

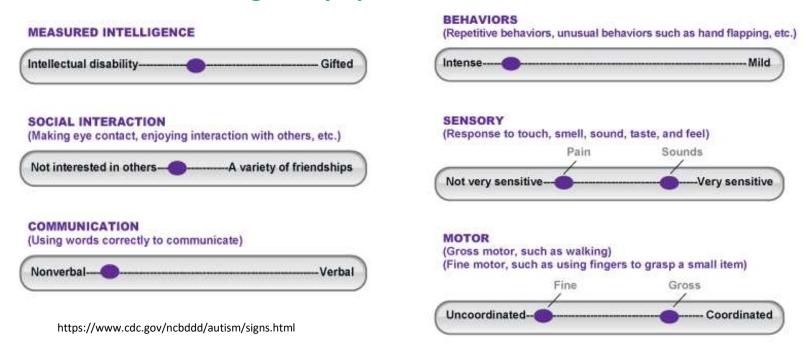
Yimiao Ou

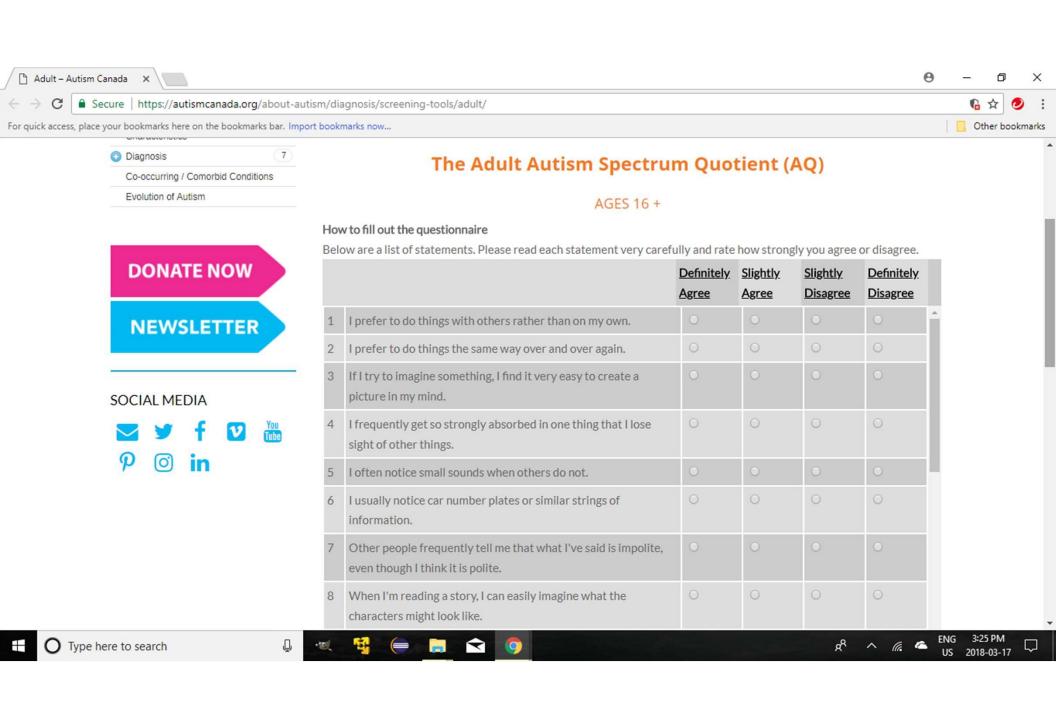
Autism Spectrum Disorder (ASD or autism):

- A neurodevelopmental disorder that impacts brain development.
- Affecting individual's learning ability, interaction and communication with others.
- It begins early in childhood and lasts throughout an individual's life.
- About 1 in 68 children has been identified with ASD.

 Centers for Disease Control and Prevention (CDC) https://www.cdc.gov/ncbddd/autism/index.html

Range of Symptoms







Can we correctly predict autism based on signs and symptoms? What are the most important symptoms(features) in autism prediction?

Dataset: Autism Screening Adult Data Set

https://archive.ics.uci.edu/ml/datasets/Autism+Screening+Adult

Data mining software: Scikit-Learn(python), RapidMiner

Data mining functions:

Classification: Decision Tree (autism/not autism) Naïve Bayes

Logistic Regression

Support Vector Machine

Feed-forward Neural Network

k-NN

Association rules(RapidMiner)

Data attributes

A1	A2	A3	A4	A5	A6	Α7	A8	A9	A10	age	gender	ethnicity	jaundice	autism	country_of_res	used_app_before	result	age_desc	relation	class
1	1	1	1	0	0	1	1	0	0	26	f	White-European	no	no	'United States'	no	6	'18 and more'	Self	NO
1	1	0	1	1	0	1	1	1	1	27	m	Latino	yes	yes	Spain	no	8	'18 and more'	Parent	YES

A1: I often notice small sounds when others do not.

A2: I usually concentrate more on the whole picture, rather than the small details.

A3: I find it easy to do more than one thing at once.

A4: If there is an interruption, I can switch back to what I was doing very quickly.

A5: I find it easy to read between the lines when someone is talking to me.

A6: I know how to tell if someone listening to me is getting bored.

A7: When I am reading a story I find it difficult to work out the character's intentions.

A8: I like to collect information about categories of things.

A9: I find it easy to work out what someone is thinking or feeling just by looking at their face.

A10: I find it difficult to work out people's intentions.

age: age in years

gender: male(m) or female(f)

ethnicity: string

jaundice: whether the case was born with jaundice(yes/no)

autism: whether any immediate family member has autism(yes/no)

country_of_res: country of residence

used_app_before: used the screening app before(yes/no)

result: final score(total number of 1) obtained from A1 to A10

age_desc: string

relation: who is completing the test

class: target label(YES/NO)

Dataset:

704 samples

Data preprocessing

data = pandas.read csv(file path)

A1	A2	А3	A4	A5	A6	A7	A8	A9	A10	age	gender	ethnicity	jaundice	autism	country_of_res	used_app_before	result	age_desc	relation	class
1	1	1	1	0	0	1	1	0	0	26	f	White-European	no	no	'United States'	no	6	18 and more'	Self	NO
1	1	0	1	1	0	1	1	1	1	27	m	Latino	yes	yes	Spain	no	8	'18 and more'	Parent	YES
1	0	0	0	0	0	0	0	0	0	383	f	Pasifika	no	no	'New Zealand'	no	1	'18 and more'	Self	NO

Remove samples with missing value

data.replace('?', numpy.nan); data.dropna(axis = 0, inplace = True)

Remove non-informative attributes

data.drop(['ethnicity'], 1)

						V								
A1	A2	А3	A4	A5	A6	A7	A8	A9	A10	age	gender	jaundice	autism	class
1	1	1	1	0	0	1	1	0	0	26	f	no	no	NO
1	1	0	1	1	0	1	1	1	1	27	m	yes	yes	YES
1	0	0	0	0	0	0	0	0	0	383	f	no	no	NO

Replace abnormal value with average value data.loc[data.age == 383, 'age'] = 30

A1	A2	А3	A4	A5	A6	A7	A8	A9	A10	age	gender	jaundice	autism	class
1	1	1	1	0	0	1	1	0	0	26	f	no	no	NO
1	1	0	1	1	0	1	1	1	1	27	m	yes	yes	YES
1	0	0	0	0	0	0	0	0	0	(30)	f	no	no	NO

Change attributes to numeric

data.replace({'gender' : 'f'}, 1).replace({'gender' : 'm'}, 0)

A1	A2	А3	A4	A5	A6	A7	A8	A9	A10	age	gender	jaundice	autism	class
1	1	1	1	0	0	1	1	0	0	26	1	0	0	0
1	1	0	1	1	0	1	1	1	1	27	0	1	1	1
1	0	0	0	0	0	0	0	0	0	30	1	0	0	0

Scikit-Learn

Read csv file

```
data = pandas.read_csv(file_path)
dataX = data.drop(['class'], 1)
dataY = data['class']
```

Split data into training set and test set

```
x_train, x_test, y_train, y_test = train_test_split(dataX, dataY, test_size = 0.3, random_state = 40)
```

Normalize data(optional)

```
scaler = StandardScaler()
scaler.fit(x_train)
x_train = scaler.transform(x_train)
x_test = scaler.transform(x_test)
```

Parameter optimization with 10-fold cross validation

```
svm = SVC(C=1, kernel = 'linear', degree = 1, cache_size = 200, max_iter = 100000)
scores = cross_val_score(svm, x_train, y_train, cv = 10)
print scores.mean(), scores.std()*2
```

Model fitting and test data prediction

```
svm.fit(x_train, y_train)
prediction = svm.predict(x_test)
accuracy = svm.score(x_test, y_test)
```

Classification Result

S	cikit-Learn	Decision Tree	Naïve Bayes	Logistic Regression	Support Vector Machine	Neural Network	k-NN
	Accuracy	0.92	0.94	1	1	0.99	0.98
	Precision	0.94	0.89	1	1	0.98	0.98
	Recall	0.81	0.93	1	1	1	0.95
	F1-score	0.87	0.91	1	1	0.99	0.97

Ra	apidMiner	Decision Tree	Naïve Bayes	Logistic Regression	Support Vector Machine	Neural Network	k-NN
	Accuracy	0.91	0.95	1	1	1	0.95
	Precision	0.88	0.92	1	1	1	0.89
	Recall	0.81	0.91	1	1	1	0.94
	F1-score	0.84	0.92	1	1	1	0.91

```
RapidMiner
                      A6 = 0: 0 \{0=225, 1=8\}
                          A5 = 0: 0 \{0=14, 1=1\}
                           A5 = 1
                               age > 19.500
                                    A2 = 0
                                        A7 = 0: 0 \{0=7, 1=0\}
                                            A4 = 0: 0 \{0=2, 1=0\}
                                            A4 = 1: 1 \{0=0, 1=3\}
                                            gender = 0: 0 \{0=2, 1=0\}
                                            gender = 1: 1 \{0=0, 1=2\}
                                       A1 = 1: 1 \{0=0, 1=10\}
                               age \leq 19.500: 0 {0=3, 1=0}
                       \top A3 = 0: 0 {0=18, 1=0}
                           A3 = 1
Decision Tree
                                    age > 35: 1 \{0=1, 1=2\}
                                    age \leq 35: 0 {0=9, 1=0}
                                                                   A5, A6, A9, A10
                               A7 = 1: 1 \{0=0, 1=3\}
                       A5 = 1
                               age > 19.500
                                    A8 = 0: 0 \{0=6, 1=1\}
                                        A4 = 0
                                            A3 = 0: 0 \{0=4, 1=0\}
                                                A7 = 0: 0 \{0=1, 1=1\}
                                                A7 = 1: 1 \{0=0, 1=2\}
                                        A4 = 1
                                           A1 = 0: 0 \{0=2, 1=1\}
                                       A1 = 1: 1 \{0=0, 1=16\}
                               age \leq 19.500: 0 {0=2, 1=0}
                               A10 = 0
                                \top A8 = 0: 0 {0=3, 1=1}
                                    A8 = 1
                                       jundice = 0: 1 \{0=0, 1=7\}
                                   | jundice = 1: 0 {0=1, 1=1}
                               A10 = 1: 1 \{0=0, 1=67\}
```

Association Rules

```
[A5, class] --> [A9] (confidence: 0.807)
[class] --> [A9] (confidence: 0.811)
[A6] --> [A10] (confidence: 0.818)
[A6] --> [A5] (confidence: 0.824)
[A10, A9] --> [A5] (confidence: 0.831)
[A5, A9] --> [A10] (confidence: 0.831)
[A5, A9] --> [class] (confidence: 0.831)
[class] --> [A10, A5] (confidence: 0.844)
[class] --> [A10] (confidence: 0.883)
[A5, class] --> [A10] (confidence: 0.889)
[A9, class] --> [A5] (confidence: 0.945)
[class] --> [A5] (confidence: 0.956)
```

Classification result with attributes A5, A6, A9, A10

Sc	ikit-Learn	Decision Tree	Naïve Bayes	Logistic Regression	Support Vector Machine	Neural Network	k-NN
	Accuracy	0.91	0.81	0.9	0.9	0.9	0.87
	Precision	0.91	0.83	0.91	0.91	0.91	0.88
	Recall	0.91	0.81	0.9	0.9	0.9	0.87
	F1-score	0.91	0.81	0.9	0.9	0.9	0.87

Attention to detail

A1: I often notice small sounds when others do not.

A2: I usually concentrate more on the whole picture, rather than the small details.

Attention switching

A3: I find it easy to do more than one thing at once.

A4: If there is an interruption, I can switch back to what I was doing very quickly.

Communication

A5: I find it easy to read between the lines when someone is talking to me.

A6: I know how to tell if someone listening to me is getting bored.

Imagination

A7: When I am reading a story I find it difficult to work out the character's intentions.

A8: I like to collect information about categories of things.

Social interaction

A9: I find it easy to work out what someone is thinking or feeling just by looking at their face.

A10: I find it difficult to work out people's intentions.

Conclusions

With attributes A1, A2, A3, A4, A5, A6, A7, A8, A9, A10, age, gender, jaundice and relative_has_autism, we can predict autism with high accuracy using Decision Tree, Naïve Bayes, Logistic Regression, Support Vector Machine, Feed-forward Neural Network and k-NN.

Models based on A5, A6, A9 and A10 alone produced comparable accuracy to models using more attributes, suggesting that communication and social interaction are two important factors for correctly predicting autism.

Dataset: Autism Screening Adult Data Set

Dataset: Autistic Spectrum Disorder Screening Data for Children Data Set

Accuracy (with attributes A1, A2, A3, A4, A5, A6, A7, A8, A9, A10, age)

		Naïve Bayes	Logistic Regression	Support Vector Machine	Neural Network	k-NN
Scikit-Learn	0.9	0.76	1	1	1	0.9
RapidMiner	0.88	0.94	1	1	1	0.88

Accuracy (with attributes A3, A4, A7, A9, A10)

	Decision Tree	Naïve Bayes	Logistic Regression	Support Vector Machine	Neural Network	k-NN
Scikit-Learn	0.88	0.68	0.88	0.88	0.88	0.88

Attention to detail

Attention switching

[A1: S/he often notices small sounds when others do not.

 $oldsymbol{\mathsf{A2}}$: S/he usually concentrates more on the whole picture, rather than the small details.

A3: In a social group, s/he can easily keep track of several different people's conversations.

A4: S/he finds it easy to go back and forth between different activities.

A5: S/he doesn't know how to keep a conversation going with his/her peers.

A6: S/he is good at social chit-chat.

Imagination

Communication

Social interaction

A7: When s/he is reading a story, s/he finds it difficult to work out the character's intentions or feelings.

A8: When s/he was in preschool, s/he used to enjoy playing games involving pretending with other children.

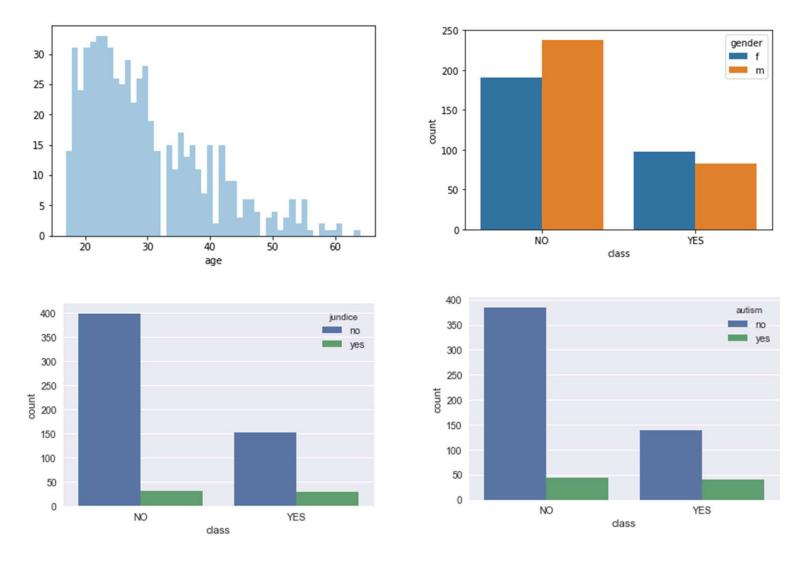
A9: S/he finds it easy to work out what someone is thinking or feeling just by looking at their face.

.A10: S/he finds it hard to make new friends.

Future Work and Challenges

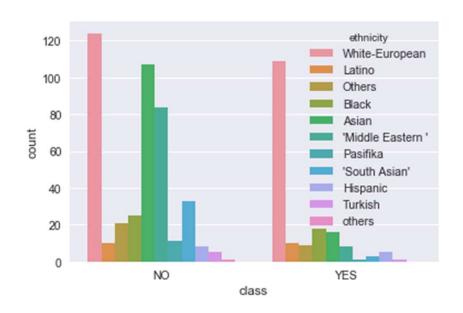
- 1. How well do the models generalize to data from other sources?
- 2. Where to get a larger data set? Finding a larger **clinical** data set would be difficult or might not be possible.
- 3. Data preprocessing was another challenge. How to handle missing data?

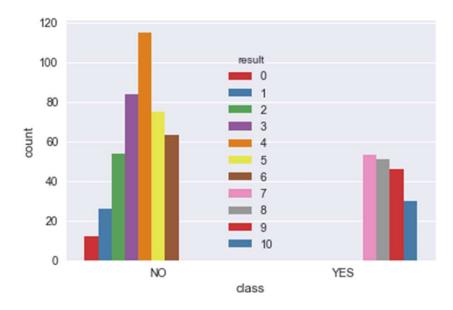
Thank you!



seaborn. distplot(data['age'], bins = 50, kde = False)

seaborn.countplot(x = 'class', hue = 'gender', data)





Scikit-Learn

Decision Tree

dcTree = tree.DecisionTreeClassifier()

Naïve Bayes

NB = MultinomialNB(alpha = 1, class_prior = None, fit_prior = True)

Logistic Regression

LR = LogisticRegression(penalty = 'l1')

Support Vector Machine

svm = SVC(C=1, kernel = 'linear', degree = 1, cache_size = 200, max_iter = 100000)

Feed_forward Neural Network

NN = MLPClassifier(hidden_layer_sizes = (10), max_iter = 1000, learning_rate_init = 0.05, momentum = 0.1)

K-NN

knn = KNeighborsClassifier(n_neighbors = 10)

RapidMiner with attributes A5, A6, A9, A10

Decision Tree

PerformanceVector: accuracy: 84.70% ConfusionMatrix:

True: 0 1 0: 113 12 1: 16 42

Logistic Regression

PerformanceVector: accuracy: 85.79% ConfusionMatrix:

True: 0 1 0: 115 12 1: 14 42

Support Vector Machine

PerformanceVector: accuracy: 85.25% ConfusionMatrix:

True: 0 1 0: 113 11 1: 16 43

```
A9 = 0

| A6 = 0: 0 {0=225, 1=8}

| A6 = 1

| A5 = 0: 0 {0=14, 1=1}

| A5 = 1: 1 {0=14, 1=15}

A9 = 1

| A5 = 0: 0 {0=28, 1=5}

| A5 = 1

| A6 = 0

| A10 = 0: 0 {0=6, 1=2}

| A10 = 1: 1 {0=9, 1=19}

| A6 = 1: 1 {0=4, 1=76}
```