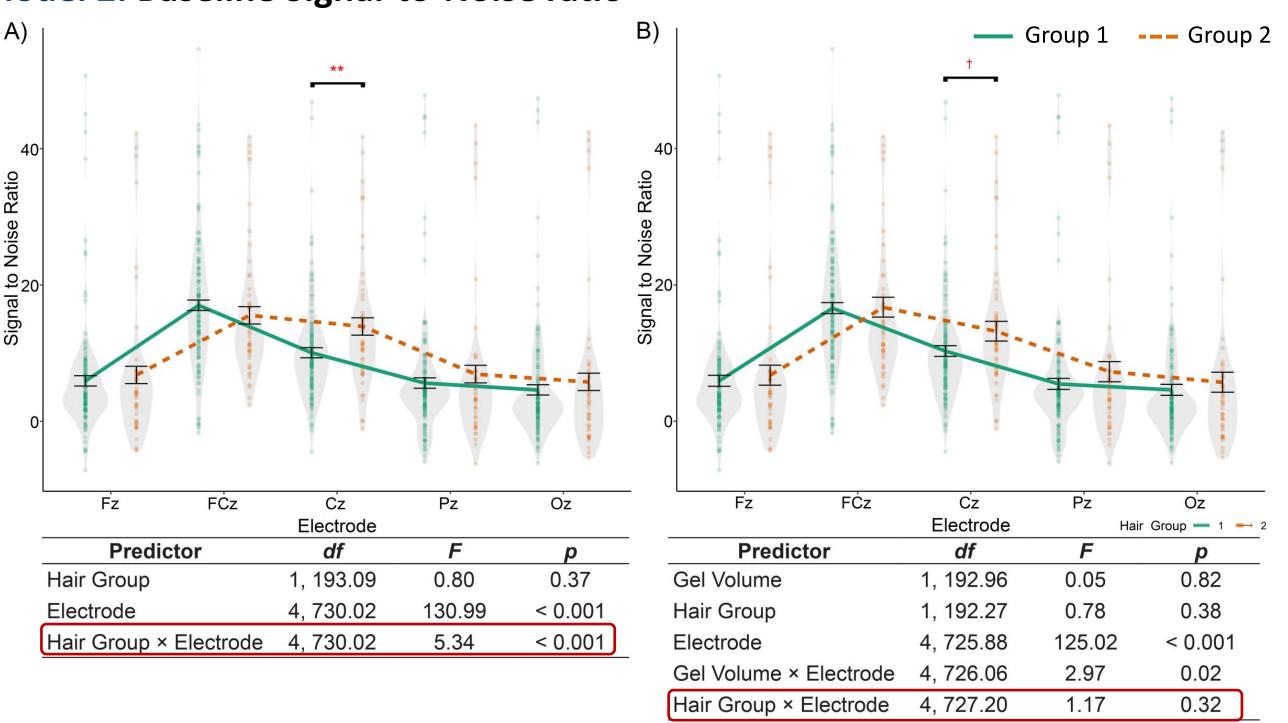
3. P1 and P3b ERP amplitude was used to represent analysis ready data.

# Results

**Model 1:** Baseline DC Offset values

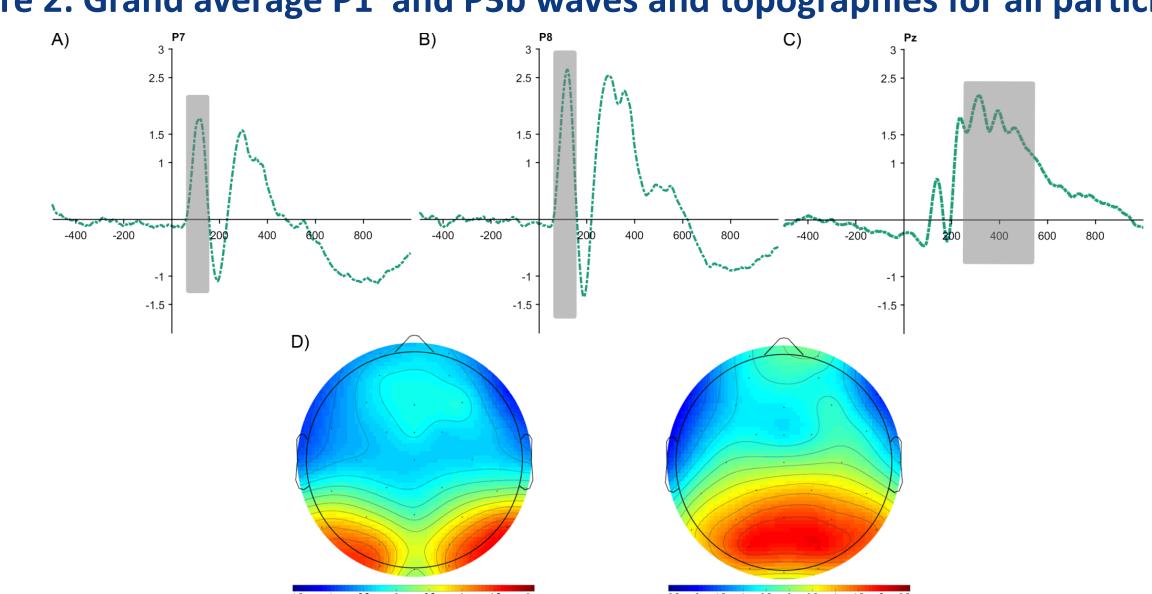




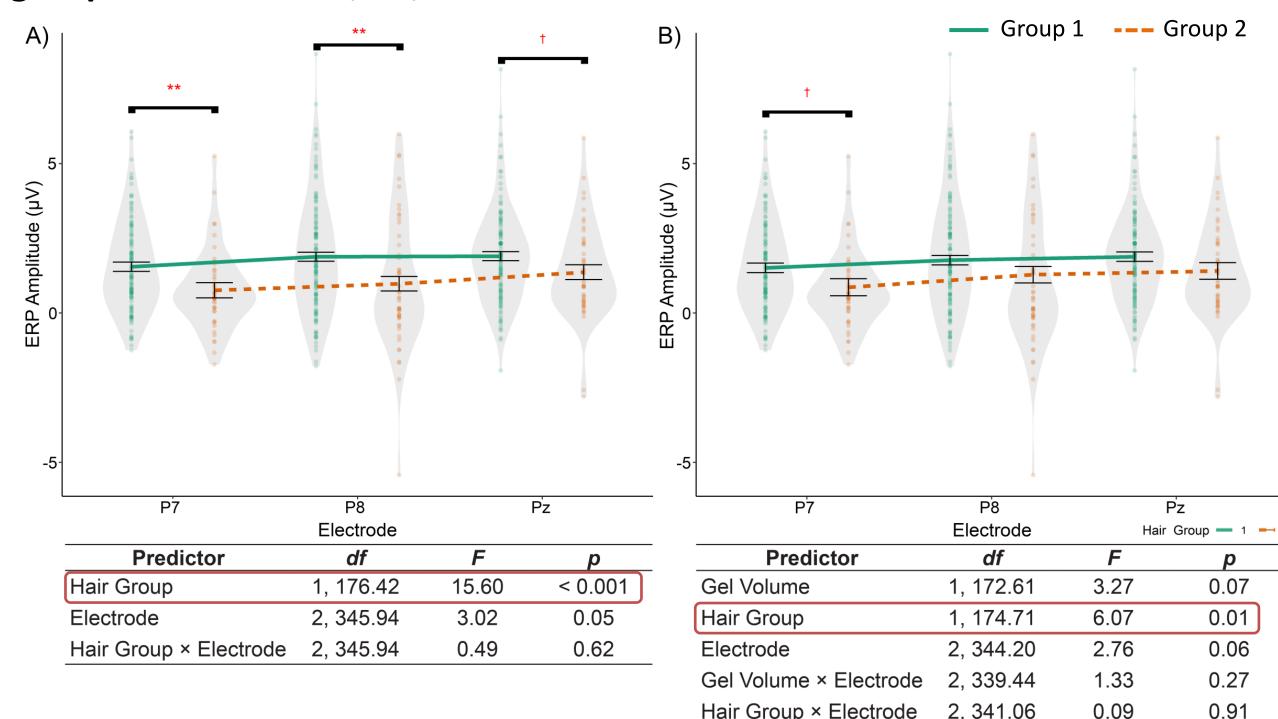


- Hair groups differ in data quality metrics, but <u>these differences are</u>
  <u>mitigated or eliminated when accounting for gel volume</u>.
- Comparable results were observed using data from the MID task.

Figure 2: Grand average P1 and P3b waves and topographies for all participants



Model 3: Comparison of ERP amplitude values between self-categorized hair groups across the P7, P8, and Pz electrodes



- Group differences in ERP amplitude also exist, such that on average participants with more textured hair had lower ERP amplitudes.
- These differences are reduced by accounting for gel volume.

### Conclusions

Measuring gel volume and including it as a covariate may help to account for individual differences in hair volume and significantly reduces group differences.

- Such differences are highly problematic as they could be misattributed to cognitive differences among groups.
- It is worth noting that these results apply to our system and ERP analysis. Additional research is needed to evaluate:
  - Other EEG data-collection systems.
  - EEG derived indices of analysis e.g., spectral features.
  - Alternative measures of hair volume.