

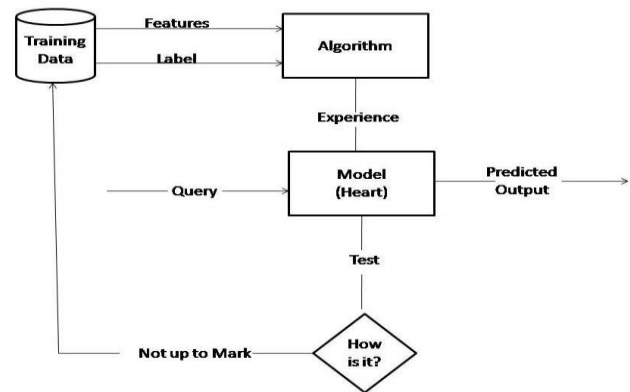
Classification of Spinal Muscle Atrophy Disease using SVM in Machine Learning



B. Ganga Bhavani, G. L. N. V. S. Kumar, M. L. Rekha, B. P. N. Madhu Kumar, Raja Rao P. B. V.

Abstract: SMA is a genetic neuromuscular disease. It is a rare disease. It is caused by mutations in the survival motorneuron (SMN) gene that encodes SMN Protein. Main difficult area of SMA is muscle weakness, causing with difficulty with moving, swallowing or breathing. There are four types of SMA's. The primary objective of this paper is to classify the SMA's by using support vector machine classifier. Then we can easily predict the life span of the children based on the group of SMA. This disease is classified on the basis of age of onset and clinical course.

Keywords: SMN1, SMN2, SVM, SVC, CPK, SMA Linear, RBF, Polynomial.

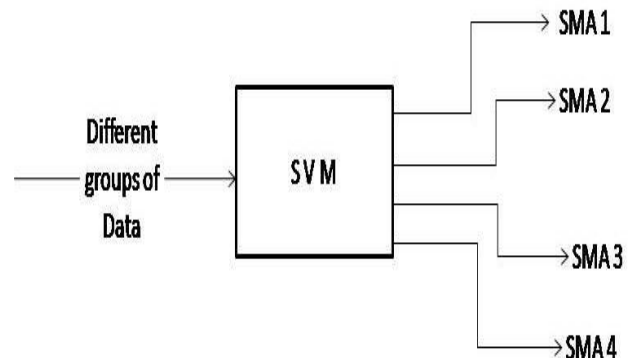


I. INTRODUCTION

The task of constructing groups of SMAs can be categorized into subtasks. There are four different types of SMAs. To classify the groups of SMAs we use Support vector machine algorithm. Support vector machine algorithm is purely supervised machine learning algorithm. In this first we take the training data, then trains the algorithm with proper features and labels then we get important output call it as model. This model is the heart of the machine. This model is trained by the experienced algorithm. We just pass a query to this model. It produces a predicted output. In some cases, the model is not up to mark we can again retrain the model. This process is continuous of testing and training.

This paper here focusses on the different groups of SMA's Such as SMA0, SMA1, SMA2, SMA3, SMA4 using support vector machine algorithm. SMA0: The most severe form. Children usually succumb to the disease before the age of 6 months.

- SMA1: Symptoms appear in the first few months of life. Children rarely survive passed their second birthday
- SMA2: Serious muscle weakness symptoms usually appears between 7 to 18 months of age.
- SMA3: children are able to stand and walk, but worsen with time. Symptoms usually appear after 18 months of age.
- SMA4: Not a lifetime threatening condition. Symptoms appear in adulthood.



II. SVM ALGORITHM

2.1 The Basis of SVM Algorithm:

Fundamentally, SVM is used for both classification and regression problems. Support vector machine is a supervised classification method that separates data using hyperplanes. The distance between the support vectors and hyperplanes is maximum. So margin is always maximum. There are two types of SVM's. Linear SVM and Nonlinear SVM. We can separate the data into two groups in linear SVM by using straight line. In non linear SVM we can separate the data into two groups by using 3D. Steps in Supervised learning algorithm:

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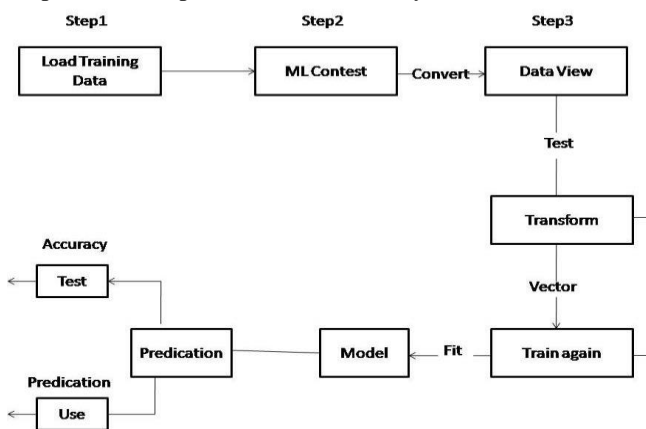
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- Step1: Load the training data
- Step2: Prepare ML context
- Step3: Convert the ML context into Dataview
- Step4: transform the text into vector form
- Step5: prepare the training algorithm
- Step6: fit that training and testing the data into the MODEL.
- Step7: make the prediction for accuracy



2.2 Support vector machine algorithm:

SVM is used to classify data by a hyper plane that linearly separates the data from different classes.

- Step 1: labelled the sample data
- Step 2: hyper plane can be represented by $\langle w, x \rangle + b = 0$ where w represents weights
- Step 3: dimensional vectors is defined as $\langle x, y \rangle = x_1y_1 + x_2y_2 + \dots + x_ny_n$
- For 2D Plane $\langle w, x \rangle = w_1x_1 + w_2x_2$
- Hyperplane is $w_1x_1 + w_2x_2 + b = 0$
- Step 4: For Linear SVM $Y_i(\langle x_i, w \rangle + b) - 1 \geq 0$
- The points closest to hyperplane are called support vectors $\text{Min} 1/2 \|w\|^2$ such that $y_i(x_i \cdot w + b) - 1 \geq 0$
- Step 5: For nonlinear SVM
- According to soft-margin classification
- Step 6: Apply kernel function
- According to Linear
- According to RBF $K(x, x_1) = \exp(-\|x_1 - x\|^2 / 2\mu^2)$
- According to Polynomial kernel $K(x, x_1) = (x \cdot x_1 + c)^d$

III. CLASSIFICATION OF SMA DISEASE

SMA is a spinal muscle atrophy. the main goal of applying support vector machine algorithm is to classify the groups of SMA's to survive the life time of human beings.

Dataset:

SMA calculations of the performance of SVM algorithm are carried out on 50 real world data sets drawn from the Rainbow hospital, NIMBS hospital, neuro muscular disorder university of california at data repository []. The datasets are chosen vary across a number of dimensions including the type of the blood samples and the number of instances. These are datasets with both discrete and continuous data. After loading the datasets into our model the graphical representation of our data is as shown in table 1

Table 1: Loading the dataset

	type	AGE	cpk	GMFM	HFMS	FVC	disease
0	SMA1	0.1	250	22	8	10	1
1	SMA1	0.2	500	12	14	12	1
2	SMA1	0.5	243	14	18	14	1
3	SMA1	0.4	287	15	17	10	1
4	SMA1	0.3	296	17	2	12	1

Statistics:

For the disease SMA following attributes are considered in the calculations which plays an important role for the classification of SMA.

- CPK: Creatine phosphokinase (based on muscle biopsy)
- GMFM: gross motor function measure
- HFMS: Hamersmith functional motor scale
- FVC: forced vital capacity

After applying statistical method on the datasets the results are shown in table 2

Table 2: Attribute values after applying statistical methods

	AGE	cpk	GMFM	HFMS	FVC	disease
count	59.000000	59.000000	59.000000	59.000000	59.000000	59.000000
mean	11.218644	1926.135593	23.966102	33.152542	29.067797	2.677966
std	13.062495	1784.400000	15.963283	19.941132	12.455250	1.209740
min	0.100000	224.000000	10.000000	2.000000	10.000000	1.000000
25%	0.750000	624.500000	13.500000	16.500000	17.500000	2.000000
50%	4.000000	1223.000000	18.000000	28.000000	31.000000	3.000000
75%	23.000000	2555.500000	26.000000	46.500000	40.000000	4.000000
max	50.000000	6047.000000	84.000000	84.000000	49.000000	4.000000

Graphical Representation:

After loading the datasets into our model the graphical representation of our data is as shown in fig 1

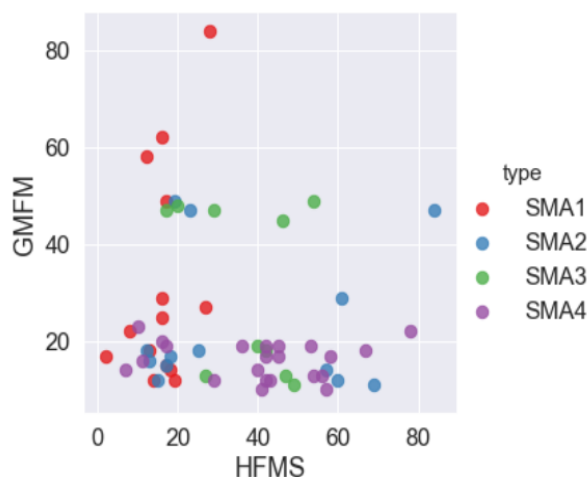


Fig 1: load the data

Fit the Model:

Here we apply the SVM (Support vector machine) to fit the model and apply the SVC (Support vector classifier) to classify the data.



After applying support vector machine the hyperplane divides the data into two parts as shown in fig2.

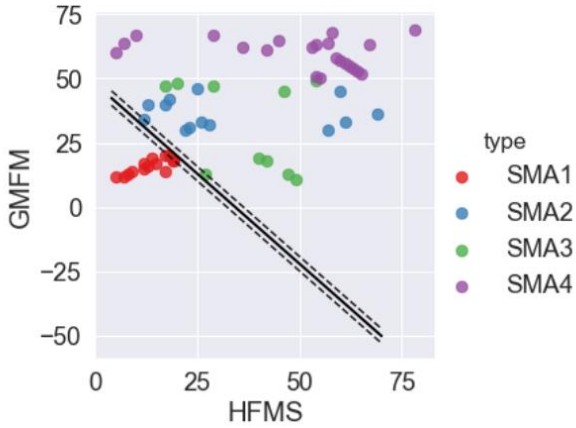


Fig2: Hyperplane divides the data into two sets.

The points which are nearest to dashed line indicates support vectors. The distance between the two hyperplanes indicates the margin.

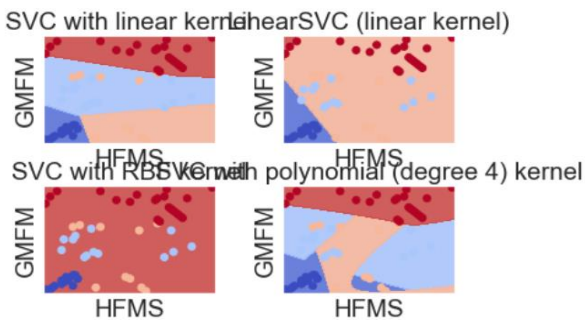
The darked line indicates decision boundary.

The points which are below the line comes under SMA1. The points which are above the line comes under SMA2, SMA3 and SMA4.

Here SVC (Support vector classifier) classify the data into two parts. For multiclassification we use kernel functions. Applying Kernel functions:

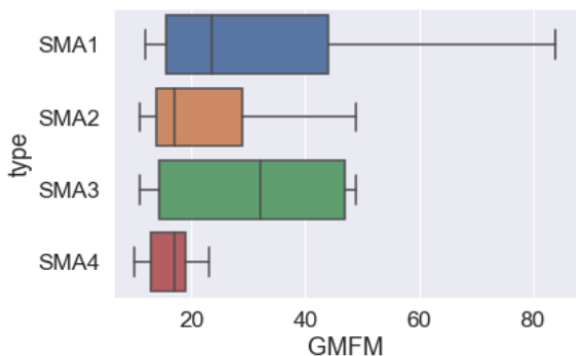
1. SVC with linear kernel
2. linear svc
3. svc with RBF
4. SVC with polynomial kernel

Among all those 4 types of kernel functions SVC with polynomial is the best kernel function to classify the types of SMA's disease

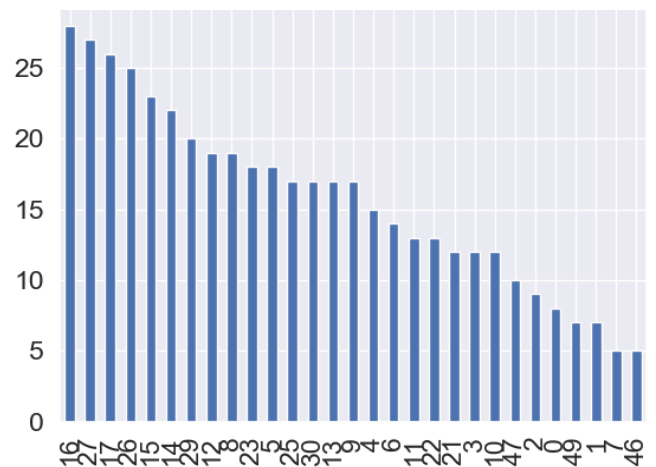
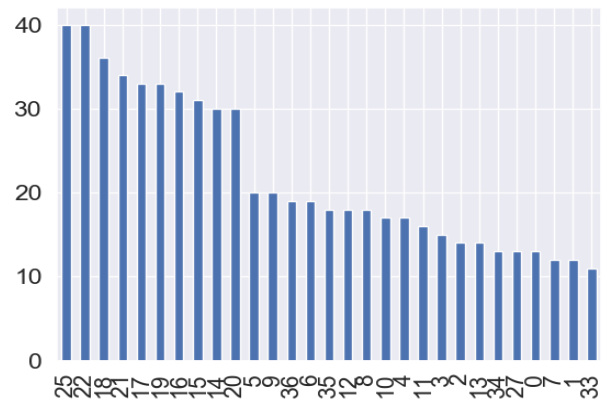
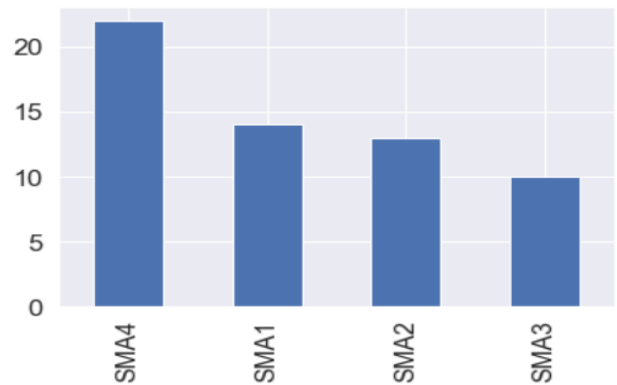


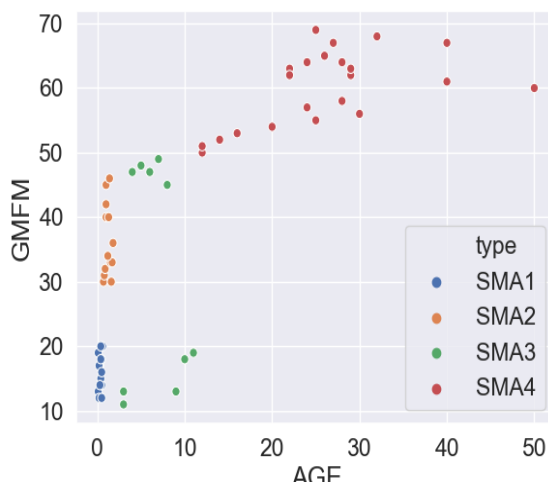
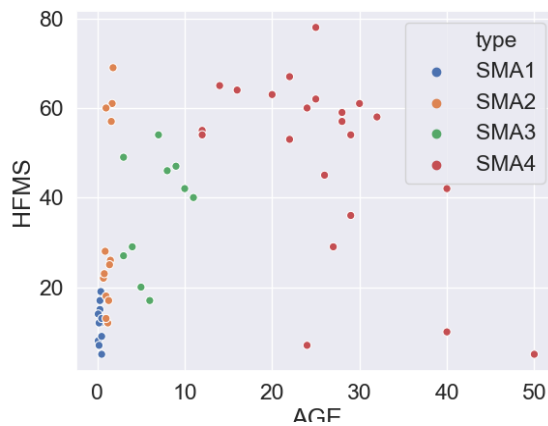
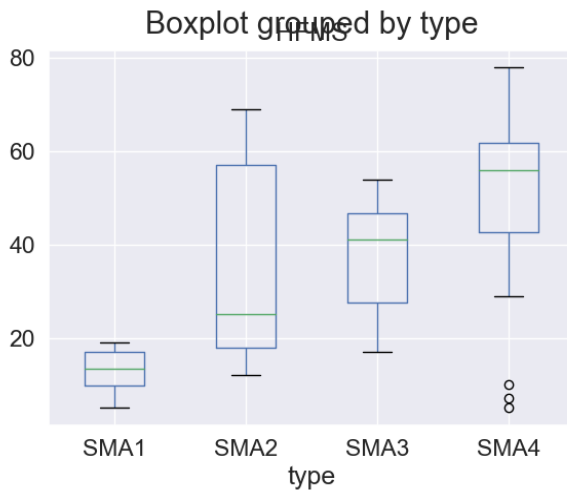
Experimental results:

Graphical representation



AGE	1	0.84	0.8	0.37	0.67	0.83
cpk	0.84	1	0.81	0.43	0.6	0.8
GMFM	0.8	0.81	1	0.5	0.65	0.85
HFMS	0.37	0.43	0.5	1	0.6	0.63
FVC	0.67	0.6	0.65	0.6	1	0.87
disease	0.83	0.8	0.85	0.63	0.87	1
	AGE	cpk	IFM	MS	FVC	ase





Prediction calculation by using Gaussian NB:

- Step1:Load dataset
- Step2:organize our data
- Step3:split our data
- Step3:initialise our classifier
- Step4:train our classifier
- Step5:make predictions
- Step6:evaluate accuracy

```
x=df[['HFMS','GMFM']].as_matrix()
y=df['disease'].as_matrix()
```

```
train, test, train_labels, test_labels = train_test_split(x,
y,
test_size=0.33,
random_state=42)
```

```
gnb = GaussianNB()
```

```
model = gnb.fit(train, train_labels)
```

```
model
```

```
GaussianNB(priors=None, var_smoothing=1e-09)
```

```
preds = gnb.predict(test)
print(preds)
```

```
[1 1 3 1 4 4 4 2 4 1 4 1 3 4 1 2 1 1 4 2]
```

```
print(accuracy_score(test_labels, preds))
```

```
0.95
```

IV. CONCLUSION

By giving the blood samples of children with respect to CPK,AGE,HFMS and GMFM my model easily classify the groups of SMA's and also identify whether the person is effected by which group of spinal muscle atropy.According to this classification we can easily survive the lifespan of children caused by the disease.

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