

Homework Assignment for Chapter 13

13.1 Hepatic Injury

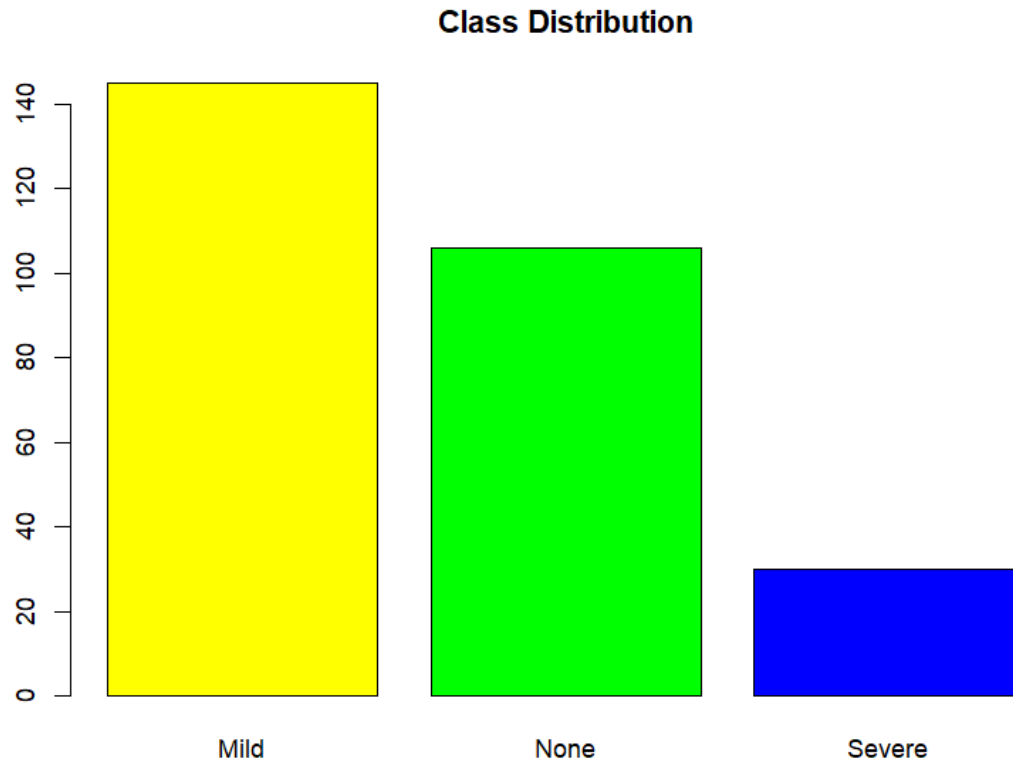


Fig 1: Distribution of Difference classes in the response variable injury

a) Model building

MDA

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Mixture Discriminant Analysis
225 samples
96 predictor
3 classes: 'Mild', 'None', 'Severe'

No pre-processing
Resampling: Repeated Train/Test Splits Estimated (25 reps, 75%)
Summary of sample sizes: 169, 169, 169, 169, 169, 169, ...
Resampling results across tuning parameters:

subclasses logloss AUC prAUC Accuracy Kappa Mean_F1 Mean_Sensitivity
1 4.178933 0.5376305 0.3495797 0.4035714 0.00803808 0.3477111 0.3529611
2 5.961177 0.5429546 0.3587080 0.4214286 0.03755265 0.3817567 0.3688342
3 7.266982 0.5597702 0.3671220 0.4450000 0.05735752 0.3953140 0.3676190
4 8.231726 0.5635855 0.3708688 0.4450000 0.06064775 0.3999887 0.3811823
5 9.267994 0.5602360 0.3627392 0.4457143 0.05788837 0.3977113 0.3821675
6 10.265879 0.5678326 0.3589256 0.4414286 0.04318430 0.3941713 0.3790695
7 11.079285 0.5662708 0.3478225 0.4428571 0.04885807 0.4002747 0.3716913
8 12.140675 0.5517320 0.3333661 0.4371429 0.03142476 0.3904543 0.3652107
9 12.605192 0.5693813 0.3252889 0.4528571 0.05298365 0.4100283 0.3770772
10 13.320107 0.5609963 0.3172712 0.4421429 0.04028470 0.3940754 0.3680679
11 13.663628 0.5674444 0.3172243 0.4514286 0.04937839 0.4197958 0.3833826
12 14.407416 0.5614268 0.2988344 0.4428571 0.03636062 0.4154443 0.3761138
13 15.234782 0.5544155 0.2831353 0.4364286 0.02097973 0.3872224 0.3589491
14 15.708051 0.5470113 0.2780329 0.4371429 0.02896333 0.3884731 0.3586973
Mean_Specificity Mean_Pos_Pred_Value Mean_Neg_Pred_Value Mean_Precision Mean_Recall
0.6677220 0.3477920 0.6677495 0.3477920 0.3529611
0.6788092 0.3633894 0.6780216 0.3633894 0.3688342
0.6869474 0.3639282 0.6860365 0.3639282 0.3676190
0.6862674 0.3797372 0.6859527 0.3797372 0.3811823
0.6845982 0.3789535 0.6848840 0.3789535 0.3821675
0.6789093 0.3739619 0.6794499 0.3739619 0.3790695
0.6817115 0.3680451 0.6821630 0.3680451 0.3716913
0.6752720 0.3588369 0.6758429 0.3588369 0.3652107
0.6834004 0.3803254 0.6834082 0.3803254 0.3770772
0.6790688 0.3673439 0.6783030 0.3673439 0.3680679
0.6808296 0.3851673 0.6809530 0.3851673 0.3833826
0.6759252 0.3777984 0.6765182 0.3777984 0.3761138
0.6717150 0.3593790 0.6719557 0.3593790 0.3589491
0.6748289 0.3627629 0.6749317 0.3627629 0.3586973
Mean_Detection_Rate Mean_Balanced_Accuracy
0.1345238 0.5103416
0.1345238 0.5103416
```

0.1343238	0.5103418
0.1404762	0.5238217
0.1483333	0.5272832
0.1483333	0.5337248
0.1485714	0.5333829
0.1471429	0.5289894
0.1476190	0.5267014
0.1457143	0.5202413
0.1509524	0.5302388
0.1473810	0.5235683
0.1504762	0.5321061
0.1476190	0.5260195
0.1454762	0.5153320
0.1457143	0.5167631

Accuracy was used to select the optimal model using the largest value.
The final value used for the model was subclasses = 9.

Confusion Matrix and Statistics

	Reference		
Prediction	Mild	None	Severe
Mild	14	11	4
None	12	9	0
Severe	3	1	2

Overall Statistics

Accuracy : 0.4464
95% CI : (0.3134, 0.5853)
No Information Rate : 0.5179
P-Value [Acc > NIR] : 0.8856

Kappa : 0.0451

Mcnemar's Test P-Value : 0.7563

Statistics by Class:

	Class: Mild	Class: None	Class: Severe
Sensitivity	0.4828	0.4286	0.33333
Specificity	0.4444	0.6571	0.92000
Pos Pred Value	0.4828	0.4286	0.33333
Neg Pred Value	0.4444	0.6571	0.92000
Prevalence	0.5179	0.3750	0.10714
Detection Rate	0.2500	0.1607	0.03571
Detection Prevalence	0.5179	0.3750	0.10714
Balanced Accuracy	0.4636	0.5429	0.62667

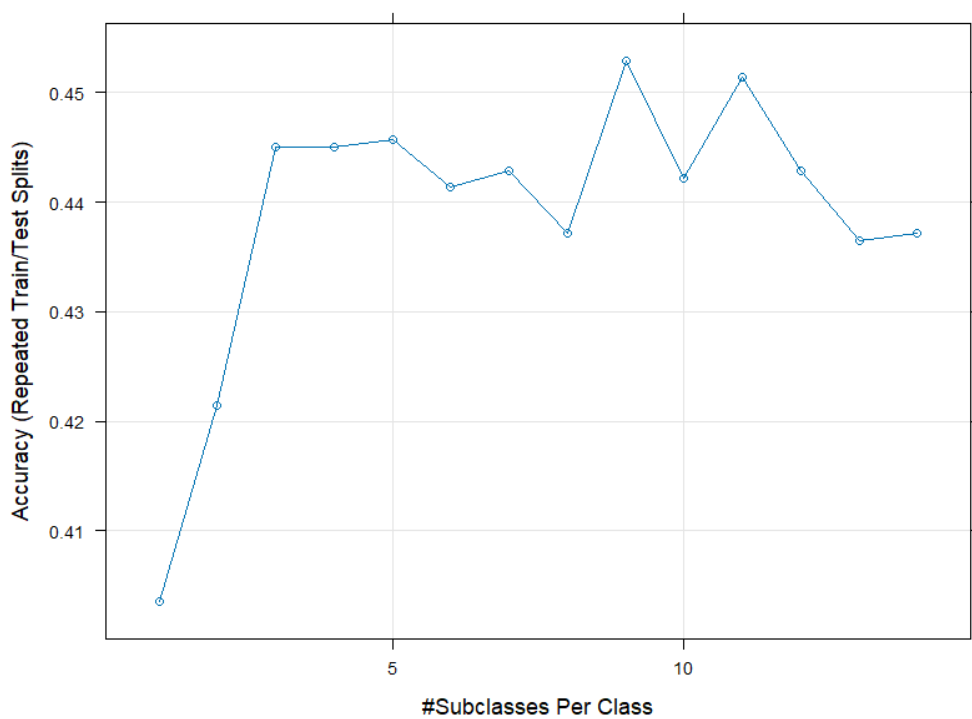


Fig 2: Plot of Accuracy vs Subclasses for MDA model

RDA

Regularized Discriminant Analysis

225 samples
96 predictor
3 classes: 'Mild', 'None', 'Severe'

No pre-processing

Resampling: Bootstrapped (25 reps)

Summary of sample sizes: 225, 225, 225, 225, 225, ...

Resampling results across tuning parameters:

gamma	lambda	Accuracy	Kappa
0.1	0.01	0.4874741	0.05031715
0.1	0.02	0.4868943	0.05191302
0.1	0.03	0.4884157	0.05904163
0.1	0.04	0.4850267	0.05626032
0.1	0.05	0.4829684	0.05444879
0.1	0.06	0.4809693	0.05216068
0.1	0.07	0.4796206	0.05089113
0.1	0.08	0.4797085	0.05179773
0.1	0.09	0.4778953	0.04958311
0.1	0.10	0.4750367	0.04660817
0.2	0.01	0.4808991	0.06328720
0.2	0.02	0.4822783	0.06719870
0.2	0.03	0.4804187	0.06504344
0.2	0.04	0.4800186	0.06752597
0.2	0.05	0.4801165	0.06808543
0.2	0.06	0.4782244	0.06639986
0.2	0.07	0.4777063	0.06754982
0.2	0.08	0.4767251	0.06590818
0.2	0.09	0.4785982	0.06991458
0.2	0.10	0.4761319	0.06687623
0.3	0.01	0.4765511	0.07184971
0.3	0.02	0.4761650	0.07162001
0.3	0.03	0.4733846	0.06854029
0.3	0.04	0.4699619	0.06340166
0.3	0.05	0.4684744	0.06092653
0.3	0.06	0.4689290	0.06149693
0.3	0.07	0.4669791	0.06043329
0.3	0.08	0.4670166	0.06148844
0.3	0.09	0.4659894	0.06104998
0.3	0.10	0.4669284	0.06460421
0.4	0.01	0.4677126	0.06715168
0.4	0.02	0.4653623	0.06442657
0.4	0.03	0.4649470	0.06473755
0.4	0.04	0.4639436	0.06431922
0.4	0.05	0.4634011	0.06423774

0.9	0.05	0.4338284	0.05131360
0.9	0.06	0.4338284	0.05131360
0.9	0.07	0.4328438	0.04923561
0.9	0.08	0.4323787	0.04813720
0.9	0.09	0.4333396	0.04961283
0.9	0.10	0.4328004	0.04858869
1.0	0.01	0.4335286	0.04451921
1.0	0.02	0.4335286	0.04451921
1.0	0.03	0.4335286	0.04451921
1.0	0.04	0.4335286	0.04451921
1.0	0.05	0.4340414	0.04504590
1.0	0.06	0.4340414	0.04504590
1.0	0.07	0.4340414	0.04504590
1.0	0.08	0.4340414	0.04504590
1.0	0.09	0.4354856	0.04652657
1.0	0.10	0.4354856	0.04652657

Accuracy was used to select the optimal model using the largest value.
The final values used for the model were gamma = 0.1 and lambda = 0.03.

Confusion Matrix and Statistics

		Reference		
Prediction		Mild	None	Severe
Mild		19	12	5
None		8	9	0
Severe		2	0	1

Overall Statistics

Accuracy : 0.5179
 95% CI : (0.3803, 0.6534)
 No Information Rate : 0.5179
 P-Value [Acc > NIR] : 0.5537

Kappa : 0.1194

Mcnemar's Test P-Value : NA

Statistics by Class:

	Class: Mild	Class: None	Class: Severe
Sensitivity	0.6552	0.4286	0.16667
Specificity	0.3704	0.7714	0.96000
Pos Pred Value	0.5278	0.5294	0.33333
Neg Pred Value	0.5000	0.6923	0.90566
Prevalence	0.5179	0.3750	0.10714
Detection Rate	0.3393	0.1607	0.01786
Detection Prevalence	0.6429	0.3036	0.05357
Balanced Accuracy	0.5128	0.6000	0.56333

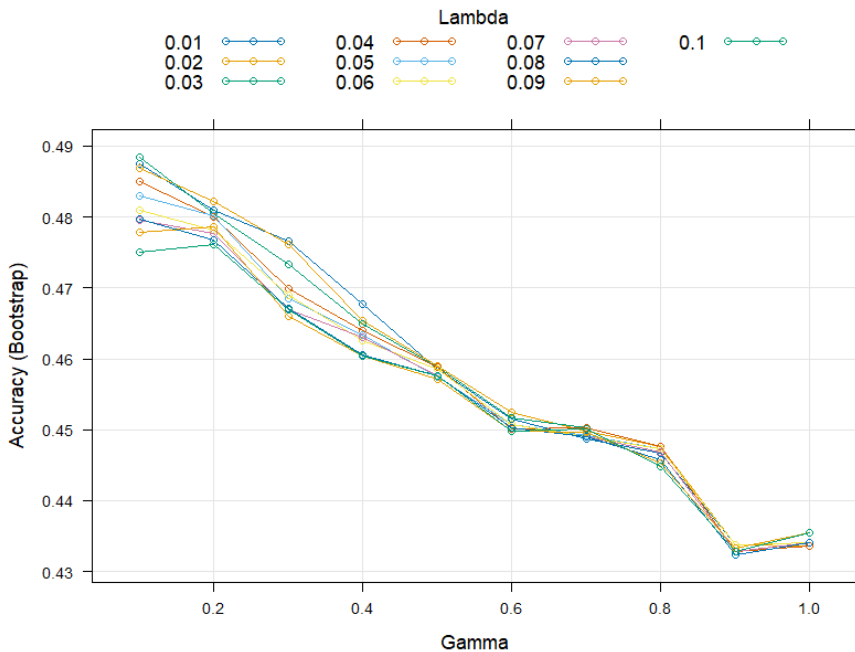


Fig 3: Accuracy vs Gamma for RDA model

Neural Networks

Neural Network

225 samples
96 predictor
3 classes: 'Mild', 'None', 'Severe'

Pre-processing: centered (96), scaled (96), spatial sign transformation (96)
Resampling: Repeated Train/Test Splits Estimated (25 reps, 75%)
Summary of sample sizes: 169, 169, 169, 169, 169, 169, ...
Resampling results across tuning parameters:

size	decay	logLoss	AUC	prAUC	Accuracy	Kappa	Mean_F1	Mean_Sensitivity
1	0.0	6.9067257	0.5146908	0.2130704	0.4085714	0.015590316	0.3044242	0.3452436
1	0.1	1.4213475	0.5219499	0.3453776	0.4764286	0.033731933	NaN	0.3488779
1	1.0	0.9479833	0.5209672	0.3530608	0.5178571	0.000000000	NaN	0.3333333
1	2.0	0.9489150	0.5241853	0.3358213	0.5178571	0.000000000	NaN	0.3333333
2	0.0	11.3579795	0.5401471	0.2758850	0.4342857	0.073753866	0.3748754	0.3751724
2	0.1	1.5610775	0.5609116	0.3610982	0.4635714	0.058987106	0.4116880	0.3659880
2	1.0	0.9459389	0.5600953	0.3677187	0.5178571	0.001709760	NaN	0.3340339
2	2.0	0.9485471	0.5216771	0.3333431	0.5178571	0.000000000	NaN	0.3333333
3	0.0	15.0001059	0.5243451	0.2894801	0.4292857	0.030423121	0.3732720	0.3520307
3	0.1	1.6036631	0.5561929	0.3564728	0.4650000	0.063372105	0.4083882	0.3668966
3	1.0	0.9457460	0.5599702	0.3652539	0.5178571	0.001709760	NaN	0.3340339
3	2.0	0.9483343	0.5217013	0.3327701	0.5178571	0.000000000	NaN	0.3333333
4	0.0	14.0627456	0.5531144	0.3082265	0.4692857	0.078292832	0.4026298	0.3821894
4	0.1	1.5837189	0.5614894	0.3614097	0.4592857	0.051165150	0.3833732	0.3623536
4	1.0	0.9457460	0.5603142	0.3651098	0.5192857	0.005043185	NaN	0.3353038
4	2.0	0.9482002	0.5224848	0.3333829	0.5178571	0.000000000	NaN	0.3333333
5	0.0	13.9984631	0.5419934	0.3043854	0.4571429	0.073064915	0.4016557	0.3773618
5	0.1	1.5927907	0.5603283	0.3599704	0.4585714	0.046366378	0.3793465	0.3617187
5	1.0	0.9457042	0.5599158	0.3651410	0.5200000	0.006727395	NaN	0.3359387
5	2.0	0.9481112	0.5215531	0.3322930	0.5178571	0.000000000	NaN	0.3333333
6	0.0	14.4905680	0.5544497	0.3017605	0.4507143	0.064592687	0.3794914	0.3723481
6	0.1	1.5640626	0.5662920	0.3656767	0.4771429	0.079852674	0.3950074	0.3785988
6	1.0	0.9457180	0.5594470	0.3652778	0.5200000	0.006727395	NaN	0.3359387
6	2.0	0.9480494	0.5209716	0.3328669	0.5178571	0.000000000	NaN	0.3333333
7	0.0	14.1970160	0.5608607	0.3093591	0.4500000	0.069983536	0.4040049	0.3896771
7	0.1	1.5801650	0.5627807	0.3605709	0.4657143	0.065260520	0.3952149	0.3703558
7	1.0	0.9457266	0.5593762	0.3655169	0.5200000	0.006727395	NaN	0.3359387
7	2.0	0.9480056	0.5179086	0.3308895	0.5178571	0.000000000	NaN	0.3333333
8	0.0	13.9653341	0.5644837	0.3129434	0.4450000	0.051565325	0.3873478	0.3691954
8	0.1	1.5905984	0.5615815	0.3606004	0.4642857	0.056999157	0.3889334	0.3689217
8	1.0	0.9457425	0.5599985	0.3657543	0.5200000	0.006727395	NaN	0.3359387
8	2.0	0.9479734	0.5175584	0.3307302	0.5178571	0.000000000	NaN	0.3333333
9	0.0	13.6280599	0.5653923	0.3188494	0.4578571	0.064592732	0.4016033	0.3825835
9	0.1	1.6061619	0.5650530	0.3632408	0.4657143	0.061713769	0.3890032	0.3687794

0.1530952	0.5227733
0.1730952	0.5017048
0.1726190	0.5000000
0.1523810	0.5341391
0.1528571	0.5215846
0.1733333	0.5022692
0.1726190	0.5000000
0.1502381	0.5303801
0.1590476	0.5355985
0.1733333	0.5022692
0.1726190	0.5000000
0.1500000	0.5392463
0.1552381	0.5291088
0.1733333	0.5022692
0.1726190	0.5000000
0.1483333	0.5260242
0.1547619	0.5269356
0.1733333	0.5022692
0.1726190	0.5000000
0.1526190	0.5347810
0.1552381	0.5278698
0.1733333	0.5022692
0.1726190	0.5000000
0.1492857	0.5293807
0.1559524	0.5297055
0.1733333	0.5022692
0.1726190	0.5000000

Accuracy was used to select the optimal model using the largest value.
The final values used for the model were size = 5 and decay = 1.

Confusion Matrix and Statistics

	Reference		
Prediction	Mild	None	Severe
Mild	27	20	6
None	2	1	0
Severe	0	0	0

Overall Statistics

Accuracy : 0.5
95% CI : (0.3634, 0.6366)
No Information Rate : 0.5179
P-Value [Acc > NIR] : 0.6562

Kappa : -0.0208

Mcnemar's Test P-Value : NA

Statistics by Class:

	Class: Mild	Class: None	Class: Severe
Sensitivity	0.93103	0.04762	0.0000
Specificity	0.03704	0.94286	1.0000
Pos Pred Value	0.50943	0.33333	NaN
Neg Pred Value	0.33333	0.62264	0.8929
Prevalence	0.51786	0.37500	0.1071
Detection Rate	0.48214	0.01786	0.0000
Detection Prevalence	0.94643	0.05357	0.0000
Balanced Accuracy	0.48404	0.49524	0.5000

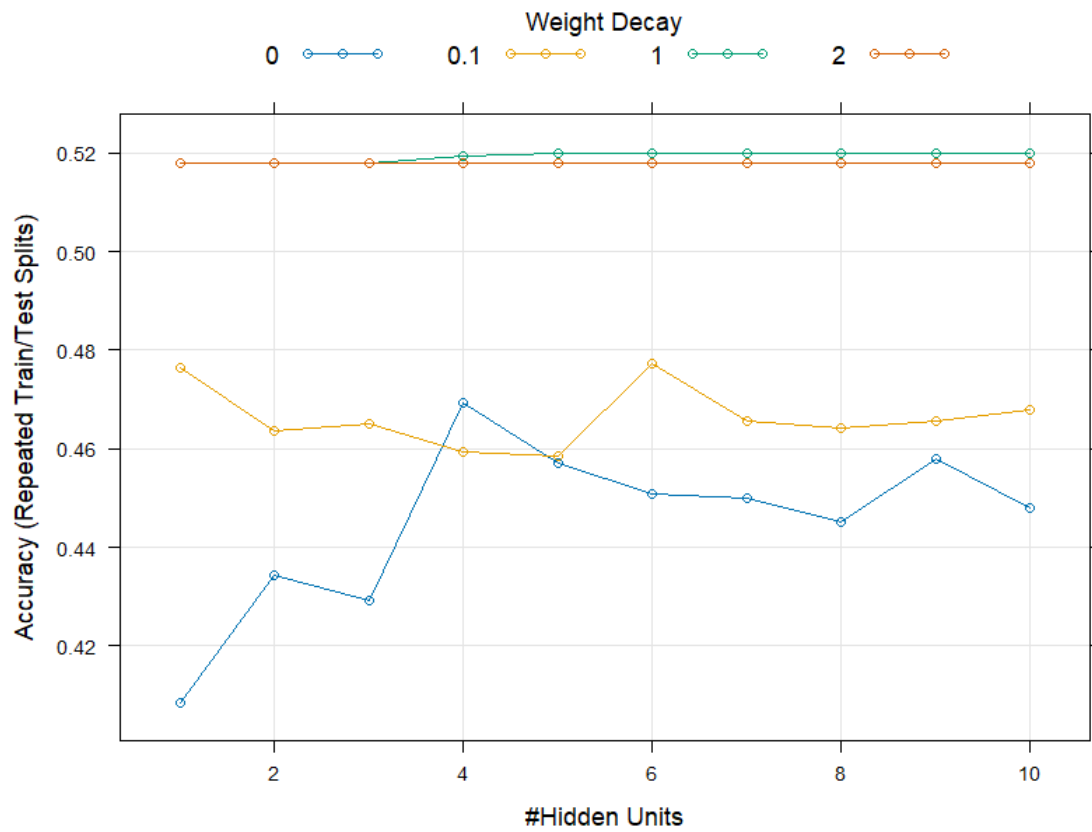


Fig 4: Accuracy vs Number of Hidden Units for Neural Networks

FDA

Flexible Discriminant Analysis

225 samples

96 predictor

3 classes: 'Mild', 'None', 'Severe'

No pre-processing

Resampling: Cross-Validated (10 fold)

Summary of sample sizes: 204, 203, 201, 202, 203, 202, ...

Resampling results across tuning parameters:

degree	nprune	Accuracy	Kappa
1	2	0.5075216	0.004884179
1	3	0.5118695	0.016206664
1	4	0.5162173	0.044165306
1	5	0.5120671	0.046545197
1	6	0.4944782	0.031727485
1	7	0.4901303	0.026468028
1	8	0.4901303	0.026468028
1	9	0.4857825	0.022579589
1	10	0.4903280	0.035177854
1	11	0.4946758	0.040171469
1	12	0.4859801	0.029673108
1	13	0.4859801	0.029673108
1	14	0.4859801	0.029673108
1	15	0.4859801	0.029673108
1	16	0.4859801	0.029673108
1	17	0.4859801	0.029673108
1	18	0.4859801	0.029673108
1	19	0.4859801	0.029673108
1	20	0.4859801	0.029673108
1	21	0.4859801	0.029673108
1	22	0.4859801	0.029673108
1	23	0.4859801	0.029673108
1	24	0.4859801	0.029673108
1	25	0.4859801	0.029673108
1	26	0.4859801	0.029673108
1	27	0.4859801	0.029673108
1	28	0.4859801	0.029673108
1	29	0.4859801	0.029673108
1	30	0.4859801	0.029673108
1	31	0.4859801	0.029673108
1	32	0.4859801	0.029673108
1	33	0.4859801	0.029673108
1	34	0.4859801	0.029673108
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2	10	0.4670078	0.005718410
2	11	0.4806442	0.032643610
2	12	0.4806442	0.032643610
2	13	0.4806442	0.028159764
2	14	0.4851896	0.040090981
2	15	0.4851896	0.040090981
2	16	0.4988260	0.062523694
2	17	0.4988260	0.062523694
2	18	0.4988260	0.062523694
2	19	0.4988260	0.062523694
2	20	0.4988260	0.062523694
2	21	0.4988260	0.062523694
2	22	0.4988260	0.062523694
2	23	0.4988260	0.062523694
2	24	0.4988260	0.062523694
2	25	0.4988260	0.062523694
2	26	0.4988260	0.062523694
2	27	0.4988260	0.062523694
2	28	0.4988260	0.062523694
2	29	0.4988260	0.062523694
2	30	0.4988260	0.062523694
2	31	0.4988260	0.062523694
2	32	0.4988260	0.062523694
2	33	0.4988260	0.062523694
2	34	0.4988260	0.062523694
2	35	0.4988260	0.062523694
2	36	0.4988260	0.062523694
2	37	0.4988260	0.062523694
2	38	0.4988260	0.062523694

Accuracy was used to select the optimal model using the largest value.
The final values used for the model were degree = 1 and nprune = 4.

Confusion Matrix and Statistics

	Reference		
Prediction	Mild	None	Severe
Mild	24	17	4
None	2	1	0
Severe	3	3	2

Overall Statistics

Accuracy : 0.4821
95% CI : (0.3466, 0.6197)
No Information Rate : 0.5179
P-Value [Acc > NIR] : 0.74822

Kappa : 0.0558

McNemar's Test P-Value : 0.00183

Statistics by Class:

	Class: Mild	Class: None	Class: Severe
Sensitivity	0.8276	0.04762	0.33333
Specificity	0.2222	0.94286	0.88000
Pos Pred Value	0.5333	0.33333	0.25000
Neg Pred Value	0.5455	0.62264	0.91667
Prevalence	0.5179	0.37500	0.10714
Detection Rate	0.4286	0.01786	0.03571
Detection Prevalence	0.8036	0.05357	0.14286
Balanced Accuracy	0.5249	0.49524	0.60667

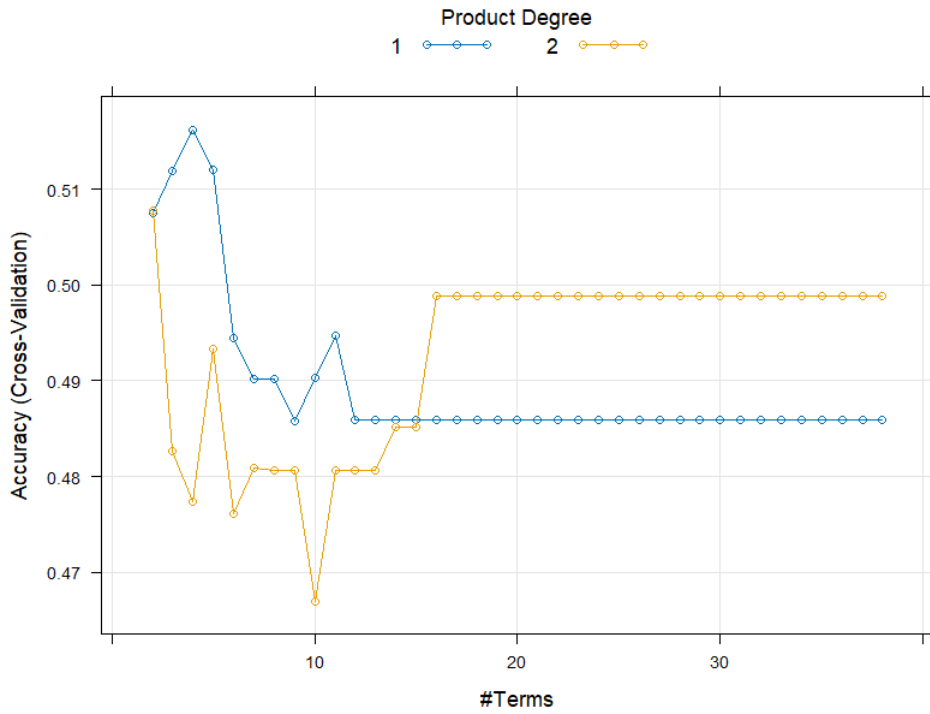


Fig 5: Accuracy vs Number of Terms for FDA model

SVM

Support Vector Machines with Radial Basis Function Kernel

225 samples
96 predictor
3 classes: 'Mild', 'None', 'Severe'

Pre-processing: centered (96), scaled (96)
Resampling: Repeated Train/Test Splits Estimated (25 reps, 75%)
Summary of sample sizes: 169, 169, 169, 169, 169, 169, ...
Resampling results across tuning parameters:

C	logLoss	AUC	prAUC	Accuracy	Kappa	Mean_F1	Mean_Sensitivity
0.0625	0.9451827	0.5487008	0.3530637	0.5114286	-0.0004991687	NaN	0.3330487
0.1250	0.9475734	0.5462783	0.3567186	0.5078571	0.0031918851	NaN	0.3339026
0.2500	0.9475976	0.5547264	0.3579220	0.5014286	-0.0113130144	NaN	0.3287137
0.5000	0.9510538	0.5464115	0.3558365	0.5042857	0.0034489606	0.3886207	0.3342419
1.0000	0.9483945	0.5396770	0.3495982	0.5085714	0.0005879288	NaN	0.3326108
2.0000	0.9470089	0.5522489	0.3603225	0.5142857	0.0159662973	NaN	0.3390914
4.0000	0.9460979	0.5546404	0.3527642	0.5164286	0.0123684544	NaN	0.3374932
8.0000	0.9508793	0.5412411	0.3435207	0.5085714	-0.0002286653	NaN	0.3326108
16.0000	0.9459449	0.5479181	0.3484454	0.5057143	-0.0123423802	NaN	0.3288451
32.0000	0.9455865	0.5527670	0.3541242	0.5021429	-0.0158950304	NaN	0.3265463
64.0000	0.9507604	0.5376813	0.3419671	0.5064286	-0.0038100361	NaN	0.3310564
128.0000	0.9485094	0.5456791	0.3493259	0.5021429	-0.0175304205	NaN	0.3265463
256.0000	0.9501055	0.5459662	0.3466640	0.4978571	-0.0114460868	NaN	0.3285167
512.0000	0.9466037	0.5552995	0.3542447	0.4992857	-0.0188461116	NaN	0.3261084
1024.0000	0.9519703	0.5424939	0.3436692	0.4978571	-0.0208656997	NaN	0.3246634
2048.0000	0.9474419	0.5418183	0.3443299	0.5021429	-0.0069294694	NaN	0.3302244
4096.0000	0.9490168	0.5441076	0.3508953	0.5042857	-0.0031066229	NaN	0.3314286
8192.0000	0.9487912	0.5467959	0.3455668	0.5107143	0.0069171950	NaN	0.3353914
16384.0000	0.9493023	0.5361636	0.3420038	0.4992857	-0.0151819113	NaN	0.3271593
Mean_Specificity	Mean_Pos_Pred_Value	Mean_Neg_Pred_Value	Mean_Precision	Mean_Recall			
0.6664014	0.3190788	0.6411652	0.3190788	0.3330487			
0.6676628	0.3376378	0.6338724	0.3376378	0.3339026			
0.6629898	NaN	0.6135092	NaN	0.3287137			
0.6674610	0.4558620	0.6655773	0.4558620	0.3342419			
0.6669178	0.4097222	0.6297972	0.4097222	0.3326108			
0.6715372	0.5092593	0.7010024	0.5092593	0.3390914			
0.6707626	NaN	0.6518247	NaN	0.3374932			
0.6669108	0.2818930	0.6299098	0.2818930	0.3326108			
0.6627866	NaN	0.6331339	NaN	0.3288451			
0.6617905	0.3189300	0.6325303	0.3189300	0.3265463			
0.6654321	0.3379630	0.6608749	0.3379630	0.3310564			
0.1704762		0.4997251					
0.1692857		0.5007827					
0.1671429		0.4958518					
0.1680952		0.5008515					
0.1695238		0.4997643					
0.1714286		0.5053143					
0.1721429		0.5041279					
0.1695238		0.4997608					
0.1685714		0.4958158					
0.1673810		0.4941684					
0.1688095		0.4982442					
0.1673810		0.4938848					
0.1659524		0.4957038					
0.1664286		0.4933561					
0.1659524		0.4924220					
0.1673810		0.4973619					
0.1680952		0.4984797					
0.1702381		0.5021733					
0.1664286		0.4945581					

Tuning parameter 'sigma' was held constant at a value of 0.003923164
Accuracy was used to select the optimal model using the largest value.
The final values used for the model were sigma = 0.003923164 and C = 4.

Confusion Matrix and Statistics

		Reference		
Prediction		Mild	None	Severe
Mild		23	15	6
None		6	6	0
Severe		0	0	0

Overall Statistics

Accuracy : 0.5179
 95% CI : (0.3803, 0.6534)
 No Information Rate : 0.5179
 P-Value [Acc > NIR] : 0.5537

Kappa : 0.0597

McNemar's Test P-Value : NA

Statistics by Class:

	Class: Mild	Class: None	Class: Severe
Sensitivity	0.7931	0.2857	0.0000
Specificity	0.2222	0.8286	1.0000
Pos Pred Value	0.5227	0.5000	NaN
Neg Pred Value	0.5000	0.6591	0.8929
Prevalence	0.5179	0.3750	0.1071
Detection Rate	0.4107	0.1071	0.0000
Detection Prevalence	0.7857	0.2143	0.0000
Balanced Accuracy	0.5077	0.5571	0.5000

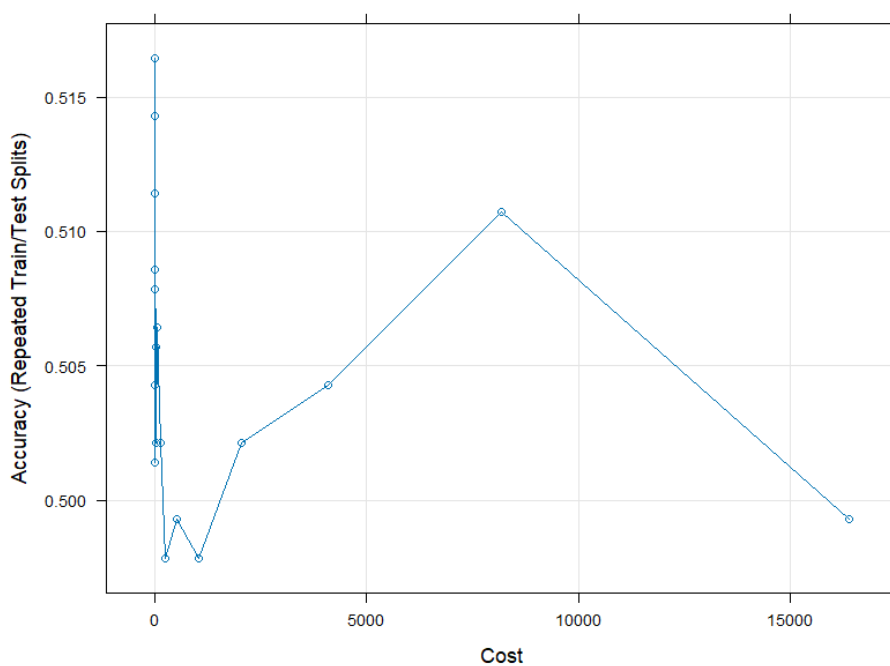


Fig 6: Accuracy vs Cost for SVM model

KNN

k-Nearest Neighbors

225 samples

96 predictor

3 classes: 'Mild', 'None', 'Severe'

Pre-processing: centered (96), scaled (96)

Resampling: Repeated Train/Test Splits Estimated (25 reps, 75%)

Summary of sample sizes: 169, 169, 169, 169, 169, 169, ...

Resampling results across tuning parameters:

k	logLoss	AUC	prAUC	Accuracy	Kappa	Mean_F1	Mean_Sensitivity
1	18.9963270	0.5310711	0.1976138	0.4500000	0.053717583	0.3959582	0.3803503
2	11.3365627	0.5363018	0.2793809	0.4450000	0.048486713	0.3880853	0.3646196
3	7.3426198	0.5453810	0.3242071	0.4671429	0.066964085	0.4000421	0.3703996
4	5.0532838	0.5479730	0.3243801	0.4628571	0.046260806	0.3925813	0.3516037
5	3.5530166	0.5464189	0.3224836	0.4514286	0.023511925	0.4005005	0.3398577
6	2.6265778	0.5506113	0.3295840	0.4528571	0.015766640	0.3760155	0.3328517
7	2.0626004	0.5540827	0.3349943	0.4692857	0.036992421	0.3942972	0.3480131
8	1.8020623	0.5537039	0.3379326	0.4635714	0.018628177	0.3967499	0.3392337
9	1.6800351	0.5581678	0.3433392	0.4814286	0.043088900	0.3853038	0.3532020
10	1.5334626	0.5654217	0.3492420	0.4857143	0.040582373	0.3899712	0.3534975
11	1.4054843	0.5751680	0.3511997	0.4957143	0.059547125	0.4200033	0.3634483
12	1.3313716	0.5793099	0.3610371	0.5014286	0.065836519	0.4213799	0.3666010
13	1.3028985	0.5827957	0.3636950	0.5121429	0.081692382	0.4244858	0.3708484
14	1.2359870	0.5801250	0.3543548	0.5121429	0.078877813	0.4513379	0.3682102
15	1.1662873	0.5810417	0.3587676	0.5178571	0.086657887	0.4423274	0.3733005
16	1.1013638	0.5756614	0.3580422	0.5114286	0.069761292	NaN	0.3615982
17	1.1017772	0.5732400	0.3525260	0.4964286	0.040371924	NaN	0.3498413
18	1.0587828	0.5703087	0.3582241	0.5042857	0.053261487	NaN	0.3554242
19	1.0358610	0.5698573	0.3577822	0.5050000	0.051870680	NaN	0.3541325
20	1.0351075	0.5711431	0.3609683	0.5064286	0.055014249	NaN	0.3554023
21	1.0327581	0.5731885	0.3567892	0.5000000	0.041034067	NaN	0.3496880
22	1.0329074	0.5719367	0.3576886	0.4992857	0.040034498	NaN	0.3499288
23	1.0333360	0.5704217	0.3610005	0.5028571	0.044409162	NaN	0.3511768
24	1.0129892	0.5675493	0.3602013	0.4964286	0.032469647	NaN	0.3468637
25	1.0138692	0.5679919	0.3622698	0.4957143	0.029842873	NaN	0.3457033
26	1.0128046	0.5689153	0.3668276	0.4957143	0.028126724	NaN	0.3451779
27	0.9898370	0.5732480	0.3665719	0.4985714	0.029765880	NaN	0.3457909
28	0.9680365	0.5730823	0.3636153	0.5007143	0.035717724	NaN	0.3482211
29	0.9421735	0.5779769	0.3717445	0.5071429	0.046236034	NaN	0.3523591
30	0.9424923	0.5774535	0.3749702	0.5021429	0.036001862	NaN	0.3486152
31	0.9406746	0.5810367	0.3779631	0.4992857	0.028389580	NaN	0.3455501
32	0.9403977	0.5820836	0.3724056	0.5014286	0.032846997	NaN	0.3471045
33	0.9396675	0.5840786	0.3772028	0.5042857	0.036340334	NaN	0.3489436

0.1680952	0.5192782
0.1683333	0.5185971
0.1688095	0.5196694
0.1666667	0.5146465
0.1664286	0.5145200
0.1676190	0.5159270
0.1654762	0.5117176
0.1652381	0.5107212
0.1652381	0.5101762
0.1661905	0.5106803
0.1669048	0.5128619
0.1690476	0.5164758
0.1673810	0.5129319
0.1664286	0.5103553
0.1671429	0.5118733
0.1680952	0.5131596
0.1669048	0.5094114
0.1669048	0.5083158
0.1666667	0.5068620
0.1673810	0.5079791
0.1652381	0.5019792
0.1657143	0.5027070
0.1626190	0.4952826
0.1650000	0.4994298
0.1669048	0.5031371
0.1664286	0.5012066
0.1645238	0.4972677
0.1647619	0.4972560
0.1654762	0.4985171
0.1664286	0.4997976
0.1680952	0.5020512
0.1664286	0.4995407
0.1678571	0.5014304

Accuracy was used to select the optimal model using the largest value.
The final value used for the model was k = 15.

Confusion Matrix and Statistics

	Reference		
Prediction	Mild	None	Severe
Mild	22	13	4
None	7	8	1
Severe	0	0	1

Overall Statistics

Accuracy : 0.5536
95% CI : (0.4147, 0.6866)
No Information Rate : 0.5179
P-Value [Acc > NIR] : 0.34484

Kappa : 0.1581

McNemar's Test P-Value : 0.07855

Statistics by Class:

	Class: Mild	Class: None	Class: Severe
Sensitivity	0.7586	0.3810	0.16667
Specificity	0.3704	0.7714	1.00000
Pos Pred Value	0.5641	0.5000	1.00000
Neg Pred Value	0.5882	0.6750	0.90909
Prevalence	0.5179	0.3750	0.10714
Detection Rate	0.3929	0.1429	0.01786
Detection Prevalence	0.6964	0.2857	0.01786
Balanced Accuracy	0.5645	0.5762	0.58333

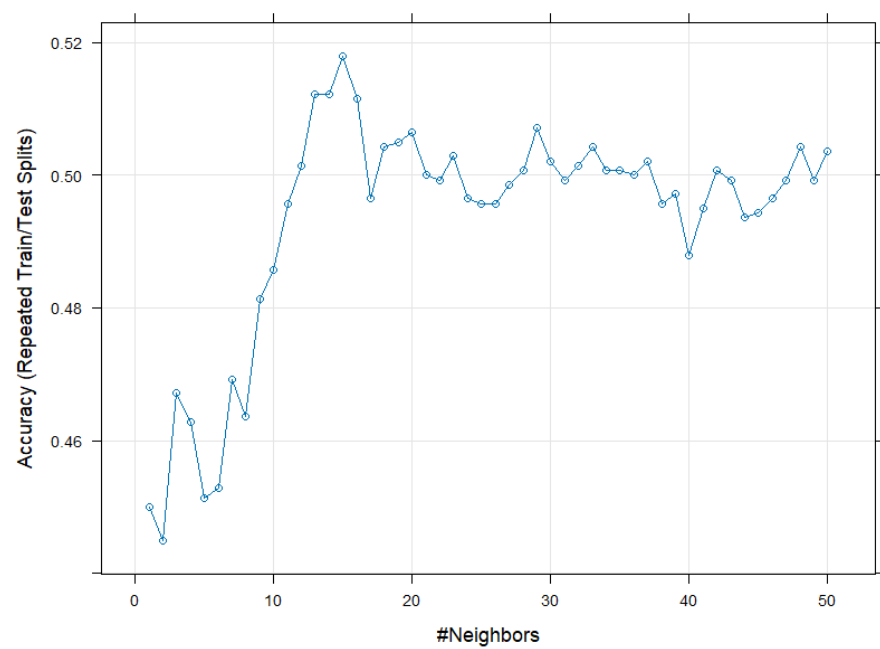


Fig 7: Accuracy vs Number of Neighbors for KNN model

Naïve Bayes

Naïve Bayes

225 samples
96 predictor
3 classes: 'Mild', 'None', 'Severe'

No pre-processing

Resampling: Repeated Train/Test Splits Estimated (25 reps, 75%)

Summary of sample sizes: 169, 169, 169, 169, 169, 169, ...

Resampling results across tuning parameters:

fL	logLoss	AUC	prAUC	Accuracy	Kappa	Mean_F1	Mean_Sensitivity	Mean_Specificity						
2	15.67154	0.4958025	0.3056379	0.3314286	-0.03444965	0.3017131	0.3225616	0.650781						
3	15.67154	0.4958025	0.3056379	0.3314286	-0.03444965	0.3017131	0.3225616	0.650781						
4	15.67154	0.4958025	0.3056379	0.3314286	-0.03444965	0.3017131	0.3225616	0.650781						
Mean_Pos_Pred_Value		Mean_Neg_Pred_Value		Mean_Precision		Mean_Recall		Mean_Detection_Rate						
0.3154702		0.6535535		0.3154702		0.3225616		0.1104762						
0.3154702		0.6535535		0.3154702		0.3225616		0.1104762						
0.3154702		0.6535535		0.3154702		0.3225616		0.1104762						
Mean_Balanced_Accuracy														
0.4866713														
0.4866713														
0.4866713														

Tuning parameter 'usekernel' was held constant at a value of TRUE

Tuning parameter 'adjust' was held
constant at a value of TRUE

Accuracy was used to select the optimal model using the largest value.

The final values used for the model were fL = 2, usekernel = TRUE and adjust = TRUE.

```
> ConfusionMatrix(fittedData = fitted, reference = y.test)
Confusion Matrix and Statistics
```

	Reference		
Prediction	Mild	None	Severe
Mild	15	6	2
None	8	8	1
Severe	6	7	3

Overall Statistics

Accuracy : 0.4643
95% CI : (0.3299, 0.6026)
No Information Rate : 0.5179
P-Value [Acc > NIR] : 0.82537

Kappa : 0.1667

McNemar's Test P-Value : 0.07905

Statistics by Class:

	Class: Mild	Class: None	Class: Severe
Sensitivity	0.5172	0.3810	0.50000
Specificity	0.7037	0.7429	0.74000
Pos Pred Value	0.6522	0.4706	0.18750
Neg Pred Value	0.5758	0.6667	0.92500
Prevalence	0.5179	0.3750	0.10714
Detection Rate	0.2679	0.1429	0.05357
Detection Prevalence	0.4107	0.3036	0.28571
Balanced Accuracy	0.6105	0.5619	0.62000

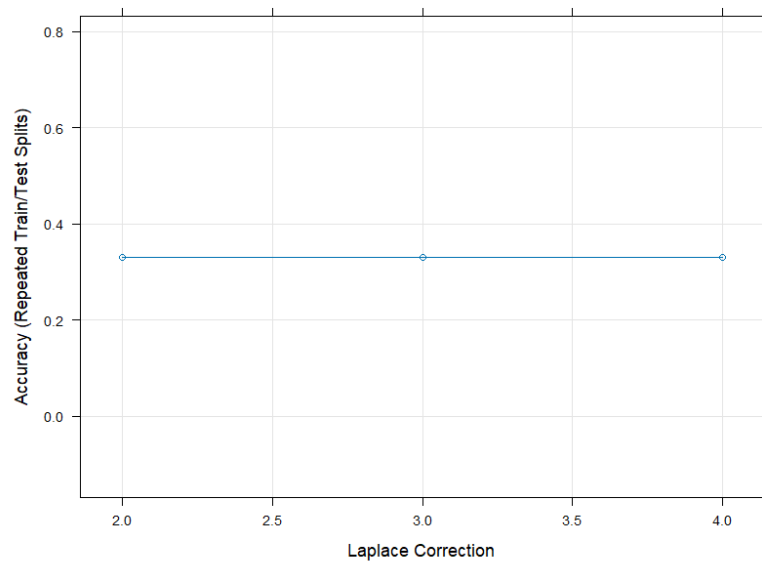


Fig 8: Accuracy for Naïve Bayes Model

Model	Best Tuning Parameter	Training		Testing	
		Kappa	Accuracy	Kappa	Accuracy
MDA	Subclasses=9	0.05298365	0.458571	0.0451	0.4464
RDA	Gamma=0.1, lambda=0.03	0.05904163	0.4884157	0.1194	0.5179
Neural Networks	Size=5, decay=1	0.006727395	0.5200	-0.0208	0.5000
FDA	Degree =1, nprune=4	0.044165306	0.5162173	0.0558	0.4821
SVM	Sigma=0.003923164, C=4	0.0123684544	0.5164286	0.0597	0.5179
KNN	K=15	0.086657887	0.5178571	0.1581	0.5536
Naïve Bayes	2	-0.03444965	0.3314286	0.1667	0.4643

Table 1: Model Summary

Table 1 gives a summary of all the models and their performance on both the training and testing datasets. From the table, KNN is the best performing model on the testing set with an accuracy of 55.36% and Kappa of 0.0579. On the training set, Neural Networks is the best performing model, with an accuracy of 52% and Kappa of 0.006727395, followed closely by KNN with an accuracy of 51.79% on the training set. RDA and SVM are the closest models to KNN on testing with an accuracy of 51.79% for both. Looking at the performance of the models, the best model would be KNN as it has a higher accuracy rate on testing and second-best accuracy on training which makes it the ideal model.

b) Improve classification

From the results of the model performance, it is evident that nonlinear structure of these models helps improve the classification performance. Nonlinear models have an average accuracy of 0.49746 while the linear models had an average accuracy of 0.4776. Average accuracy of the nonlinear models is much higher compared to the linear models.

c) Top five important predictors

KNN: Top 5 Important Predictors

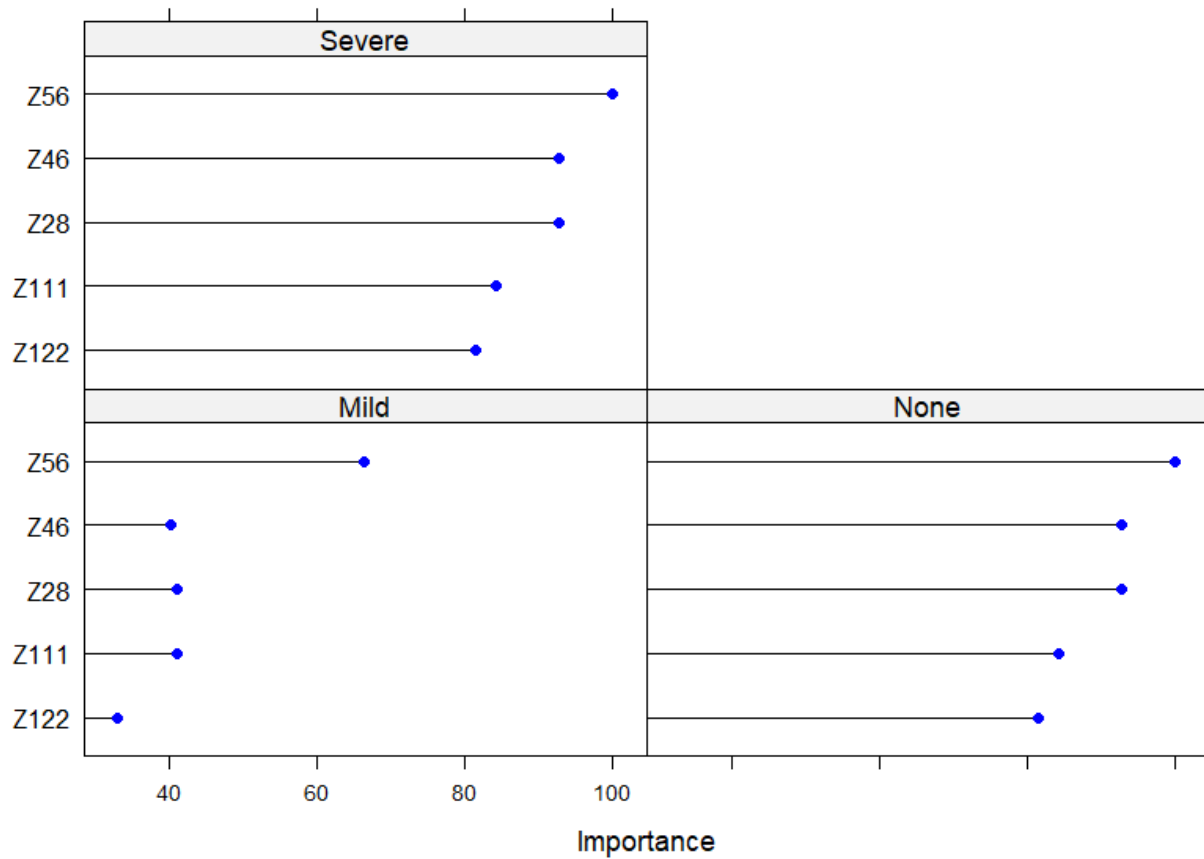


Fig 9: Top 5 important predictors for the best model(KNN)

The table shows the top five most important predictors for the KNN model for the three different classes.

Appendix

#####Question 13.1#####

```
library(mda)
```

```
library(caret)
```

```
library(AppliedPredictiveModeling)
```

```
data(hepatic)
```

```
injury
```

```
table(injury)
```

```
barplot(table(injury), col=c('yellow','green','blue'),main='Class Distribution')
```

```
#part c: Nearzero & Corr
```

```
bio <- bio[, -nearZeroVar(bio)]
```

```
highCorrbio <- findCorrelation(cor(bio), cutoff=0.90)
```

```
bio<- bio[, -highCorrbio]
```

```
#checking for negative values
```

```
negatives_values <- sapply(bio, function(col) any(col < 0))
```

```
print(negatives_values)
```

```
library(e1071)
```

```
skewValues <- apply(bio, 2, skewness)
```

```
head(skewValues)
```

```
#####Transformation#####
```

```
bio_trans <- preProcess(bio, method = c("BoxCox", "center", "scale", "spatialSign"))
```

```
bio_trans
```

```
bio_trsfmd <- predict(bio_trans, bio)
```

```
head(bio_trsfmd)
```

```
set.seed(100)
```

```
trainR <- createDataPartition(injury, p=0.8, list=FALSE)
```

```
X.train <- bio_trsfmd[trainR, ]
```

```
y.train <- injury[trainR]
```

```
X.test <- bio_trsfmd[-trainR, ]
y.test <- injury[-trainR]
ctrl <- trainControl(summaryFunction = multiClassSummary,classProbs = TRUE,
method='LGOCV',savePredictions = TRUE)
```

```
#####Nonlinear Classification Models#####
```

```
#####MDA#####
```

```
library(caret)
set.seed(100)
mdaFit <- caret::train(x = X.train,
                      y = y.train,
                      method = "mda",
                      metric = "Accuracy",
                      tuneGrid = expand.grid(.subclasses = 1:14),
                      trControl = ctrl)
mdaFit
plot(mdaFit)
pred_bio<-predict(mdaFit,X.test)
confusionMatrix(data=pred_bio,
                reference=y.test)
```

```
##RDA##
```

```
install.packages("rda")
install.packages("rrcov")
library(rda)
library(caret)
```

```

set.seed(100)

ctrl <- trainControl(summaryFunction = defaultSummary,
                      classProbs = TRUE)

library(rrcov)

tuneGrid <- expand.grid(
  .gamma = seq(0.1, 1, by = 0.1), # Example values for gamma
  .lambda = seq(0.01, 0.1, by = 0.01) # Example values for lambda
)

rdaFit <- caret::train(X.train,
                       y.train,
                       method = "rda",
                       metric = "Accuracy",
                       tuneGrid = tuneGrid,
                       trControl = ctrl)

rdaFit

plot(rdaFit)

# prediction

predrda <- predict(rdaFit,newdata=X.test)


# confusion Matrix

confusionMatrix(predrda,y.test)


#####NeuralNetworks#####

nnetGrid <- expand.grid(.size = 1:10, .decay = c(0, .1, 1, 2))

maxSize <- max(nnetGrid$.size)

numWts <- (maxSize * (96+ + 1) + (maxSize+1)*3)

nnetFit <- caret::train(x = X.train,

```

```

      y = y.train,
      method = "nnet",
      metric = "Accuracy",
      preProc = c("center", "scale", "spatialSign"),
      tuneGrid = nnetGrid,
      trace = FALSE,
      maxit = 2000,
      MaxNWts = numWts,
      trControl = ctrl)

nnetFit
plot(nnetFit)
nnetpred <- predict(nnetFit,X.test)
nnetpred
confusionMatrix(nnetpred,y.test)

#####FDA#####
library(mda)
library(earth)
marsGrid <- expand.grid(.degree = 1:2, .nprune = 2:38)
fdaTuned <- caret::train(x = X.train,
      y = y.train,
      method = "fda",
      metric="Accuracy",
      tuneGrid = marsGrid,
      trControl = trainControl(method = "cv"))

fdaTuned
plot(fdaTuned)
fdaPred <- predict(fdaTuned, newdata = X.test)

```

```
confusionMatrix(data = fdaPred,reference =y.test)
```

```
#####SVM#####
```

```
set.seed(100)
```

```
library(kernlab)
```

```
library(caret)
```

```
sigmaRangeReduced <- sigest(as.matrix(X.train[,1:96]))
```

```
svmRGridReduced <- expand.grid(.sigma = sigmaRangeReduced[1],  
                              .C = 2^seq(-4, 14))
```

```
svmRModel <- caret::train(x = X.train,  
                          y =y.train,  
                          method = "svmRadial",  
                          metric = "Accuracy",  
                          preProc = c("center", "scale"),  
                          tuneGrid = svmRGridReduced,  
                          fit = FALSE,  
                          trControl = ctrl)
```

```
svmRModel
```

```
plot(svmRModel)
```

```
svmPred <- predict(svmRModel, newdata = X.test)
```

```
confusionMatrix(data = svmPred,reference =y.test)
```

```
#####KNN#####
```

```
library(caret)
```

```
set.seed(100)
```

```
knnFit <- caret::train(x = X.train,
```



```

        y = y.train,
        method = "knn",
        metric = "Accuracy",
        preProc = c("center", "scale"),
        ##tuneGrid = data.frame(.k = c(4*(0:5)+1, 20*(1:5)+1, 50*(2:9)+1)), ## 21 is the
best
        tuneGrid = data.frame(.k = 1:50),
        trControl = ctrl)

knnFit
plot(knnFit)
knnpred <- predict(knnFit,X.test)
knnpred
confusionMatrix(knnpred,
                y.test)

#####Naive Bayes#####
#install.packages("klaR")
# Create a tuning grid for Naive Bayes
nbGrid <- expand.grid(
  .fL=c(2,3,4),
  .usekernel = TRUE,
  .adjust = TRUE
)
library(klaR)
set.seed(100)
nbFit <- caret::train( x = X.train,
        y = y.train,
        method = "nb",
        metric = "Accuracy",
        ## preProc = c("center", "scale"),

```

```

best      ##tuneGrid = data.frame(.k = c(4*(0:5)+1, 20*(1:5)+1, 50*(2:9)+1)), ## 21 is the
tuneGrid = nbGrid,
trControl = ctrl)

nbFit
plot(nbFit)
nbPred <- predict(nbFit, newdata = X.test)
confusionMatrix(data = nbPred,reference =y.test)

####Important Predictors####

knn_ImpVals=varImp(knnFit)
knn_ImpVals
# top 5.

plot(knn_ImpVals,
     top = 5,
     scales = list(y = list(cex = .95)),
     col = "blue",
     main="KNN: Top 5 Important Predictors "
)

```