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Spatial Frequency Implications for Global and Local Processing in Autistic Children

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Main Finding

- Paired t-test indicated that the difference between the *entire image's SF* and the *average hotspot SF* was **statistically significant**

SF of Entire Image vs. Hotspots for Filtered Images

- Mean: whole image: 144756.193, hotspot: 15540.080
- T-test: t-statistic = -63.135, p-value < 0.001

SF of Entire Image vs. Hotspots for Raw Images

- Mean: whole image: 172865.634, hotspot: 15579.260
- T-test: t-statistic = 77.919, p-value < 0.001

SF of Entire Image in Filtered vs. Unfiltered Images

- Mean (SD): Raw: 172865.634, 13823.480; Filtered: 144756.193, 13957.312
- T-test: t-statistic = 14.864, p-value < 0.001

SF of Hotspots in Filtered vs. Unfiltered Images

- T-test: t-statistic = 0.631, p-value = 0.531

Discussion

Implications

- The SF of an image and its hotspots **affect** the filter as it is designed

Weaknesses

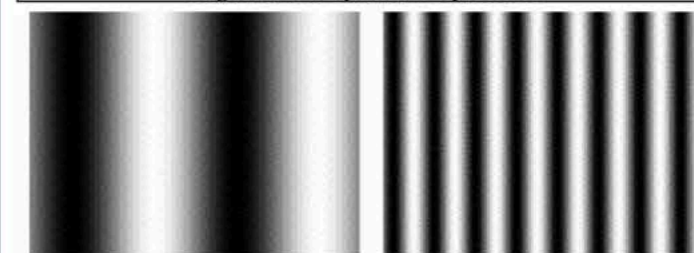
- The use of the head-mounted tracker did not allow for fixations to be automatically recorded by eye-tracking software

Future Work

- Investigate the role of SF in participants encountering hotspots
- This research lays the groundwork for potentially improving the filter to create a significant **decrease** in *local interference*
- Once COVID restrictions allow and a screen-mounted eye gaze device is available, the team plans to test more participants

Extra Tables & Figures

High vs. low spatial frequencies



Low

High

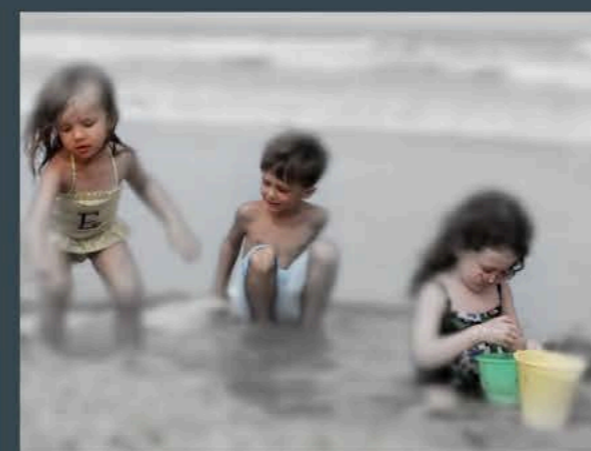
Formula used to calculate SF of entire image and hotspots

```
def spatial_freq(image_rect):
    #M = rows, N = columns
    M,N = image_rect.shape
    #Calculating row frequency
    row_sum = 0
    for j in range(0,M-1):
        for k in range(1,N-1):
            row_sum = row_sum +
                (image_rect[j,k] - image_rect[j,k - 1])**2
    row_freq = math.sqrt((1/M*N)*row_sum)
    #Calculating column frequency
    col_sum = 0
    for k in range(0,N-1):
        for j in range(0,M-1):
            col_sum = col_sum +
                (image_rect[j,k] - image_rect[j-1,k])**2
    col_freq = math.sqrt((1/M*N)*col_sum)
    #Calculating final spatial freq
    SF = math.sqrt(row_freq**2 + col_freq**2)
    return(SF)
```

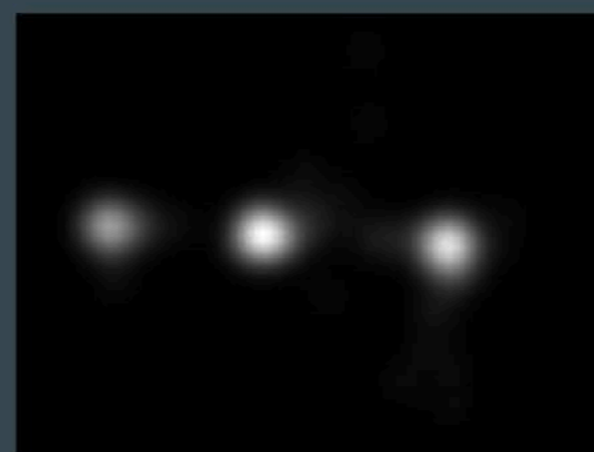
Main finding: There was a **statistically significant** difference between the *spatial frequency* for the *entire image* and the *average spatial frequency* for the *normative visual fixations areas*.



Raw, unfiltered image



Filtered image



Heat map of hotspots where normative individuals fixated



Spatial frequency implications for global and local processing in Autistic children

Riya Mody, Ayra Tusneem, Dr. LouAnne Boyd, Dr. Vincent Berardi

Introduction

Visual Perception (VP)

- VP requires integration of both **global** and **local** sensory input
- Local interference:** local details prioritized over global input
- Overall project aim:** decrease local interference in **Autism Spectrum Disorder (ASD)** by using a filter to *highlight global features*

Research Question

- Does **spatial frequency (SF)**, a measure of *contrast*, affect the filter's functionality and an individual's likelihood of *fixating* on *normative* areas?

Methods

Participants and Test Procedures

- N=10, Ages 9 -18 , ASD diagnosis
- Conducted two sessions over two days
- Randomly presented 50 filtered and unfiltered images on Day 1 and another version on Day 2

Data Collection

- Head-mounted eye gaze device to track *fixations*
- Assistants scored each video on whether the participant hit a **hotspot** (areas where normative individuals looked the most)

Statistical Analyses

- Used the **OpenCV** image processing package in Python to isolate each image's hotspots, draw **contours** around them and create *rectangles* around each
- Used a formula to calculate the **SF** of each image (filtered and unfiltered) and hotspot
- Created a dataframe of all SF results
- Performed **paired sample t-tests** to examine differences in SF by *hotspot vs. entire image* and *filtered vs. unfiltered image*

Spatial Frequency Formula

In a $M \times N$ image where M = number of rows and N = number of columns, row and column frequencies are:

$$\text{Row_Freq} = \sqrt{\frac{1}{MN} \sum_{j=0}^{M-1} \sum_{k=1}^{N-1} [F(j,k) - F(j,k-1)]^2}$$

$$\text{Column_Freq} = \sqrt{\frac{1}{MN} \sum_{k=0}^{N-1} \sum_{j=1}^{M-1} [F(j,k) - F(j-1,k)]^2}$$

The **total spatial frequency** is then:

$$= \sqrt{(\text{Row_Freq})^2 + (\text{Column_Freq})^2}$$