# Towards Improved Recognition and Diagnosis of Autism among Females – A Novel Approach using Machine Learning

Aanika Tangirala, Lakeside School, Seattle, WA, USA | BEHA023

# BACKGROUND

**Autism Spectrum** Disorder (ASD) affects 1 in every 36 children<sup>1</sup> **AUTISM SPECTRUM** DISORDER

Current diagnostic process is **time** consuming, inaccessible, costly, and **subjective** 

Leading Theories

Autism Parenting Magazine, 2024

It can take up to 13 **months** to get a diagnosis

ASD is diagnosed solely on behavioral **observation** – no reliable biomarkers

**Early intervention** 

key for successful

outcomes in life<sup>2</sup>

undiagnosed by

intervention has

age 8, therapeutic

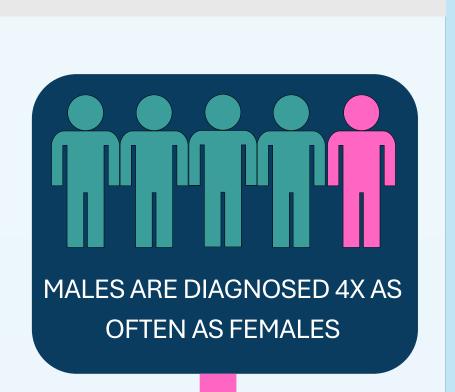
For the **27**%

dissipated<sup>3</sup>

and **diagnosis** is

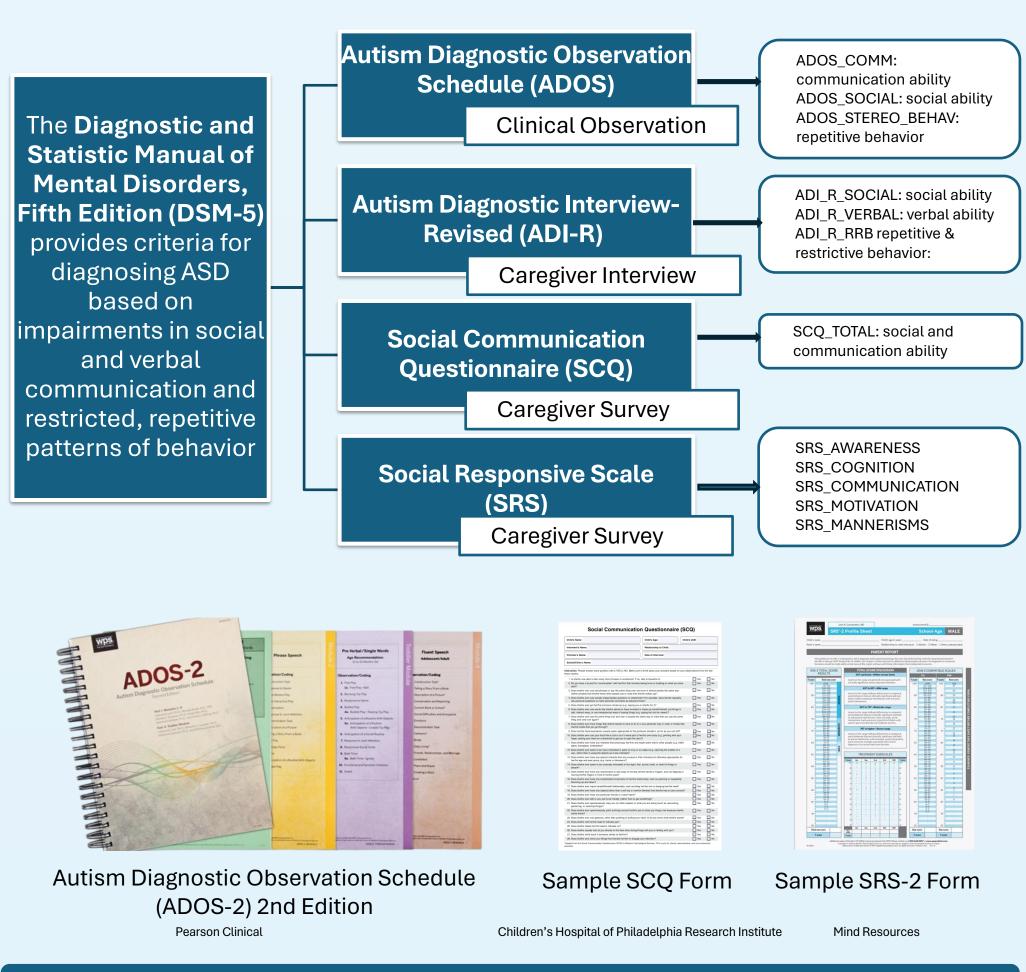
#### THE HIDDEN ISSUE: GENDER BIAS IN DIAGNOSIS

- Females with autism are often misdiagnosed or miss diagnosis due to biased diagnostic tests<sup>4</sup>
- Female Autistic Phenotype exists<sup>5</sup>
- Females tend to camouflage and **mask** symptoms<sup>5</sup>
- Delayed diagnoses and support leads to higher % of **anxiety** and depression in teen girls with autism compared to male peers<sup>6</sup>



Difference cannot ONLY be attributed to something inherent in females that decreases the chances of developing autism.

# **CURRENT AUTISM DIAGNOSTIC TESTS**



Clinicians use a combination of above assessments to diagnose autism

## RESEARCH QUESTION AND GOALS

Research Question: Which features of existing Autism Spectrum Disorder (ASD) diagnostic tests are most useful for predicting autism in **female** populations?



1) Establish diagnostic gender bias and behavioral differences between females and males with autism in diagnostic tests.



2) Identify **minimal** set of behaviors and features most predictive of autism in females.



3) Create a more accurate diagnostic **predictive** machine learning model for females.

# DATA AND METHODS

#### **DATA SOURCE**

- ABIDE the Autism Brain Imaging Data Exchange<sup>7</sup>
- Public online repository of phenotypic autism behavioral data collected over **17** global sites
- Contains 73 features across ADOS, ADI-R, SRS, SCQ
- Dataset size = 1112 individuals, 948 males, 164 females
- 539 individuals diagnosed with autism and 573 without

- Remove outliers
- Remove features not related to ADOS, ADI-R, SCQ, SRS tests, from 73 to 23 features

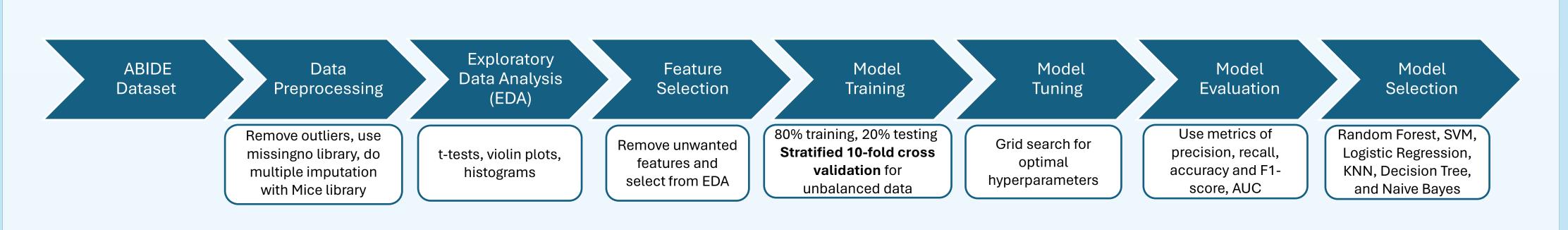
DATA PREPROCESSING

 Use multiple imputation methods for missing data based on correlation

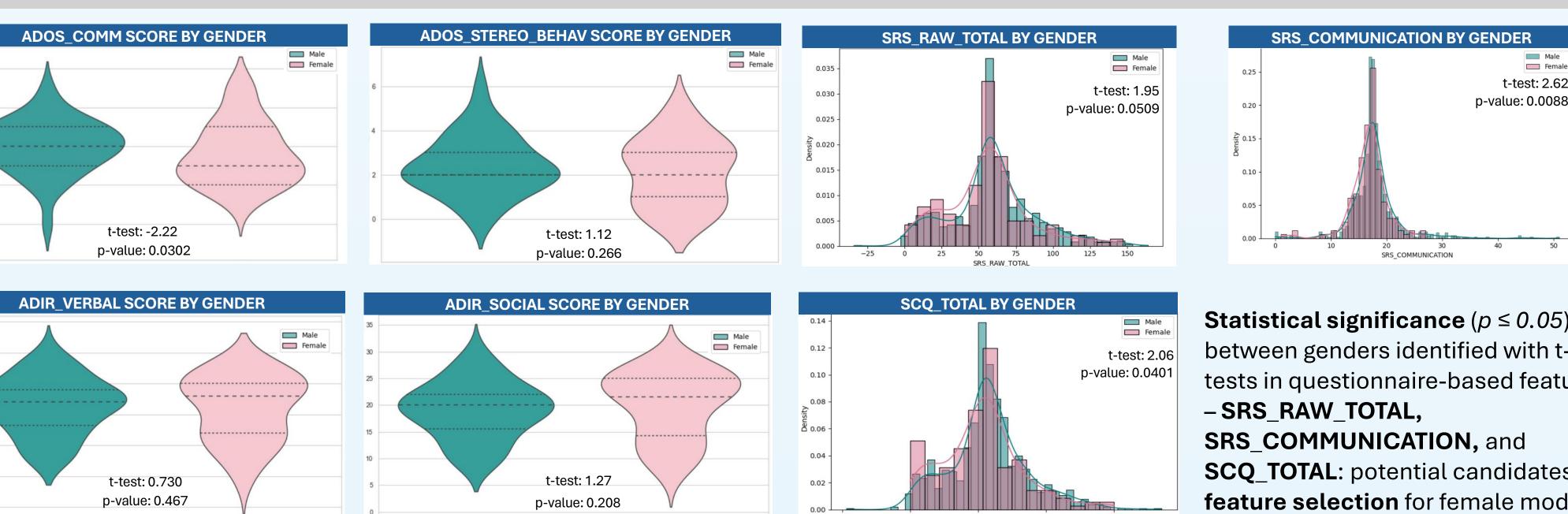
### **EXPLORATORY DATA ANALYSIS (EDA)**

- Descriptive statistics of the features
- Visualization of diagnostic features by gender using histograms and violin plots
- Check for candidate gender-specific features for feature selection and towards Goal 1

#### **MACHINE LEARNING PIPELINE**



# DATA ANALYSIS AND RESULTS



 Males have much sharper distributions centered around the

(1) Built Random Forest

classifier on processed

Features

ABIDE dataset with all 23

features to predict autism

Females have bimodal distributions

1) ESTABLISHED DIAGNOSTIC GENDER BIAS AND BEHAVIORAL

DIFFERENCES IN CURRENT TESTS

(2) Trained and tested on male and

female data combined (Model 1)

on female dataset only (Model 2),

on male dataset only (Model 3)

**Comparison of Evaluation Metrics for All-Feature Models** 

Females have higher scores in ADI\_R

 Statistical difference in ADOS COMM between males and females

(3) Checked model

evaluation metrics:

accuracy, precision,

ВОТН

MALE

Female dataset has least accuracy on

all-feature model, Males have most

accuracy

Validates gender bias of tests

recall, etc.

**Statistical significance** ( $p \le 0.05$ ) between genders identified with ttests in questionnaire-based features

SCQ\_TOTAL: potential candidates for feature selection for female model

## Diagnostic tests capture and classify female autistic behavior differently than males

#### 2) IDENTIFIED MINIMAL FEATURE SET FOR AUTISM PREDICTION IN FEMALES

(4) Checked Top 10 most impactful features for each model. Gender based differences observed

**(5)** Using Top 10 + candidates from EDA, built minimal feature set of 12 features

**Comparison of Top 10 Impactful Features Across Genders** Top 10 features for Top 10 features for Top 10 features for males + females females males AQ\_TOTAL AQ\_TOTAL ADI\_R\_SOCIAL\_TOTA ADOS\_GOTHAM\_SEVERITY ADI\_R\_VERBAL\_TOTAL ADI\_R\_VERBAL\_TOTAL ADOS\_SOCIAL ADOS\_SOCIAL SCQ\_TOTAL ADI\_RRB\_TOTAL ADOS\_COMM ADI\_RRB\_TOTAL SCQ\_TOTAL ADI\_R\_VERBAL\_TOTAL SRS\_RAW\_TOTAL ADOS\_COMM AQ\_TOTAL ADOS\_COMM ADOS\_TOTAL ADOS\_GOTHAM\_SEVERITY ADOS\_SOCIAL ADOS\_GOTHAM\_RRB SRS\_RAW\_TOTAL ADOS\_GOTHAM\_SOCAFFECT ADI\_R\_SOCIAL\_TOTA SCQ\_TOTAL SRS\_RAW\_TOTAL ADOS\_GOTHAM\_SEVERITY SRS\_COGNITION Interview (ADI\_R) and

ADOS\_TOTAL, ADIR\_RRB not impactful for prediction in female dataset

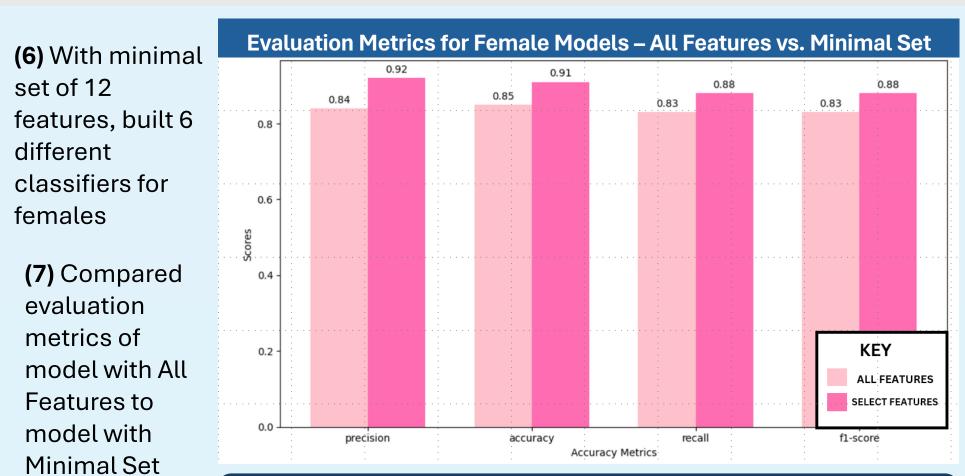
Questionnaire (SCQ, SRS) features more impactful for females

**Minimal Set of 12** most predictive features for females established

**ADOS COMM** SCQ TOTAL **AQ TOTAL SRS RAW TOTAL ADOS SOCIAL** 

**SRS AWARENESS** SRS COMMUNICATION ADOS GOTHAM RRB ADOS GOTHAM SOCAFFECT ADIR SOCIAL TOTAL ADIR VERBAL TOTAL ADOS GOTHAM SEVERITY

## 3) CREATED MORE ACCURATE AUTISM DIAGNOSTIC PREDICTIVE ML MODEL FOR FEMALES



Minimal Set feature model for females more accurate than

All Features model (91% vs. 85%)

0.84

**Evaluation Metrics and AUROC for different classifiers for Females Autism Prediction with Minimal Set** Random Forest SVM Logistic Regression KNN Decision Tree Naive Bayes

# Random Forest (AUROC = 0.97) SVM (AUROC = 0.63) Decision Tree (AUROC = 0.88) — Naive Baves (AUROC = 0.93)

Random Forest is most accurate classifier and has largest AUROC on Minimal Features **Set model for Females** 

CONCLUSIONS

- Significant differences in behavior of males and females with autism and their performance on social, communication, and repetitive subtests; supports Female Autistic Phenotype hypothesis
- Gender bias in current diagnostic tests proven, more accurate in diagnosing males than females (95% vs 85%)
- Random Forest classifier with minimal set of 12 features (from original 23, 48% reduction) was more accurate in predicting female autism (91% vs 85%)
- Questionnaire and caregiver interview-based tests were better indicators for female autism; supports camouflaging and masking behavior in females.
- More female data across multiple sites and access to item-specific scores would potentially lead to even deeper insights

# **FUTURE RESEARCH**

- Working with Children's National Neurodevelopmental Program (Washington DC) for deeper item level sub-score analysis
- Similar analysis and models for removing bias based on race and socioeconomic factors
- Generative adversarial networks (GANs) to address lack of female data issues
- Combination of behavioral and neuroimaging data to provide more accurate diagnosis

# **APPLICATIONS OF RESEARCH**

- Modify DSM-5 and current diagnostic tests for females remove overreliance on ADOS test and repetitive behaviors sub-tests
- Weight questionnaire and interview-based responses more heavily; use these for early screening
- Develop digital and mobile health approaches for faster and expedient screening
- Raise awareness for caregivers, teachers, and parents to **identify female autistic behaviors** and flag for earlier diagnosis

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