Feature replacement methods enable reliable home video analysis for machine learning detection of autism

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SUPPLEMENTARY FILE

Additional Information

Classic feature imputation techniques: Multivariate

The first multivariate feature imputation method we selected is **ridge regression**. Ridge regression predicts the target value (here, the missing value) based on the other available features in the model's dataset. It includes a regularization parameter λ in the estimation procedure that reduces the risk of overfitting on the training set.

Gaussian mixtures assume that data points are generated from a mixture of K Gaussian distributions, i.e. clusters, each with unknown parameters (mean and standard deviation). Based on an estimation of these parameters, we are able to predict the missing values in our dataset. Our Gaussian mixture object implements the expectation-maximization (EM) algorithm for parameter estimation. The EM algorithm iterates between two modes: 1) the expectation step (E) fixes the Gaussian model parameters and computes the conditional probability of each data point according to these parameters and 2) the maximisation step (M) computes the Gaussian model parameters that maximize the probabilities found in E. Once clusters of data are learned, examples with missing entries can be identified as parts of a given cluster and completed with the most likely values given the non-missing entries for this cluster. We opted for this method since the significant difference in the number of missing values between ASD and NT in part "Results - Dataset analysis" implies that our data may live on separate clusters that unsupervised learning methods can discover.

Finally, **decision trees** are a non-parametric supervised learning method aiming to learn simple decision rules for prediction inferred from the data. Decision tree is another common predictive model used to impute missing values using approaches such as the *missForest*¹ technique that builds a random forest model for each variable.

Mathematical Formulation

With X, a dataset of n features and k records, let us note $x_{i,j}$ as the value of the j th feature for the i th record in X. We call Y the target binary vector containing the ASD diagnosis of all k records in X (0 corresponding to NT and 1 to ASD). In the current context, X corresponds to our ADOS or ADI-R training dataset and Y to the instrument-level ASD diagnosis. If X contains a missing value in position (i, j), we call $x_{i,j}^{\text{nomiss}}$ the actual value that should be in $x_{i,j}$ and $\hat{x}_{i,j}^{\text{nomiss}}$ our estimation of $x_{i,j}^{\text{nomiss}}$ through f_{imp} , the feature imputation method. We are looking for f_{imp} such that we minimize the error between X^{nomiss} (version of X containing all $x_{i,j}^{\text{nomiss}}$) and \hat{X}^{nomiss} (our estimation of X^{nomiss} containing all $\hat{x}_{i,j}^{\text{nomiss}}$) and \hat{X}^{nomiss} (our estimation of X nomiss containing all $\hat{x}_{i,j}^{\text{nomiss}}$). However, since we do not have access to X^{nomiss} in practice (we do not know the "correct" value to fill NAs with) and since our main goal is the final accuracy of the model, we look for f_{imp} such that we maximize the unweighted average recall (UAR) between Y and \hat{Y} , our prediction of the ASD diagnosis from the model h (either LR9 and ADTree7) trained on \hat{X}^{nomiss} . We then test this process on Z, the video ratings containing missing values and for which we want to predict the ASD diagnosis. Similarly, if Z contains a missing value in feature j, we call z_j^{nomiss} the actual missing value and $\hat{z}_j^{\text{nomiss}}$ our estimation of z_j^{nomiss} through f_{imp} , feature

imputation method. To summarize, we look for f_{imp}^* such that:

$$f_{\text{imp}}^* = \underset{f_{\text{imp}}}{\arg\min}(\|X^{\text{nomiss}} - \hat{X}^{\text{nomiss}}\|) = \underset{f_{\text{imp}}}{\arg\min}(\|X^{\text{nomiss}} - f_{\text{imp}}(X)\|)$$

However, we do not have access to X^{nomiss} , so we use:

$$\begin{split} f^*_{\text{imp}} &= \underset{f_{\text{imp}}}{\arg\max}(UAR(Y,\hat{Y})) \\ &= \underset{f_{\text{imp}}}{\arg\max}(UAR(Y,h(\hat{X}^{\text{nomiss}}))) \\ &= \underset{f_{\text{imp}}}{\arg\max}(UAR(Y,h(f_{\text{imp}}(X)))) \end{split}$$

with:

- UAR is unweighted average recall
- X is the training dataset containing NULL values and Y is the associated ASD diagnosis
- X^{nomiss} is the theoretical version of X containing "correct" missing values
- \hat{X}^{nomiss} is our estimation of X^{nomiss} using feature imputation method f_{imp}
- h is the ASD classifier, either LR9 or ADTree7, with a MinMaxScaler

We then test on a record Z containing missing values by establishing an ASD prediction using both f_{imp}^* and h:

ASD prediction for $Z = h(f_{imp}^*(Z))$

General feature replacement methods

Let us consider Z a new test record for which we wish to predict ASD class and that has a missing value in feature j. Our estimation of $\hat{z}_j^{\text{nomiss}}$ will be z_{j*} if feature j^* is the closest feature to j as per score s in the training set X. Using mutual information (MI), for example, the replaced feature would be:

$$j^* = \underset{l \in [\text{ all } n^* \text{ features available }]}{\operatorname{arg max}} (MI(X_j, X_l))$$

Dynamic feature replacement

Let us consider Z a new test record for which we wish to predict ASD class and that has a missing value in feature j. Our estimation of $\hat{z}_j^{\text{nomiss}}$ will be z_{j*} if feature j^* is the closest feature to j as per score s in the <u>subset</u> of the training set \tilde{X} . Using mutual information (MI), for example, the predicted feature would be:

$$j^* = \operatorname*{arg\,max}_{l \in [\text{ all } n^* \text{ features available }]} (MI(X_j, X_l))$$

 \tilde{X} is defined as all records in X with a similar rating of Z, i.e. for all features l in Z, $x_{i,l}$ is in a -1/+1 range of z_l based on the ADOS or ADI-R questions' ordinal scale.

 $\tilde{X} = \{ \text{all records } i \text{ in } X \text{ such that } \forall l, x_{i,l} - 1 \le z_l \le x_{i,l} + 1 \}$

As defined above, \tilde{X} 's size varies the number of ratings in the training set similar to Z.

References

 Stekhoven, D. J. & Bühlmann, P. Missforest—non-parametric missing value imputation for mixed-type data. *Bioinformatics* 28, 112–118 (2012).

Figures and Tables

Diagnosis / Gender	Female	Male	Diagnosis / Gender	Female M	Male
ASD	594	2,883	ASD	1,897 9	,409
NT	201	419	NT	191 2	241
		1 4 4	(b) ADID 2002 AD		

(a) ADOS Module 2 - LR9 training dataset

(b) ADI-R 2003 - ADTree7 training dataset

Table 1. Gender and Diagnosis in training datasets

Age / Diagnosis	<1	1-3	4-6	7-10	11-15	16-18	>18	NULL
ASD	172	452	1,439	698	229	31	33	610
NT	124	331	184	29	8	1	1	1

(a) ADOS Module 2 - LR9 training dataset								
Age / Diagnosis	<1	1-3	4-6	7-10	11-15	16-18	>18	NULL
ASD	13	1,010	2,530	2,567	1,659	403	343	2,896
NT	0	70	146	119	81	10	9	1

(b) ADI-R 2003 - ADTree7 training dataset

Table 2. Age Group and Diagnosis in training datasets (age in years)

Diagnosis / Gender	Female	Male
ASD	26	44
NT	32	38

(a) Gender and Diagnosis

Diagnosis / Age	<1	1-3	4-6	7-10	11-15	16-18	>18	NULL
ASD	0	32	34	4	0	0	0	0
NT	0	43	26	1	0	0	0	0

(b) Age Group and Diagnosis (age in years)

 Table 3. YouTube testing dataset description

Rated feature	Used in LR9	Used in ADTree7
Echolalia		
Expressive Language	B10 Amount of reciprocal social communication	35 - conver5 Reciprocal conversation of simple language: answer most abnormal between 4 and 5
Speech Patterns		
Communicative Engagement		
Aggression		
Entertains Self		
Understands Language		29 - compsl5 Comprehension of simple language: answer most abnormal between 4 and 5
Eye Contact	B01 Unusual eye contact	50 - gaze5 Direct gaze: answer most abnormal between 4 and 5
Responsiveness		

Developmental Delay		86 - ageabn
		Age when abnormality first evident
Comforts Others		
Social Participation		64 - grplay5 Group play with peers: answer most abnormal between 4 and 5
Sensory Aversion		
Imitates Actions		
Emotion Expression		
Sensory Seeking		
Pretend Play		48 - play5 Imaginative play: answer most abnormal between 4 and 5
Shakes Head YesNo		
Responsive Social Smile		
Calls Attention to Objects		
Joint Attention Pointing	B06 Spontaneous initiation of joint attention	
Appropriate Play		
Creativity		
Stereotyped Speech	A05 Stereotyped-idiosyncratic use of words or phases	
Spontaneous Gestures	A08 Descriptive conventional instrumental or informative gestures	
Indicates Pleasure to Others	B03 Shared enjoyment in interaction	49 - peerpl5 Imaginative play with peers: answer most abnormal between 4 and 5
Social Overtures	B08 Quality of social overtures	
Complex Mannerisms	D02 Hand and finger and other complex mannerisms	
Stereotyped Interests Actions	D04 Unusually repetitive interests or stereotyped behaviors	

Table 4. 30 features scored by video raters and, if they are features of the LR9 or ADTree7 algorithms, their mapped ADOS and ADI-R features

LR9 Feature	ADOS Module 2 Replacement feature
Rated feature used	Rated Replacement feature used
B10 - Amount of reciprocal social communication	A02 - Amount of social overtures / maintenance of attention
Rated feature used: Expressive Language	Rated feature used: Social Overtures
D02 - Hand and finger and other complex mannerisms	D01 - Unusual sensory interest in play material / person.
Rated feature used: Complex mannerisms	Rated feature used: Sensory Seeking
A08 - Descriptive conventional instrumental or informative gestures Rated feature used: Spontaneous gestures	B10 - Amount of reciprocal social communication <i>Rated feature used:</i>
B08 - Quality of social overtures	A02 - Amount of social overtures / maintenance of attention
Rated feature used: Social overtures	Rated feature used: Social Overtures
D04 - Unusually repetitive interests or stereotyped behaviors Rated feature used: Stereotyped Interests Actions	D01 - Unusual sensory interest in play material / person. Rated feature used: Sensory Seeking
B03 - Shared enjoyment in interaction	A02 - Amount of social overtures / maintenance of attention
Rated feature used: Indicates Pleasure to Others	Rated feature used: Social Overtures
B06 - Spontaneous initiation of joint attention	A07 - Pointing
Rated feature used: Joint Attention Pointing	Rated feature used: Calls Attention to Objects
A05 - Stereotyped/idiosyncratic use of words or phases	B09 - Quality of social response
Rated feature used: Stereotyped Speech	Rated feature used: Responsiveness
B01 - Unusual eye contact	B08 - Quality social overtures
Rated feature used: Eye contact	Rated feature used: Social Overtures

 Table 5. LR9 Features and replacements - correlation based selection

ADTree7 Feature	ADI-R 2003 Replacement feature
Rated feature used	Rated Replacement feature used
86 - Age when abnormality first evident Rated feature used: Developmental Delay	69.2 - Ever repetitive use of objects or interests in parts of objects Rated feature used: Stereotyped Interests Actions
29 - Comprehension of simple language: answer most abnormal between 4 and 5 <i>Rated feature used: Understands Language</i>	41.2 - At 5 current communicative speech <i>Rated feature used: Expressive Language</i>
48 - Imaginative play:	49 - Imaginative play with peers:
answer most abnormal between 4 and 5	answer most abnormal between 4 and 5
Rated feature used: Pretend Play	Rated feature used: Indicates Pleasure to Others
49 - Imaginative play with peers:	48 - Imaginative play:
answer most abnormal between 4 and 5	answer most abnormal between 4 and 5
Rated feature used: Indicates Pleasure to Others	Rated feature used: Pretend Play
64 - Group play with peers: answer most abnormal between 4 and 5 <i>Rated feature used: Social Participation</i>	63.2 - At 4-5 response to approaches of other children Rated feature used: Comforts Others
35 - Reciprocal conversation of simple language: answer most abnormal between 4 and 5 <i>Rated feature used: Expressive Language</i>	34.2 - Ever social verbalization / chat Rated feature used: Understands Language
50 - Direct gaze: answer most abnormal between 4 and 5 <i>Rated feature used: Eye contact</i>	56.2 - At 4-5 quality of social overtures <i>Rated feature used: Social Overtures</i>

 Table 6. ADTree7 Features and replacements - correlation based selection

LR9 Feature	ADOS M2 Replacement feature
Rated feature used	Rated Replacement feature used
B10 - Amount of reciprocal social communication	B09 - Quality of social response
Rated feature used: Expressive Language	Rated feature used: Social Participation
D02 - Hand and finger and other complex mannerisms	D01 - Unusual sensory interest in play material/person
Rated feature used: Complex mannerisms	Rated feature used: Sensory Seeking
A08 - Descriptive conventional instrumental or informative gestures Rated feature used: Spontaneous gestures	B10 - Amount reciprocal social communication Rated feature used: Expressive Language
B08 - Quality of social overtures	B09 - Quality of social response.
Rated feature used: Social overtures	Rated feature used: Social Participation
D04 - Unusually repetitive interests or stereotyped behaviors Rated feature used: Stereotyped Interests Actions	D01 - Unusual sensory interest in play material/person. Rated feature used: Sensory Seeking
B03 - Shared enjoyment in interaction	A02 - Amount of social overtures/maintenance of attention.
Rated feature used: Indicates Pleasure to Others	Rated feature used: Responsiveness
B06 - Spontaneous initiation of joint attention	A02 - Amount of social overtures/maintenance of attention.
Rated feature used: Joint Attention Pointing	Rated feature used: Responsiveness
A05 - Stereotyped/idiosyncratic use of words or phases	B09 - Quality of social response.
Rated feature used: Stereotyped Speech	Rated feature used: Social Participation
B01 - Unusual eye contact	A06 - Conversation.
Rated feature used: Eye contact	Rated feature used: Understands Language

 Table 7. LR9 Features and replacements - nearest neighbor selection

ADTree7 Feature	ADI-R 2003 Replacement feature
Rated feature used	Rated Replacement feature used
86 - Age when abnormality first evident Rated feature used: Developmental Delay	35 - Reciprocal conversation of simple language: answer most abnormal between 4 and 5 <i>Rated feature used: Expressive Language</i>
29 - Comprehension of simple language: answer most abnormal between 4 and 5 <i>Rated feature used: Understands Language</i>	54.2 - At 4-5 seeking to share his/her enjoyment with others Rated feature used: Shares Excitement
48 - Imaginative play:	49 - Imaginative play with peers:
answer most abnormal between 4 and 5	answer most abnormal between 4 and 5
Rated feature used: Pretend Play	Rated feature used: Indicates Pleasure to Others
49 - Imaginative play with peers:	48 - Imaginative play:
answer most abnormal between 4 and 5	answer most abnormal between 4 and 5
Rated feature used: Indicates Pleasure to Others	Rated feature used: Pretend Play
64 - Group play with peers:	49 - Imaginative play with peers:
answer most abnormal between 4 and 5	answer most abnormal between 4 and 5
Rated feature used: Social Participation	Rated feature used: Indicates Pleasure to Others
35 - Reciprocal conversation of simple language: answer most abnormal between 4 and 5 <i>Rated feature used: Expressive Language</i>	34.2 - Ever social verbalization/chat Rated feature used: Understands Language
50 - Direct gaze: answer most abnormal between 4 and 5 <i>Rated feature used: Eye contact</i>	63.2 - At 4-5 response to approaches of other children <i>Rated feature used: Social Participation</i>

Table 8. ADTree7 Features and replacements - nearest neighbor selection

LR9 Feature	ADOS M2 Replacement feature
Rated feature used	Rated Replacement feature used
B10 - Amount of reciprocal social communication	A02 - Amount of social overtures/maintenance of attention
Rated feature used: Expressive Language	Rated feature used: Social Overtures
D02 - Hand and finger and other complex mannerisms	D01 - Unusual sensory interest in play material/person
Rated feature used: Complex mannerisms	Rated feature used: Sensory Seeking
A08 - Descriptive conventional instrumental or informative gestures <i>Rated feature used: Spontaneous gestures</i>	A06 - Conversation Rated feature used: Communicative Engagement
B08 - Quality of social overtures	A02 - Amount of social overtures/maintenance of attention
Rated feature used: Social overtures	Rated feature used: Social Overtures
D04 - Unusually repetitive interests or stereotyped behaviors Rated feature used: Stereotyped Interests Actions	D01 - Unusual sensory interest in play material/person Rated feature used: Sensory Seeking
B03 - Shared enjoyment in interaction	A02 - Amount of social overtures/maintenance of attention
Rated feature used: Indicates Pleasure to Others	Rated feature used: Social Overtures
B06 - Spontaneous initiation of joint attention	A07 - Pointing
Rated feature used: Joint Attention Pointing	Rated feature used: Calls Attention to Objects
A05 - Stereotyped/idiosyncratic use of words or phases	B09 - Quality of social response
Rated feature used: Stereotyped Speech	Rated feature used: Responsiveness
B01 - Unusual eye contact	B09 - Quality of social response
Rated feature used: Eye contact	Rated feature used: Responsiveness

 Table 9. LR9 Features and replacements - mutual information selection

ADTree7 Feature	ADI-R 2003 Replacement feature
Rated feature used	Rated Replacement feature used
86 - Age when abnormality first evident	04a - Coded: Onset as perceived with hindsight
Rated feature used: Developmental Delay	Rated feature used: Developmental Delay
29 - Comprehension of simple language:	41.2 - At 5 current communicative speech Rated feature used: Communicative Engagement
answer most abnormal between 4 and 5	
Rated feature used: Understands Language	
48 - Imaginative play:	47.2 - At 4-5 spontaneous imitation of actions Rated feature used: Imitates Actions
answer most abnormal between 4 and 5	
Rated feature used: Pretend Play	
49 - Imaginative play with peers:	56.2 - At 4-5 response to approaches of other children Rated feature used: Social Overtures
answer most abnormal between 4 and 5	
Rated feature used: Indicates Pleasure to Others	
64 - Group play with peers:	63.2 - At 4-5 response to approaches of other children <i>Rated feature used: Social Participation</i>
answer most abnormal between 4 and 5	
Rated feature used: Social Participation	
35 - Reciprocal conversation of simple language:	34.2 - Ever social verbalization/chat Rated feature used: Understands Language
answer most abnormal between 4 and 5	
Rated feature used: Expressive Language	
50 - Direct gaze:	56.2 - At 4-5 quality of social overtures <i>Rated feature used: Social Overtures</i>
answer most abnormal between 4 and 5	
Rated feature used: Eye contact	

Table 10. ADTree7 Features and replacements - mutual information selection



Figure 1. Percentage of missing values per model feature in training datasets





(b) ADTree7 - YouTube ratings testing dataset

Figure 2. Percentage of missing values per model feature in testing datasets