



Classification of Key Risk Factors for Alzheimer's From Questionnaire Data

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Period 4





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01

Project Objective





Alzheimer's Disease

"...a general term for memory loss and other cognitive abilities serious enough to interfere with daily life" (Alzheimer's Association).

- not a normal part of aging
- no known cure
- if no medical breakthrough is made, 13.8 million Americans could have Alzheimer's by 2060 (NIH)

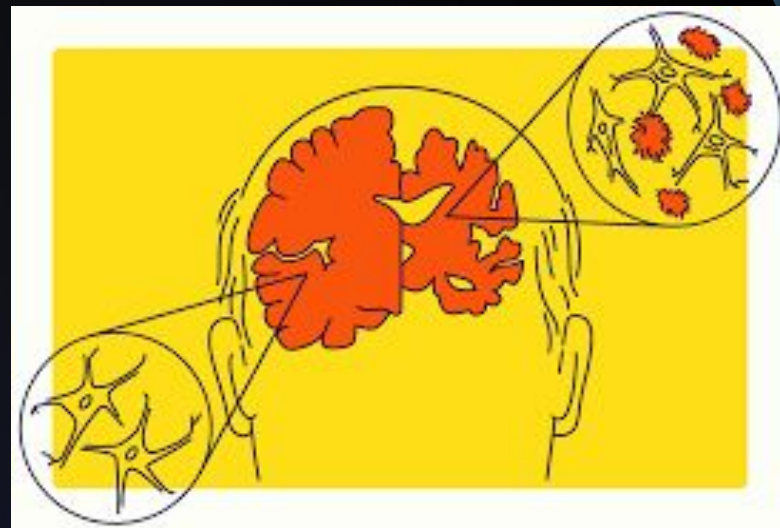






Image source: [Alzheimer's Disease Causes, Symptoms, Treatments](#)



Objective: To categorize results from a telephone questionnaire into six risk factors for Alzheimer's.





02

Background



Datasource

- Collected from questionnaire by the Behavioral Risk Factor Surveillance System
- The 'Question' attribute asked questions regarding health
 - Health-related risk behaviors
 - chronic health conditions
 - health-care access
- 30 attributes; 284142 instances



Overall Classification

- Each instance is an individual
- classifying “Class” attribute based on age, location, datasource, etc
- Name is misleading;
- Not classifying Alzheimer's, instead indicator of healthy aging



Attributes

RowId	Dataset row identifier
YearStart	Year Start
YearEnd	Year End
LocationAbbr	Location Abbreviation
LocationDesc	Location Description
Datasource	Data Source
Class	Class description
Topic	Topic description

Question	Question
Data_Value_Unit	The unit, such as "%" for percentage
DataValueTypeID	Identifier for the Data Value Type
Data_Value_Type	The data value type, such as age-adjusted prevalence or crude prevalence
Data_Value	Data Value, such as 14.7
Data_Value_Alt	Equal to data value, but format is numeric
Data_Value_Footnote_Symbol	Footnote Symbol

Class

Health indicators:

- Screenings and Vaccines 62153
- Nutrition/Physical Activity/Obesity 33194
- Caregiving 25493
- Mental Health 22184
- Smoking and Alcohol Use 22183
- Cognitive Decline 22182





03

Preprocessing



Cleaning Up Redundant Columns



```
['Data_Value_Footnote_Symbol', 'Data_Value_Footnote', 'ClassID',  
'QuestionID', 'TopicID', 'RowId', 'Data_Value_Alt',  
'StratificationCategoryID1', 'StratificationCategoryID2', 'Data_Value_Alt',  
'LocationAbbr', 'LocationID', 'StratificationID1', 'StratificationID2']
```

- Attributes that described other columns in the dataset
- Fourteen columns removed

Processing Strings

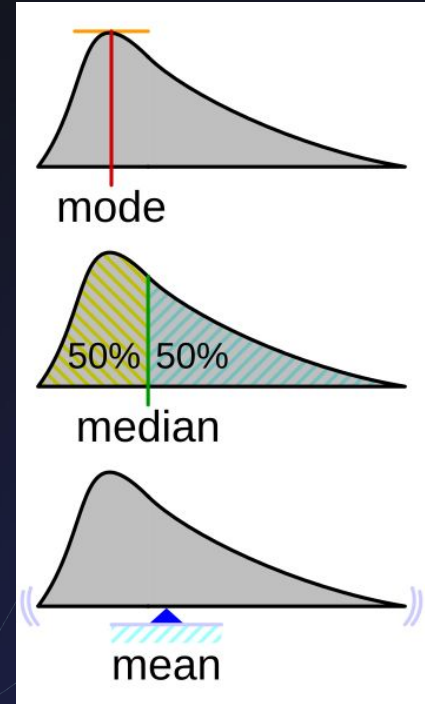
```
df['Stratification2'] = df['Stratification2'].str.replace(',', ' ',  
regex=False)  
df['DataValueTypeID'] = df['DataValueTypeID'].str.replace(',', ' ',  
regex=False)  
df['Question'] = df['Question'].str.replace(',', ' ', regex=False)
```

- Commas, quotation marks, blank attribute names
- Pandas add an extra column

Replacing Nans/Null values

- Quantitative data filled by mean
- Qualitative data filled by mode
- Skipping columns without missing values by using

```
df[column].isnull().sum() == 0:
```



Normalizing Features

- We normalize attributes that are non-categorical
- `['Data_Value', 'Low_Confidence_Limit',
 'High_Confidence_Limit']`
- We use the Z-score normalization from `scipy.stats`

```
df[val] = zscore(df[val])
```

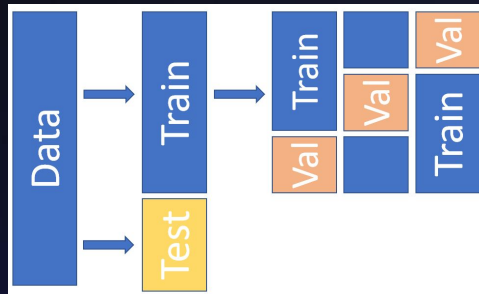
```
id,addr,datasource,class,topic,question,da  
.,0.015822219,0.009009595,0.020437664,1,  
.,0.021894508,-0.031394665,0.069022964,1,  
.,-0.097957964,-0.081032733,-0.113027022  
.,-1.381443874,-1.328727975,-1.403744719  
.,-1.545006359,-1.433643978,-1.640223152  
.,-0.035231781,0.017336395,-0.087968589,  
.,-0.780111209,-0.884083961,-0.620120107  
.,-1.333337261,-1.238799972,-1.413018383  
.,0.105050474,0.125108071,0.08467836,1,3  
.,1.163395965,0.639696087,1.531369951,1,  
.,1.625219452,0.844532094,2.027510977,1,  
.,1.740675323,1.36411611,1.920863841,1,2  
.,1.278851836,1.164276104,1.34589667,1,2  
.,0.716004462,-0.20962394,1.508185791,1,  
.,-1.29485197,-1.17884797,-1.408381551,1  
1,0.225317007,0.015196067,0.432440762,1  
.,0.629412558,0.235020074,0.974950108,1,  
.,1.129721335,1.309160108,0.951765948,1,  
.,0.903620253,0.584740085,1.160423389,1,  
.,1.312526466,1.469032113,1.155786557,1,  
.,0.355204863,0.269992075,0.432440762,1,  
0.470660734 0.195052073 0.724561179 1
```

- , 1, 1, 1, 1, 1, 1
 E, 1, 2, 2, 2, 1,
 , 1, 3, 4, 4, 1, 1
 , 1, 3, 4, 4, 1, 1
 , 1, 3, 4, 4, 1, 1
 , 1, 3, 4, 4, 1, 1
 , 1, 3, 5, 5, 1, 1
 , 1, 3, 5, 5, 1, 1
 , 1, 3, 5, 5, 1, 1
 , 1, 3, 5, 5, 1, 1
 1 3 5 5 1 1

Splitting the Dataset

- We will do a 70%/20%/10% split for the train and test datasets.
- Use sklearn's `train_test_split` to split the data with the `stratify` argument to ensure the ratios of the class are the same.
- The lengths are 198899, 57112, and 28131 respectively. 284,142 total.

```
train_test_split(X_temp, y_temp, stratify=y_temp, test_size=0.33, random_state=42)
```





04

Attribute Selection



Attribute Selection Algorithms Used

Kendall Rank Correlation Coefficient

CorrelationAttributeEval

- Finds correlation based on rank

$$\tau = \frac{\text{Number of concordant pairs} - \text{Number of discordant pairs}}{\frac{n(n-1)}{2}}$$

Concordant Pair: $(x_1, y_1)(x_2, y_2)$ is **concordant** if it satisfies one of the following conditions:

$x_1 > x_2, y_1 > y_2$ and $y_1 > y_2, x_1 > x_2$

$x_1 < x_2, y_1 < y_2$ and $y_1 < y_2, x_1 < x_2$

Discordant Pair: A pair of observations $(x_1, y_1)(x_2, y_2)$ is **discordant** if it satisfies one of the following conditions:

$x_1 > x_2, y_1 < y_2$ and $y_1 < y_2, x_1 < x_2$

$x_1 < x_2, y_1 > y_2$ and $y_1 > y_2, x_1 > x_2$

n: The total number of observations.

Attribute Selection Algorithms Used

Pearson's Correlation

CorrelationAttributeEval

- ranks attributes based on correlation coefficient

$$r = \frac{n(\sum xy) - (\sum x)(\sum y)}{\sqrt{[n\sum x^2 - (\sum x)^2][n\sum y^2 - (\sum y)^2]}}$$

Subset Analysis

CfsSubsetEval

Selects attributes by:

1. grouping attributes as subsets
2. in each subset, evaluate attributes individually
3. choose a subset with high correlation with class but low intercorrelation

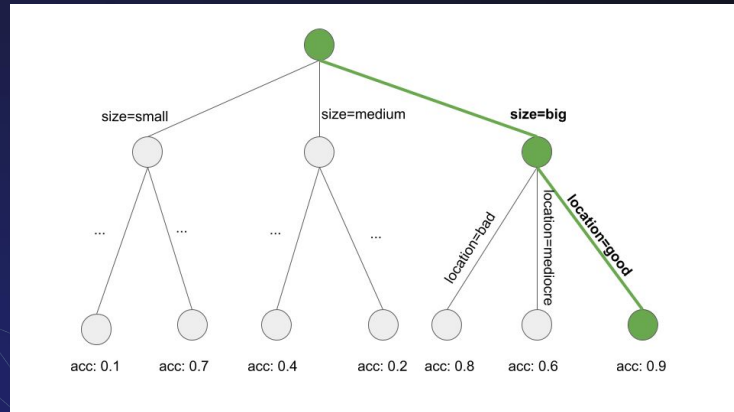
Searches with greedy-stepwise search

Attribute Selection Algorithms Used

Selection by OneR

OneRAttributeEval

- ranks attributes based on classification error from the OneR algorithm



Entropy/Information Gain

InfoGainAttributeEval

- ranks attributes based on information gained with respect to the class

Attributes Selected: Kendall Correlation

```
YearStart SignificanceResult(statistic=-0.008292277413063754, pvalue=9.933366399952563e-09)
YearEnd SignificanceResult(statistic=-0.037674185690728605, pvalue=1.8876229756547483e-149)
LocationDesc SignificanceResult(statistic=0.004162443388828436, pvalue=0.002471294877482304)
Datasource SignificanceResult(statistic=nan, pvalue=nan)
Topic SignificanceResult(statistic=-0.14172617218857977, pvalue=0.0)
Question SignificanceResult(statistic=-0.14172617218857977, pvalue=0.0)
Data_Value_Unit SignificanceResult(statistic=0.03031410117404321, pvalue=1.2063404836511375e-73)
DataValueTypeID SignificanceResult(statistic=0.03031410117404321, pvalue=1.2063404836511375e-73)
Data_Value_Type SignificanceResult(statistic=0.03031410117404321, pvalue=1.2063404836511375e-73)
Data_Value SignificanceResult(statistic=0.21041334613980775, pvalue=0.0)
Low_Confidence_Limit SignificanceResult(statistic=0.23429182517060937, pvalue=0.0)
High_Confidence_Limit SignificanceResult(statistic=0.18575483791844988, pvalue=0.0)
StratificationCategory1 SignificanceResult(statistic=nan, pvalue=nan)
Stratification1 SignificanceResult(statistic=-0.0009736636955514779, pvalue=0.5362844239255413)
StratificationCategory2 SignificanceResult(statistic=-0.03917874396411915, pvalue=9.978459679775013e-122)
Stratification2 SignificanceResult(statistic=-0.02873497235392164, pvalue=4.8475202075739715e-86)
```

0.10 and -0.10
as bounds

Chosen Attributes: Topic, Question, Data_Value,
Low_Confidence_Limit, High_Confidence_Limit

Attributes Selected: Pearson's Correlation

Ranked attributes:

0.22043	12	Low_Confidence_Limit
0.21785	7	Question
0.21785	6	Topic
0.21248	9	DataValueTypeID
0.21248	8	Data_Value_Unit
0.21248	10	Data_Value_Type
0.20816	11	Data_Value
0.18774	13	High_Confidence_Limit
0.02513	18	Geolocation
0.02459	3	YearEnd
0.02426	16	StratificationCategory2
0.01836	2	YearStart
0.01821	17	Stratification2
0.003	4	LocationDesc
0.00213	1	Num
0.00113	15	Stratification1
0	14	StratificationCategory1
0	5	Datasource

cutoff = 0.20

Selected attributes: 12,7,6,9,8,10,11,13,18,3,16,2,17,4,1,15,14,5 : 18



Attributes Selected: Subset Analysis

```
=== Attribute Selection on all input data ===
```

```
Search Method:
```

```
    Greedy Stepwise (forwards).
```

```
    Start set: no attributes
```

```
    Merit of best subset found:    0.767
```

```
Attribute Subset Evaluator (supervised, Class (nominal): 18 Class):
```

```
    CFS Subset Evaluator
```

```
    Including locally predictive attributes
```

```
Selected attributes: 5,16,17 : 3
```

```
    Topic
```

```
    Stratification2
```

```
    Geolocation
```


Attributes Selected: OneR

Ranked attributes:

100	5	Topic
100	6	Question
61.4298	10	Data_Value
61.3478	11	Low_Confidence_Limit
60.9674	12	High_Confidence_Limit
37.7632	17	Geolocation
35.7117	16	Stratification2
34.0509	4	Datasource
34.0509	3	LocationDesc
34.0509	2	YearEnd
34.0509	9	Data_Value_Type
34.0509	7	Data_Value_Unit
34.0509	8	DataValueTypeID
34.0509	15	StratificationCategory2
34.0509	14	Stratification1
34.0509	13	StratificationCategory1
34.0509	1	YearStart

cutoff = 37.0

Selected attributes: 5,6,10,11,12,17,16,4,3,2,9,7,8,15,14,13,1 : 17

Attributes Selected: Information Gain

Ranked attributes:

2.5445	5	Topic
2.5445	6	Question
1.24906	10	Data_Value
1.23631	11	Low_Confidence_Limit
1.22586	12	High_Confidence_Limit
0.14572	17	Geolocation
0.13029	8	DataValueTypeID
0.13029	7	Data_Value_Unit
0.13029	9	Data_Value_Type
0.07083	1	YearStart
0.04949	2	YearEnd
0.03301	16	Stratification2
0.02841	3	LocationDesc
0.00292	15	StratificationCategory2
0	13	StratificationCategory1
0	14	Stratification1
0	4	Datasource

cutoff = 0.13

Selected attributes: 5,6,10,11,12,17,8,7,9,1,2,16,3,15,13,14,4 : 17



05

Results + Analysis



Classification Algorithms Used

OneR

Rule-based method selecting ONE rule based on the attribute with least classification error

Decision Table

Rule-based method that selects MULTIPLE rules based on interactions between multiple attributes

J48

Tree-based method based on the usage of the C4.5 algorithm to generate decision trees

Random Forest

Tree-based method that creates a forest of random trees

Att. Selection Algorithm	Classification Algorithm	Accuracy	MAE	RMSE	Precision	Recall
Kendall's Rank (Non-Weka)	OneR	100%	0	0	1.0	1.0
	Decision Table	100%	0.0002	0.0003	1.0	1.0
	J48	100%	0	0	1.0	1.0
	Random Forest	99.9929%	0	0.0037	1.0 (0.99 in one category)	1.0
Pearson's Correlation	OneR	100%	0	0	1.0	1.0
	Decision Table	100%	0	0	1.0	1.0
	J48	100%	0	0	1.0	1.0
	Random Forest	99.9964%	0.0001	0.0031	1.0	1.0



Att. Selection Algorithm	Classification Algorithm	Accuracy	MAE	RMSE	Precision	Recall
Subset Analysis	OneR	100%	0	0	1.0	1.0
	Decision Table	100%	0.0002	0.0003	1.0	1.0
	J48	100%	0	0	1.0	1.0
	Random Forest	100%	0.0002	0.0011	1.0	1.0
OneR Algorithm	OneR	100%	0	0	1.0	1.0
	Decision Table	100%	0.002	0.003	1.0	1.0
	J48	100%	0	0	1.0	1.0
	Random Forest	99.9929%	0.0001	0.004	1.0 (0.99 in one class)	1.0



Att. Selection Algorithm	Classification Algorithm	Accuracy	MAE	RMSE	Precision	Recall
Information Gain	OneR	100%	0	0	1.0	1.0
	Decision Table	100%	0.0002	0.0003	1.0	1.0
	J48	100%	0	0	1.0	1.0
	Random Forest	99.9929%	0.0001	0.0039	1.0 (0.99) in a class	1.0

Confusion Matrix Comparisons

```
=== Confusion Matrix ===
```

	a	b	c	d	e	f	g	<-- classified as
2196	0	0	0	0	0	0	0	a = Cognitive Decline
0	6154	0	0	0	0	0	0	b = Screenings and Vaccines
0	0	9579	0	0	0	0	0	c = Overall Health
0	0	0	2524	0	0	0	0	d = Caregiving
0	0	0	0	3286	0	0	0	e = Nutrition/Physical Activity/Obesity
0	0	0	0	0	2196	0	0	f = Mental Health
0	0	0	0	0	0	2196	0	g = Smoking and Alcohol Use

A perfect confusion matrix

Subset Analysis with OneR

```
=== Confusion Matrix ===
```

	a	b	c	d	e	f	g	<-- classified as
2196	0	0	0	0	0	0	0	a = Cognitive Decline
0	6152	0	0	0	0	0	2	b = Screenings and Vaccines
0	0	9579	0	0	0	0	0	c = Overall Health
0	0	0	2524	0	0	0	0	d = Caregiving
0	0	0	0	3286	0	0	0	e = Nutrition/Physical Activity/Obesity
0	0	0	0	0	2196	0	0	f = Mental Health
0	0	0	0	0	0	2196	0	g = Smoking and Alcohol Use

Confusion matrix with some errors

Kendall's Correlation with Random Forest

Kendall's Correlation	Pearson's Correlation	Subset Analysis	OneR Algorithm	Entropy/Info Gain
Topic	Low_Confidence_Limit	Topic	Topic	Topic
Question	Question	Stratification2	Question	Question
Data_Value	Topic	Geolocation	Data_Value	Data_Value
Low_Confidence_Limit	Data_Value_Type		Low_Confidence_Limit	Low_Confidence_Limit
High_Confidence_Limit	Data_Value_Unit		High_Confidence_Limit	High_Confidence_Limit
	DataValueTypeD		Geolocation	Geolocation
	Data_Value			DataValueTypeD
				Data_Value_Unit
				Data_Value_Type

Why did most algorithms perform perfectly?

What if we remove 'Topic'?

=== Summary ===

Correctly Classified Instances	28131	100	%
Incorrectly Classified Instances	0	0	%
Kappa statistic	1		
Mean absolute error	0		
Root mean squared error	0		
Relative absolute error	0	%	
Root relative squared error	0	%	
Total Number of Instances	28131		

=== Detailed Accuracy By Class ===

	TP Rate	FP Rate	Precision	Recall	F-Measure	MCC	ROC Area	PRC Area	Class
	1.000	0.000	1.000	1.000	1.000	1.000	1.000	1.000	Cognitive Decline
	1.000	0.000	1.000	1.000	1.000	1.000	1.000	1.000	Screenings and Vaccines
	1.000	0.000	1.000	1.000	1.000	1.000	1.000	1.000	Overall Health
	1.000	0.000	1.000	1.000	1.000	1.000	1.000	1.000	Caregiving
	1.000	0.000	1.000	1.000	1.000	1.000	1.000	1.000	Nutrition/Physical Activity/Obesity
	1.000	0.000	1.000	1.000	1.000	1.000	1.000	1.000	Mental Health
	1.000	0.000	1.000	1.000	1.000	1.000	1.000	1.000	Smoking and Alcohol Use
Weighted Avg.	1.000	0.000	1.000	1.000	1.000	1.000	1.000	1.000	

=== Confusion Matrix ===

	a	b	c	d	e	f	g	<-- classified as
2196	0	0	0	0	0	0	0	a = Cognitive Decline
0	6154	0	0	0	0	0	0	b = Screenings and Vaccines
0	0	9579	0	0	0	0	0	c = Overall Health
0	0	0	2524	0	0	0	0	d = Caregiving
0	0	0	0	3286	0	0	0	e = Nutrition/Physical Activity/Obesity
0	0	0	0	0	2196	0	0	f = Mental Health
0	0	0	0	0	0	2196	0	g = Smoking and Alcohol Use

From the Kendall
Rank-selected
attributes with
OneR:

What if we ALSO remove 'Question'?

=== Summary ===

Correctly Classified Instances	17132	60.9008 %
Incorrectly Classified Instances	10999	39.0992 %
Kappa statistic	0.4982	
Mean absolute error	0.1117	
Root mean squared error	0.3342	
Relative absolute error	49.1059 %	
Root relative squared error	99.1022 %	
Total Number of Instances	28131	

=== Detailed Accuracy By Class ===

	TP Rate	FP Rate	Precision	Recall	F-Measure	MCC	ROC Area	PRC Area	Class
	0.365	0.000	1.000	0.365	0.535	0.588	0.682	0.414	Cognitive Decline
	0.752	0.110	0.658	0.752	0.702	0.613	0.821	0.549	Screenings and Vaccines
	0.688	0.236	0.601	0.688	0.641	0.439	0.726	0.520	Overall Health
	0.342	0.005	0.867	0.342	0.491	0.521	0.669	0.356	Caregiving
	0.540	0.067	0.516	0.540	0.528	0.464	0.737	0.333	Nutrition/Physical Activity/Obesity
	0.617	0.060	0.464	0.617	0.530	0.490	0.779	0.317	Mental Health
	0.509	0.033	0.567	0.509	0.536	0.500	0.738	0.327	Smoking and Alcohol Use
Weighted Avg.	0.609	0.120	0.645	0.609	0.603	0.508	0.745	0.450	


=== Confusion Matrix ===

a	b	c	d	e	f	g	<-- classified as
801	296	503	15	237	207	137	a = Cognitive Decline
0	4629	1149	9	306	42	19	b = Screenings and Vaccines
0	1663	6591	38	794	297	196	c = Overall Health
0	297	698	864	247	295	123	d = Caregiving
0	147	1042	22	1774	217	84	e = Nutrition/Physical Activity/Obesity
0	1	465	24	56	1356	294	f = Mental Health
0	4	523	24	21	507	1117	g = Smoking and Alcohol Use

From the Kendall
Rank-selected
attributes with
OneR:

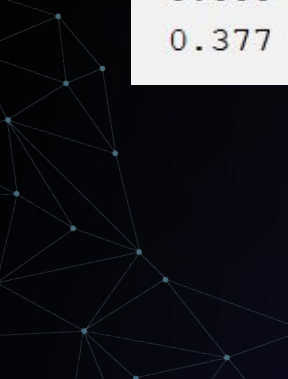
Att. Selection Algorithm	Classification Algorithm	Accuracy	MAE	RMSE	Precision	Recall
Kendall's Rank (Non-Weka)	OneR	60.9002%	0.1117	0.3342	0.645	0.609
	Decision Table	62.2445%	0.1254	0.2505	0.626	0.622
	J48	62.2516%	0.1193	0.2646	0.621	0.623
	Random Forest	61.5655%	0.1153	0.2705	0.615	0.616
Pearson's Correlation	OneR	60.9008%	0.1117	0.3342	0.645	0.609
	Decision Table	63.8904%	0.1203	0.2447	0.649	0.639
	J48	63.318%	0.1172	0.2556	0.632	0.633
	Random Forest	61.5691%	0.1144	0.2683	0.616	0.616


Att. Selection Algorithm	Classification Algorithm	Accuracy	MAE	RMSE	Precision	Recall
Subset Analysis	OneR	37.7022%	0.178	0.4219	?	0.377
	Decision Table	39.085%	0.2152	0.3279	?	0.391
	J48	39.085%	0.2175	0.3287	?	0.391
	Random Forest	39.117%	0.2151	0.3282	?	0.391
OneR Algorithm	OneR	66.0268%	0.1117	0.3342	0.645	0.609
	Decision Table	66.0268%	0.1237	0.243	0.661	0.660
	J48	67.5767%	0.1026	0.2588	0.675	0.676
	Random Forest	68.405%	0.1017	0.2391	0.683	0.684




TP Rate	FP Rate	Precision	Recall	F-Measure	MCC	ROC Area	PRC Area
0.000	0.000	?	0.000	?	?	0.500	0.078
0.114	0.016	0.667	0.114	0.195	0.214	0.549	0.270
0.947	0.844	0.367	0.947	0.529	0.150	0.552	0.366
0.241	0.035	0.406	0.241	0.302	0.262	0.603	0.166
0.000	0.000	?	0.000	?	?	0.500	0.117
0.101	0.024	0.260	0.101	0.146	0.120	0.538	0.096
0.000	0.000	?	0.000	?	?	0.500	0.078
0.377	0.296	?	0.377	?	?	0.541	0.232

Explanation of Subset Analysis





Att. Selection Algorithm	Classification Algorithm	Accuracy	MAE	RMSE	Precision	Recall
Information Gain	OneR	60.9008%	0.1117	0.3342	0.645	0.609
	Decision Table	66.9724%	0.1218	0.24	0.672	0.670
	J48	68.2663%	0.1003	0.2554	0.682	0.683
	Random Forest	68.6716%	0.1008	0.2363	0.686	0.687

The background features a dark blue gradient with decorative geometric patterns. In the top-left corner, there is a network of white dots connected by thin lines, forming a complex, organic shape. In the top-right corner, there is a cluster of white hexagons, some of which contain icons: a gear, a lightbulb, and a magnifying glass with a plus sign. In the bottom-right corner, there is another network of white dots connected by thin lines, similar to the one in the top-left.

We choose **Information Gain**
Attribute Selection with
Random Forest classification
to be our best model.

Steps for Reproducibility:

1. In Weka, load the dataset titled Alzh_train.csv. This dataset automatically includes all preprocessing steps done in Python, as outlined above.
2. Under Select Attributes, under AttributeEvaluator, choose InfoGainAttributeEval. Weka should automatically select the Ranker Search method.
3. On the left side of the screen, select the down arrow by "No class" and select "(Nom) Class" as the class variable. Click start to begin attribute selection.
4. Noting down the attributes with a value of more than 0.13, go to the Preprocess tab and remove all attributes below 0.13 by selecting such attributes and removing them. Do not remove the class. The remaining attributes should be identical to those found in the file Alzh_InfoTrain.arff.
5. Open Alzh_test.csv and remove the same attributes removed in Alzh_train.csv. The resulting dataset should be the same found in Alzh_InfoTest.arff.
6. Furthermore remove 'Topic' and 'Question' from both of the above training and testing sets if you wish to analyze without those two attributes. The resulting datasets should be equivalent to Alzh_InfoTrain2.arff and Alzh_InfoTest2.arff
7. Return to the Alzh_InfoTrain2.arff dataset. Click on the Classify tab, and under Supplied Test Set, choose Alzh_InfoTest2.arff.
8. Repeat step 3, choosing the correct class to be "Class".
9. Select the Random Forest model under rules.
10. Click Start. This model can be found at Info_RandomForest_BestModel.model



06

Conclusion



Reflection

- Understanding class variable
- Stratified sampling in code
- Dealing with qualitative data
- Being mindful of data



Next Steps

- Improve preprocessing.
 - use min-max and decimal scaling to replicate the exact scale.
- Further analysis of performance, model, and feature selector.
 - possible investigation of intercorrelation between attributes





References

Alzheimer's Association. (2023). 2023 Alzheimer's Disease Facts and Figures. *Alzheimer's & Dementia*, 19(4), 1598–1695. <https://doi.org/10.1002/alz.13016>

CDC Division of Population Health. (2024). *Alzheimer's Disease and Healthy Aging Data | Data | Centers for Disease Control and Prevention*. [Chronicdata.cdc.gov](https://chronicdata.cdc.gov).

https://chronicdata.cdc.gov/Healthy-Aging/Alzheimer-s-Disease-and-Healthy-Aging-Data/hfr9-rurv/about_data

CfsSubsetEval. (n.d.). [Weka.sourceforge.io](https://weka.sourceforge.io).

<https://weka.sourceforge.io/doc.dev/weka/attributeSelection/CfsSubsetEval.html>

GeeksForGeeks. (2020, May 15). *Python | Kendall Rank Correlation Coefficient*. [GeeksforGeeks](https://www.geeksforgeeks.org/python-kendall-rank-correlation-coefficient/).

<https://www.geeksforgeeks.org/python-kendall-rank-correlation-coefficient/>

GreedyStepwise. (2022, January 28). [Sourceforge.io](https://weka.sourceforge.io).

<https://weka.sourceforge.io/doc.dev/weka/attributeSelection/GreedyStepwise.html>

<https://www.health.com/condition/alzheimers-overview>. (2022, August 15). *What Is Alzheimer's Disease?*