

STATISTICAL COMPUTING FINAL PROJECT

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MODEL 1: BAGGING MODEL

Model 1 used these three classification ensembles models: * Bagging Model * Gradient Boosting Model * Stacking Model

Libraries Needed for Pre-processing

```
library(readr)
library(dplyr)
library(ggplot2) # for plotting
library(caret) # pre-processing and modeling
library(corrplot)
library(fastDummies) # for creating dummy variables
pacman::p_load(tidyverse)
pacman::p_load(bestNormalize)
```

DATA ACQUISITION AND PRE-PROCESSING

Dataset

radiomics_completedata.csv data was used for this project. The data has a 431 variables and 197 observations including *Failure.binary* as our target/response/outcome variable with its 430 predictors/features/independent variables. *Failure.binary* is a binary variable with 1 and 0 as its values.

```
radiomics = read.csv("radiomics_completedata.csv")
head(radiomics)
```

```
## Institution Failure.binary Failure Entropy_cooc.W.ADC GLNU_align.H.PET
## 1 A 0 49.30000 12.85352 46.25635
## 2 A 1 12.56667 12.21115 27.45454
## 3 A 0 79.80000 12.75682 90.19570
## 4 A 1 17.86667 13.46730 325.64333
## 5 A 0 39.56667 12.63733 89.57904
## 6 A 1 4.76667 13.16159 101.71345
## Min_hist.PET Max_hist.PET Mean_hist.PET Variance_hist.PET
## 1 6.249117 17.825541 9.783773 6.814365
## 2 11.005214 26.469077 15.426640 12.932074
## 3 2.777718 6.877486 4.295330 0.923425
## 4 6.296588 22.029843 10.334779 6.649795
## 5 3.583846 7.922501 4.454175 0.572094
## 6 2.597947 6.206142 3.769041 0.615282
## Standard_Deviation_hist.PET Skewness_hist.PET Kurtosis_hist.PET
## 1 2.612479 0.688533 -0.339727
## 2 3.598298 0.789526 -0.319613
```

## 3	0.962163	0.248637	-0.944246		
## 4	2.580759	0.832011	0.855861		
## 5	0.757225	1.574845	3.250288		
## 6	0.785315	0.610611	-0.090239		
##	Energy_hist.PET	Entropy_hist.PET	AUC_hist.PET	H_suv.PET	Volume.PET
## 1	0.005095	9.629587	0.506553	1.123930	13751.970
## 2	0.006297	8.072951	0.507519	1.927281	9327.705
## 3	0.005015	9.669316	0.503300	0.410573	26624.003
## 4	0.003289	10.574730	0.544274	0.919612	51058.073
## 5	0.008066	7.621834	0.543922	0.306344	29414.553
## 6	0.005237	10.589120	0.507322	0.388752	14240.032
##	X3D_surface.PET	ratio_3ds_vol.PET	ratio_3ds_vol_norm.PET	irregularity.PET	
## 1	5622.519	3.214263	15.91400	2.212137	
## 2	8356.832	4.848032	21.09429	2.348324	
## 3	16832.003	3.163721	19.52154	2.121251	
## 4	29100.294	2.027384	20.12864	1.859572	
## 5	7769.379	4.815431	21.01721	2.219725	
## 6	9563.905	3.699578	18.53249	2.136984	
##	tumor_length.PET	Compactness_v1.PET	Compactness_v2.PET		
## 1	44.04796	0.003366	0.002778		
## 2	39.39796	0.003078	0.002637		
## 3	50.91422	0.003145	0.002664		
## 4	76.23900	0.003118	0.002653		
## 5	36.93490	0.003081	0.002638		
## 6	46.00253	0.003195	0.002687		
##	Spherical_disproportion.PET	Sphericity.PET	Asphericity.PET	Center_of_mass.PET	
## 1	15.91400	0.065378	14.91400	0.811086	
## 2	21.09429	0.049942	20.09429	0.587732	
## 3	19.52154	0.053762	18.52154	0.393189	
## 4	20.12864	0.052217	19.12864	0.866799	
## 5	21.01721	0.050116	20.01721	0.525997	
## 6	18.53249	0.056497	17.53249	0.308017	
##	Max_3D_diam.PET	Major_axis_length.PET	Minor_axis_length.PET		
## 1	44.04796	34.60475	25.88546		
## 2	39.39796	35.13100	27.30539		
## 3	50.91422	48.12896	30.37293		
## 4	76.23900	64.12797	54.46594		
## 5	36.93490	35.99413	23.84296		
## 6	46.00253	42.95117	31.60120		
##	Least_axis_length.PET	Elongation.PET	Flatness.PET	Max_cooc.L.PET	
## 1	24.98484	0.750543	0.724516	0.005020	
## 2	21.15130	0.779759	0.604571	0.008190	
## 3	27.52209	0.633585	0.574348	0.005033	
## 4	51.56490	0.851856	0.806616	0.005971	
## 5	21.38912	0.664919	0.596741	0.007553	
## 6	15.99647	0.738262	0.374927	0.005396	
##	Average_cooc.L.PET	Variance_cooc.L.PET	Entropy_cooc.L.PET	DAVE_cooc.L.PET	
## 1	22.87750	205.6627	10.688721	11.857838	
## 2	21.90654	226.6299	10.291026	13.993568	
## 3	27.25065	208.9461	10.878250	12.281559	
## 4	17.81061	102.6657	10.238635	7.473982	
## 5	15.35938	142.2193	9.829042	10.237690	
## 6	23.34637	181.6257	10.702694	11.660805	
##	DVAR_cooc.L.PET	DENT_cooc.L.PET	SAVE_cooc.L.PET	SVAR_cooc.L.PET	

## 1	84.21646	4.997454	45.75246	587.8808
## 2	129.35103	5.205762	43.81055	581.4143
## 3	85.30680	5.004455	54.49878	599.6980
## 4	43.94774	4.379716	35.61869	310.8875
## 5	79.40248	4.799453	30.71623	384.7110
## 6	87.31571	4.964671	46.69022	503.2667
##	SENT_cooc.L.PET	ASM_cooc.L.PET	Contrast_cooc.L.PET	Dissimilarity_cooc.L.PET
## 1	6.530649	0.003302	234.76478	11.857838
## 2	6.489125	0.003596	325.10017	13.993568
## 3	6.587702	0.003198	236.08136	12.281559
## 4	6.108770	0.003680	99.77033	7.473982
## 5	6.049095	0.004001	184.16098	10.237690
## 6	6.460137	0.003268	223.23109	11.660805
##	Inv_diff_cooc.L.PET	Inv_diff_norm_cooc.L.PET	IDM_cooc.L.PET	
## 1	0.165784		0.858670	0.088949
## 2	0.156018		0.839093	0.085385
## 3	0.154252		0.852986	0.079027
## 4	0.228938		0.904866	0.141631
## 5	0.188717		0.875632	0.108336
## 6	0.166582		0.860102	0.090157
##	IDM_norm_cooc.L.PET	Inv_var_cooc.L.PET	Correlation_cooc.L.PET	
## 1	0.953919	0.091308		0.431777
## 2	0.937653	0.087501		0.285278
## 3	0.952616	0.084629		0.437596
## 4	0.980381	0.149832		0.516631
## 5	0.963872	0.114365		0.355073
## 6	0.955880	0.093295		0.387992
##	Autocorrelation_cooc.L.PET	Tendency_cooc.L.PET	Shade_cooc.L.PET	
## 1	611.5456		587.8808	6860.4448
## 2	543.8667		581.4143	4691.7137
## 3	833.3669		599.6980	403.0883
## 4	369.9095		310.8875	3805.6356
## 5	285.9728		384.7110	9785.4495
## 6	614.9464		503.2667	4106.7640
##	Prominence_cooc.L.PET	IC1_.L.PET	IC2_.L.PET	Coarseness_vdif_.L.PET
## 1	869822.0	-0.083966	0.789572	0.014320
## 2	803734.5	-0.096731	0.814047	0.014196
## 3	800129.8	-0.072366	0.758160	0.016269
## 4	345452.5	-0.050269	0.655209	0.004936
## 5	743501.3	-0.070677	0.727840	0.017239
## 6	708597.7	-0.073872	0.759220	0.016045
##	Contrast_vdif_.L.PET	Busyness_vdif_.L.PET	Complexity_vdif_.L.PET	
## 1	1.021460		0.087378	17053.35
## 2	1.510199		0.080209	21289.19
## 3	1.014169		0.057518	15199.89
## 4	0.306364		0.392674	10762.05
## 5	0.854170		0.081956	16796.63
## 6	0.895212		0.069338	15170.83
##	Strength_vdif_.L.PET	SRE_align.L.PET	LRE_align.L.PET	GLNU_align.L.PET
## 1	27.40494	0.986583	1.070671	10.162131
## 2	35.76496	0.989835	1.057129	8.416510
## 3	24.45341	0.989308	1.057095	9.117958
## 4	5.55092	0.973462	1.129413	94.565775
## 5	57.03783	0.986186	1.069172	10.574675

## 6	26.08534	0.985853	1.070890	10.057347
##	RLNU_align.L.PET	RP_align.L.PET	LGRE_align.L.PET	HGRE_align.L.PET
## 1	383.8912	0.981089	0.063695	590.1484
## 2	263.3486	0.985313	0.065825	560.1103
## 3	394.6779	0.984963	0.039224	781.3663
## 4	2941.3190	0.963661	0.048051	386.6793
## 5	262.4745	0.981101	0.091713	295.6003
## 6	397.9059	0.980630	0.048144	627.3399
##	LGSRE_align.L.PET	HGSRE_align.L.PET	LGHRE_align.L.PET	HGLRE_align.L.PET
## 1	0.062491	580.5855	0.068738	631.5734
## 2	0.064212	554.5346	0.072438	583.5148
## 3	0.038778	768.0350	0.041011	836.1597
## 4	0.046564	376.9558	0.054360	428.3121
## 5	0.090222	292.3243	0.097821	308.7154
## 6	0.047408	618.2607	0.051089	665.2563
##	GLNU_norm_align.L.PET	RLNU_norm_align.L.PET	GLVAR_align.L.PET	
## 1	0.027914	0.961445	201.5094	
## 2	0.033437	0.969710	214.6379	
## 3	0.024834	0.968128	216.6109	
## 4	0.032318	0.928789	107.6866	
## 5	0.041113	0.960224	121.3562	
## 6	0.026718	0.959459	187.2442	
##	RLVAR_align.L.PET	Entropy_align.L.PET	SZSE.L.PET	LZSE.L.PET
## 1	0.025908	5.586143	0.926936	1.384001
## 2	0.021453	5.385714	0.961338	1.244838
## 3	0.020843	5.702830	0.974475	1.114749
## 4	0.046375	5.480351	0.905696	1.617562
## 5	0.024509	5.053054	0.966013	1.148597
## 6	0.025153	5.622598	0.936782	1.322943
##	HGLZE.L.PET	SZLGE.L.PET	SZHGE.L.PET	LZLGE.L.PET
## 1	592.5775	0.056127	553.5787	0.089951
## 2	566.7718	0.060570	546.1829	0.086532
## 3	769.6933	0.040391	735.9377	0.040694
## 4	393.5484	0.043346	360.6300	0.076789
## 5	300.9426	0.091138	295.8022	0.101787
## 6	617.0878	0.041385	567.5274	0.065899
##	GLNU_area.L.PET			
## 1	9.166018			
## 2	7.817915			
## 3	8.877842			
## 4	83.352565			
## 5	10.245976			
## 6	9.390127			
##	ZSNU.L.PET	ZSP.L.PET	GLNU_norm.L.PET	ZSNU_norm.L.PET
## 1	301.1987	0.899841	0.027499	0.823228
## 2	233.4102	0.941158	0.032589	0.900252
## 3	372.1247	0.966472	0.024663	0.930516
## 4	2206.3053	0.860538	0.031941	0.781042
## 5	242.2684	0.956101	0.040895	0.909893
## 6	325.9069	0.913118	0.026787	0.844660
##	GLVAR_area.L.PET			
## 1	201.7881			
## 2	213.9100			
## 3	216.4466			
## 4	109.9100			
## 5	123.6639			
## 6	184.6198			
##	ZSVAR.L.PET	Entropy_area.L.PET	Max_cooc.H.PET	Average_cooc.H.PET
## 1	0.142022	5.886187	0.031232	39.87474
## 2	0.109793	5.546278	0.043568	39.22729
## 3	0.038537	5.775912	0.169447	44.90994
## 4	0.259194	5.901957	0.040212	38.15816
## 5	0.048849	5.156114	0.423535	49.45276
## 6	0.116919	5.851581	0.217884	46.26425
##	Variance_cooc.H.PET	Entropy_cooc.H.PET	DAVE_cooc.H.PET	DVAR_cooc.H.PET
## 1	255.25108	6.344137	13.397288	131.6433
## 2	259.22064	7.168339	14.938851	146.5065
## 3	226.94291	3.662030	11.817845	143.8888

## 4	276.46636	6.205163	12.489582	129.5153
## 5	65.47745	2.835302	6.261891	56.9727
## 6	174.57711	3.122212	10.059360	134.1508
##	DENT_cooc.H.PET	SAVE_cooc.H.PET	SVAR_cooc.H.PET	SENT_cooc.H.PET
## 1	4.528843	79.74696	769.9364	5.285948
## 2	2.880112	75.45206	667.2773	5.693972
## 3	4.354173	89.81735	824.2760	3.057425
## 4	4.257568	76.31379	820.4186	5.186241
## 5	3.891832	98.90299	765.7524	2.360339
## 6	1.916625	92.52596	463.0127	2.599031
##	ASM_cooc.H.PET	Contrast_cooc.H.PET	Dissemblarity_cooc.H.PET	
## 1	0.017558	311.0628	13.397288	
## 2	0.012079	369.6002	14.938851	
## 3	0.096088	283.4905	11.817845	
## 4	0.020168	285.4418	12.489582	
## 5	0.233933	96.1523	6.261891	
## 6	0.146959	235.2907	10.059360	
##	Inv_diff_cooc.H.PET	Inv_diff_norm_cooc.H.PET	IDM_cooc.H.PET	
## 1	0.240428	0.846191	0.181276	
## 2	0.198536	0.831014	0.137656	
## 3	0.439712	0.866805	0.405377	
## 4	0.279879	0.856139	0.224079	
## 5	0.576561	0.923498	0.543300	
## 6	0.516123	0.886644	0.485744	
##	IDM_norm_cooc.H.PET	Inv_var_cooc.H.PET	Correlation_cooc.H.PET	
## 1	0.940222	0.030684	0.393202	
## 2	0.929828	0.032006	0.289621	
## 3	0.944553	0.011773	0.377943	
## 4	0.945253	0.032706	0.486297	
## 5	0.980482	0.021087	0.268281	
## 6	0.953100	0.009811	0.328640	
##	Autocorrelation_cooc.H.PET	Tendency_cooc.H.PET	Shade_cooc.H.PET	
## 1	1689.514	709.9364	-2209.927	
## 2	1613.004	667.2773	-4195.799	
## 3	2101.874	624.2760	-4303.802	
## 4	1589.599	820.4186	-5395.462	
## 5	2462.728	165.7524	1099.232	
## 6	2197.079	463.0127	-2285.992	
##	Prominence_cooc.H.PET	IC1_d.H.PET	IC2_d.H.PET	Coarseness_vdif.H.PET
## 1	1028531.31	-0.043805	0.512217	0.004319
## 2	957339.84	-0.023569	0.418010	0.005180
## 3	729696.02	-0.063791	0.473698	0.003375
## 4	1434052.83	-0.069422	0.611279	0.002825
## 5	55971.88	-0.044636	0.360145	0.003902
## 6	381561.77	-0.056410	0.417972	0.003199
##	Contrast_vdif.H.PET	Busyness_vdif.H.PET	Complexity_vdif.H.PET	
## 1	49.10863	0.141647	25517.13	
## 2	28.26579	0.103194	28339.01	
## 3	220.66779	0.236919	24028.42	
## 4	40.72831	0.833266	23437.94	
## 5	32.04753	0.124684	15279.35	
## 6	271.03091	0.279836	22773.21	
##	Strength_vdif.H.PET	SRE_align.H.PET	LRE_align.H.PET	RLNU_align.H.PET
## 1	19.64713	0.917833	1.449477	291.82356

## 2	25.47241	0.953059	1.241419	227.49063		
## 3	22.15293	0.774121	2.674531	165.69391		
## 4	2.79079	0.880393	1.732322	2033.70698		
## 5	53.29819	0.741090	2.918639	99.23077		
## 6	21.85351	0.720078	3.392842	140.39293		
##	RP_align.H.PET LGRE_align.H.PET HGRE_align.H.PET LGSRE_align.H.PET					
## 1	0.888556	0.004341	1569.763	0.004198		
## 2	0.935326	0.004349	1536.186	0.004223		
## 3	0.710370	0.003527	1821.062	0.003336		
## 4	0.839415	0.005339	1588.246	0.005019		
## 5	0.684948	0.002975	2476.679	0.002849		
## 6	0.656286	0.003229	2111.778	0.003040		
##	HGSRE_align.H.PET LGHRE_align.H.PET HGLRE_align.H.PET GLNU_norm_align.H.PET					
## 1	1433.081	0.005120	2278.993	0.130158		
## 2	1472.727	0.004991	1836.812	0.108781		
## 3	1318.500	0.004849	5694.966	0.309012		
## 4	1388.818	0.007300	2734.362	0.120339		
## 5	1889.628	0.003929	6544.325	0.470904		
## 6	1501.696	0.004877	7061.132	0.374988		
##	RLNU_norm_align.H.PET GLVAR_align.H.PET RLVAR_align.H.PET Entropy_align.H.PET					
## 1	0.805658	271.94120	0.166759	3.665844		
## 2	0.881876	263.05257	0.089416	3.807145		
## 3	0.559747	231.23849	0.633026	2.962910		
## 4	0.733600	302.00409	0.279758	3.963763		
## 5	0.516961	63.36076	0.708711	2.615080		
## 6	0.492823	187.63061	0.894173	2.953297		
##	SZSE.H.PET LZSE.H.PET LGLZE.H.PET HGLZE.H.PET SZLGE.H.PET SZHGE.H.PET					
## 1	0.729896	6.346008	0.004206	1945.242	0.003751	1205.4141
## 2	0.889774	1.945761	0.004294	1541.326	0.004071	1371.5287
## 3	0.543152	38.343615	0.003595	1869.824	0.003145	833.9286
## 4	0.686000	28.192087	0.005281	2614.722	0.004412	1088.6316
## 5	0.494282	85.120177	0.002930	2778.032	0.002719	1427.6154
## 6	0.494144	151.989372	0.003258	2079.108	0.002893	988.7421
##	LZLGE.H.PET LZHGE.H.PET GLNU_area.H.PET ZSNU.H.PET ZSP.H.PET GLNU_norm.H.PET					
## 1	0.014967	9278.763	28.21123	112.61992	0.564877	0.125177
## 2	0.007054	2730.177	23.91083	171.00253	0.829245	0.106933
## 3	0.027806	99597.669	42.33586	36.25834	0.312626	0.330695
## 4	0.066848	39940.885	160.59767	604.01684	0.425782	0.117405
## 5	0.047180	166256.576	23.73782	17.00253	0.245387	0.351578
## 6	0.115459	288928.476	28.02885	17.76569	0.181354	0.371297
##	ZSNU_norm.H.PET GLVAR_area.H.PET ZSVAR.H.PET Entropy_area.H.PET					
## 1	0.492171	263.01858	3.183797	4.580974		
## 2	0.749255	257.55868	0.482612	4.158935		
## 3	0.283583	218.15517	27.944240	4.080320		
## 4	0.434586	309.53854	22.609920	5.086907		
## 5	0.252530	70.97225	68.165160	3.954518		
## 6	0.236256	205.12926	120.717731	4.002762		
##	Max_cooc.W.PET Average_cooc.W.PET Variance_cooc.W.PET Entropy_cooc.W.PET					
## 1	0.013277	8.741717	27.724284	8.310617		
## 2	0.015738	10.946398	54.254568	8.954940		
## 3	0.046074	4.019422	3.648015	5.580950		
## 4	0.013915	9.152454	25.597213	8.286935		
## 5	0.116685	2.577872	2.729045	4.706665		
## 6	0.063098	3.127779	2.391005	5.013592		

##	DAVE_cooc.W.PET	DVAR_cooc.W.PET	DENT_cooc.W.PET	SAVE_cooc.W.PET
## 1	4.361115	12.870015	3.611785	17.480905
## 2	6.845926	31.128005	4.224171	21.890266
## 3	1.595373	1.629296	2.279633	8.036314
## 4	3.728549	11.060383	3.431589	18.302378
## 5	1.376959	1.728999	2.205393	5.153215
## 6	1.306368	1.277859	2.076037	6.253029
##	SVAR_cooc.W.PET	SENT_cooc.W.PET	ASM_cooc.W.PET	Contrast_cooc.W.PET
## 1	79.024802	5.099087	0.006555	31.867274
## 2	139.053134	5.483416	0.005298	77.960077
## 3	10.420558	3.676978	0.027061	4.166444
## 4	77.440194	5.106053	0.007012	24.943599
## 5	7.293066	3.190894	0.061557	3.618055
## 6	6.581107	3.336839	0.041094	2.977854
##	Dissimilarity_cooc.W.PET	Inv_diff_cooc.W.PET	Inv_diff_norm_cooc.W.PET	
## 1	4.361115	0.306285	0.861048	
## 2	6.845926	0.244001	0.837985	
## 3	1.595373	0.503481	0.863798	
## 4	3.728549	0.343449	0.905179	
## 5	1.376959	0.558453	0.882471	
## 6	1.306368	0.553594	0.874095	
##	IDM_cooc.W.PET	IDM_norm_cooc.W.PET	Inv_var_cooc.W.PET	Correlation_cooc.W.PET
## 1	0.213874	0.955388	0.224294	0.427805
## 2	0.158456	0.936467	0.164222	0.284054
## 3	0.439777	0.957440	0.421156	0.431424
## 4	0.254836	0.980367	0.261941	0.515299
## 5	0.509374	0.964322	0.439330	0.339500
## 6	0.504966	0.961979	0.468899	0.379680
##	Autocorrelation_cooc.W.PET	Tendency_cooc.W.PET	Shade_cooc.W.PET	
## 1	88.165309	79.024802	341.143402	
## 2	135.044039	139.053134	552.913441	
## 3	17.701479	10.420558	2.361775	
## 4	96.847788	77.440194	471.374078	
## 5	7.553672	7.293066	26.823935	
## 6	10.670526	6.581107	7.170907	
##	Prominence_cooc.W.PET	IC1_d.W.PET	IC2_d.W.PET	Coarseness_vdif.W.PET
## 1	15813.1737	-0.042283	0.565302	0.015034
## 2	45767.4163	-0.044029	0.591913	0.015811
## 3	242.8423	-0.052987	0.524822	0.017811
## 4	21312.7505	-0.056187	0.630354	0.004934
## 5	276.1447	-0.033151	0.398878	0.018221
## 6	124.4042	-0.044775	0.466821	0.017235
##	Contrast_vdif.W.PET	Busyness_vdif.W.PET	Complexity_vdif.W.PET	
## 1	0.294464	0.717283	869.48613	
## 2	0.599158	0.420854	2313.88985	
## 3	0.112568	2.860859	40.08855	
## 4	0.133588	1.549091	1346.28621	
## 5	0.078944	3.650188	44.97271	
## 6	0.079545	4.181398	27.61148	
##	Strength_vdif.W.PET	SRE_align.W.PET	LRE_align.W.PET	GLNU_align.W.PET
## 1	3.919855	0.961787	1.191350	24.97624
## 2	8.341981	0.977438	1.116168	14.88136
## 3	0.511453	0.889821	1.618702	53.72505
## 4	1.384522	0.943354	1.291573	179.17215

## 5	1.109636	0.876250	1.674603	59.72108
## 6	0.444774	0.863194	1.800706	67.44333
##	RLNU_align.W.PET	RP_align.W.PET	LGRE_align.W.PET	HGRE_align.W.PET
## 1	347.5995	0.947236	0.150278	85.345885
## 2	250.6373	0.968373	0.127690	139.175484
## 3	265.0196	0.853307	0.272808	15.983362
## 4	2609.2747	0.922696	0.092857	101.288786
## 5	170.2453	0.840992	0.466475	7.937118
## 6	245.9412	0.822440	0.339659	10.636341
##	LGSRE_align.W.PET	HGSRE_align.W.PET	LGHRE_align.W.PET	HGLRE_align.W.PET
## 1	0.144360	82.365395	0.178628	98.96776
## 2	0.122525	136.722689	0.150485	150.71592
## 3	0.245883	13.790048	0.414898	28.12741
## 4	0.087782	95.978334	0.117784	126.22675
## 5	0.401364	7.231352	0.833918	11.22377
## 6	0.297964	9.120687	0.601806	18.69612
##	GLNU_norm_align.W.PET	RLNU_norm_align.W.PET	GLVAR_align.W.PET	
## 1	0.067162	0.901536	27.361255	
## 2	0.058138	0.938874	51.482886	
## 3	0.154351	0.749487	3.691659	
## 4	0.061479	0.859819	27.190856	
## 5	0.256845	0.724823	2.405984	
## 6	0.196000	0.702794	2.523334	
##	RLVAR_align.W.PET	Entropy_align.W.PET	SZSE.W.PET	LZSE.W.PET
## 1	0.069370	4.413771	0.862196	2.111226
## 2	0.043126	4.601911	0.939019	1.436265
## 3	0.229632	3.470022	0.737823	5.821460
## 4	0.107059	4.683410	0.816094	3.396694
## 5	0.239812	2.974484	0.688181	6.186741
## 6	0.289495	3.306066	0.662526	12.143891
##	HGLZE.W.PET	SZLGE.W.PET	SZHGE.W.PET	LZLGE.W.PET
## 1	88.918679	0.112325	79.094274	0.392257
## 2	138.464377	0.116457	128.987889	0.195656
## 3	14.973723	0.247502	10.310508	1.043890
## 4	106.496868	0.073436	88.831921	0.286957
## 5	9.015688	0.284427	6.692377	3.360406
## 6	10.745985	0.252353	6.482655	5.046844
##	GLNU_area.W.PET	LZHGE.W.PET	GLNU_area.W.PET	
## 1	88.918679	0.112325	79.094274	0.392257
## 2	138.464377	0.116457	128.987889	0.195656
## 3	14.973723	0.247502	10.310508	1.043890
## 4	106.496868	0.073436	88.831921	0.286957
## 5	9.015688	0.284427	6.692377	3.360406
## 6	10.745985	0.252353	6.482655	5.046844
##	ZSNU.W.PET	ZSP.W.PET	GLNU_norm.W.PET	ZSNU_norm.W.PET
## 1	224.38141	0.789816	0.065066	0.699359
## 2	211.55675	0.901447	0.056642	0.852145
## 3	121.85027	0.586665	0.160280	0.503961
## 4	1419.26821	0.697656	0.059662	0.620677
## 5	66.31832	0.545387	0.232966	0.438818
## 6	77.07583	0.451942	0.195918	0.406055
##	GLVAR_area.W.PET			
## 1	224.38141	0.789816	0.065066	0.699359
## 2	211.55675	0.901447	0.056642	0.852145
## 3	121.85027	0.586665	0.160280	0.503961
## 4	1419.26821	0.697656	0.059662	0.620677
## 5	66.31832	0.545387	0.232966	0.438818
## 6	77.07583	0.451942	0.195918	0.406055
##	ZSVAR.W.PET	Entropy_area.W.PET	Min_hist.ADC	Max_hist.ADC
## 1	0.497852	4.937916	549.00253	2268.003
## 2	0.198720	4.834988	0.00253	2211.003
## 3	2.890741	4.143192	634.00253	2860.003
## 4	1.327156	5.449999	0.00253	2869.003
## 5	2.793389	3.991207	0.00253	2389.003
## 6	7.192684	4.330361	0.00253	2498.003
##	Mean_hist.ADC			
## 1	0.497852	4.937916	549.00253	2268.003
## 2	0.198720	4.834988	0.00253	2211.003
## 3	2.890741	4.143192	634.00253	2860.003
## 4	1.327156	5.449999	0.00253	2869.003
## 5	2.793389	3.991207	0.00253	2389.003
## 6	7.192684	4.330361	0.00253	2498.003
##	Variance_hist.ADC	Standard_Deviation_hist.ADC	Skewness_hist.ADC	
## 1	113473.17		336.8603	1.05752
## 2	83953.26		289.7494	-0.49105

## 3	193194.07		439.5410	1.53649	
## 4	132561.08		364.0919	0.24067	
## 5	110268.35		332.0693	0.31916	
## 6	276984.10		526.2953	-0.19996	
##	Kurtosis_hist.ADC	Energy_hist.ADC	Entropy_hist.ADC	AUC_hist.ADC	Volume.ADC
## 1	0.39978	0.00757	7.72697	0.52307	14702.81
## 2	1.41215	0.00503	8.82392	0.49147	11850.17
## 3	2.15473	0.00426	9.42564	0.56722	26067.89
## 4	0.23359	0.00365	10.02927	0.52148	51577.90
## 5	0.50069	0.00454	9.12787	0.50458	27419.14
## 6	-1.03080	0.00413	9.41989	0.49047	16131.31
##	X3D_surface.ADC	ratio_3ds_vol.ADC	ratio_3ds_vol_norm.ADC	irregularity.ADC	
## 1	2621.908	0.39370	1.52762		1.93975
## 2	3814.097	0.27791	1.37006		1.76130
## 3	5638.645	0.21884	1.32876		1.57930
## 4	11033.100	0.21644	1.64907		1.63673
## 5	5670.769	0.22562	1.35892		1.61457
## 6	6099.528	0.30552	1.70690		1.72859
##	Compactness_v1.ADC	Compactness_v2.ADC	Spherical_disproportion.ADC		
## 1	0.03070	0.28444	1.52762		
## 2	0.03570	0.39354	1.37006		
## 3	0.03727	0.43122	1.32876		
## 4	0.02764	0.22655	1.64907		
## 5	0.03611	0.40326	1.35892		
## 6	0.02637	0.20451	1.70690		
##	Sphericity.ADC	Asphericity.ADC	Center_of_mass.ADC	Max_3D_diam.ADC	
## 1	0.65823	0.52762	0.97407	46.80855	
## 2	0.73378	0.37006	1.00173	57.64178	
## 3	0.75655	0.32876	1.48789	64.07496	
## 4	0.60987	0.64907	1.32794	85.02235	
## 5	0.73978	0.35892	0.57983	59.88998	
## 6	0.58926	0.70690	1.60559	66.42410	
##	Major_axis_length.ADC	Minor_axis_length.ADC	Least_axis_length.ADC		
## 1	45.53640	20.24517	13.58989		
## 2	35.07877	28.70241	23.63536		
## 3	42.14714	36.72698	25.93458		
## 4	58.00549	42.98623	35.06326		
## 5	39.28351	35.40209	31.13508		
## 6	52.01087	34.53146	21.82211		
##	Elongation.ADC	Flatness.ADC	Max_cooc.L.ADC	Average_cooc.L.ADC	
## 1	0.44709	0.30093	0.01362	24.26969	
## 2	0.82074	0.67629	0.00769	34.15443	
## 3	0.87392	0.61784	0.00984	17.40595	
## 4	0.74359	0.60699	0.00893	26.20041	
## 5	0.90372	0.79509	0.00863	27.03123	
## 6	0.66644	0.42207	0.00548	33.31549	
##	Variance_cooc.L.ADC	Entropy_cooc.L.ADC	DAVE_cooc.L.ADC	DVAR_cooc.L.ADC	
## 1	135.95808	9.35172	9.33833	95.10941	
## 2	60.59539	9.52569	6.58341	31.97649	
## 3	159.14565	9.93157	8.05607	81.58702	
## 4	57.02199	9.50974	5.46198	23.67951	
## 5	65.76514	9.76494	6.96837	33.58727	
## 6	176.68232	10.64861	9.13371	70.36682	
##	DENT_cooc.L.ADC	SAVE_cooc.L.ADC	SVAR_cooc.L.ADC	SENT_cooc.L.ADC	

## 1	4.68745	48.53685	361.5607	4.49616
## 2	4.18551	68.30632	167.0920	2.32433
## 3	4.48343	34.80936	490.1310	5.16708
## 4	3.95039	52.39829	174.5978	4.55938
## 5	4.26293	54.05993	180.9453	4.48500
## 6	4.65758	66.62846	552.9789	3.08233
##	ASM_cooc.L.ADC Contrast_cooc.L.ADC Dissimilarity_cooc.L.ADC			
## 1	0.00535	182.26652	9.33833	
## 2	0.00448	75.28447	6.58341	
## 3	0.00458	146.44656	8.05607	
## 4	0.00454	53.48506	5.46198	
## 5	0.00414	82.11021	6.96837	
## 6	0.00338	153.74529	9.13371	
##	Inv_diff_cooc.L.ADC Inv_diff_norm_cooc.L.ADC IDM_cooc.L.ADC			
## 1	0.23569	0.88844	0.15619	
## 2	0.24103	0.91456	0.15044	
## 3	0.24921	0.90225	0.16496	
## 4	0.27847	0.92805	0.18834	
## 5	0.23450	0.90993	0.14567	
## 6	0.20980	0.88787	0.12604	
##	IDM_norm_cooc.L.ADC Inv_var_cooc.L.ADC Correlation_cooc.L.ADC			
## 1	0.96528	0.15633	0.33222	
## 2	0.98542	0.15887	0.38132	
## 3	0.97276	0.17144	0.54243	
## 4	0.99019	0.19368	0.53355	
## 5	0.98376	0.15283	0.37826	
## 6	0.96963	0.13018	0.56744	
##	Autocorrelation_.L.ADC Tendency_cooc.L.ADC Shade_.L.ADC Prominence_cooc.L.ADC			
## 1	633.7211	361.5607	7639.8939	517154.08
## 2	1189.3065	167.0920	-1156.8109	112937.29
## 3	388.8025	490.1310	17093.4493	1296059.93
## 4	716.6097	174.5978	616.3283	88605.95
## 5	755.2618	180.9453	592.0947	113320.37
## 6	1209.5645	552.9789	-1837.1897	590287.94
##	IC1_.L.ADC IC2_.L.ADC Coarseness_vdif_.L.ADC Contrast_vdif_.L.ADC			
## 1	-0.11842	0.83912	0.02135	0.71307
## 2	-0.05061	0.63924	0.01258	0.23808
## 3	-0.07274	0.73740	0.00784	0.40394
## 4	-0.06200	0.68774	0.00556	0.15512
## 5	-0.04812	0.63329	0.01085	0.27967
## 6	-0.09225	0.81078	0.01042	0.60161
##	Busyness_vdif_.L.ADC Complexity_vdif_.L.ADC Strength_vdif_.L.ADC			
## 1	0.04811	8748.919	30.44366	
## 2	0.05243	5213.433	10.85376	
## 3	0.21602	9811.189	12.83805	
## 4	0.20181	4912.319	3.52728	
## 5	0.08515	5705.778	8.31391	
## 6	0.06946	8974.106	10.09240	
##	SRE_align.L.ADC LRE_align.L.ADC GLNU_align.L.ADC RLNU_align.L.ADC			
## 1	0.97677	1.11587	9.40856	232.7602
## 2	0.97564	1.11803	26.43616	645.9593
## 3	0.96919	1.14834	43.70925	1177.5699
## 4	0.96126	1.18592	102.31243	2562.1046
## 5	0.97703	1.11715	28.40221	788.2562

## 6	0.98211	1.08986	21.25471	890.8892		
##	RP_align.L.ADC	LGRE_align.L.ADC	HGRE_align.L.ADC	LGSRE_align.L.ADC		
## 1	0.96871	0.00908	831.5410	0.00900		
## 2	0.96669	0.00605	1191.1595	0.00602		
## 3	0.95823	0.01361	487.9258	0.01321		
## 4	0.94795	0.00810	786.0107	0.00784		
## 5	0.96795	0.00721	833.8975	0.00716		
## 6	0.97551	0.00591	1362.5846	0.00587		
##	HGSRE_align.L.ADC	LGHRE_align.L.ADC	HGLRE_align.L.ADC	GLNU_norm_align.L.ADC		
## 1	820.9252	0.00946	876.2823	0.04038		
## 2	1157.5280	0.00615	1335.5219	0.04066		
## 3	478.4817	0.01531	528.1310	0.03656		
## 4	757.7992	0.00954	909.4492	0.03841		
## 5	815.1979	0.00741	917.7657	0.03626		
## 6	1335.9421	0.00607	1478.8704	0.02516		
##	RLNU_norm_align.L.ADC	GLVAR_align.L.ADC	RLVAR_align.L.ADC	Entropy_align.L.ADC		
## 1	0.93826	154.93296	0.04141	5.29371		
## 2	0.93411	69.45486	0.04188	5.17751		
## 3	0.91877	156.30297	0.05240	5.47452		
## 4	0.90022	64.98946	0.06534	5.31012		
## 5	0.93819	78.05347	0.04295	5.30441		
## 6	0.95061	175.82591	0.03219	5.74239		
##	SZSE.L.ADC	LZSE.L.ADC	LGLZE.L.ADC	HGLZE.L.ADC	SZLGE.L.ADC	SZHGE.L.ADC
## 1	0.93703	1.33159	0.00927	858.5837	0.00905	831.8537
## 2	0.92448	1.39444	0.00624	1184.8610	0.00617	1086.4222
## 3	0.87706	1.82170	0.01338	514.4899	0.01189	468.7768
## 4	0.90217	1.59820	0.00767	792.5723	0.00686	720.2240
## 5	0.91279	1.55603	0.00757	833.3315	0.00743	760.6074
## 6	0.93634	1.29245	0.00606	1348.0807	0.00598	1247.0381
##	LZLGE.L.ADC	LZHGE.L.ADC	GLNU_area.L.ADC	ZSNU.L.ADC	ZSP.L.ADC	GLNU_norm.L.ADC
## 1	0.01042	981.8102	8.25894	197.1051	0.91304	0.03781
## 2	0.00662	1681.2171	24.10984	524.4053	0.89683	0.04002
## 3	0.02376	734.9103	34.98083	798.7819	0.82545	0.03416
## 4	0.01300	1204.1618	90.93063	1994.0215	0.86029	0.03768
## 5	0.00840	1283.7978	24.73040	600.5032	0.87065	0.03520
## 6	0.00644	1779.7534	19.65712	741.6164	0.91756	0.02479
##	ZSNU_norm.L.ADC	GLVAR_area.L.ADC	ZSVAR.L.ADC	Entropy_area.L.ADC		
## 1	0.84485	158.37071	0.12535	5.53926		
## 2	0.81809	71.19097	0.14408	5.46224		
## 3	0.72475	157.77185	0.34501	6.00431		
## 4	0.77331	66.76247	0.23904	5.67242		
## 5	0.79579	82.41219	0.22912	5.69671		
## 6	0.84241	176.08461	0.09810	6.01150		
##	Max_cooc.H.ADC	Average_cooc.H.ADC	Variance_cooc.H.ADC	Entropy_cooc.H.ADC		
## 1	0.00464	29.95976	310.9790	11.72265		
## 2	0.00420	33.61846	312.8265	11.35537		
## 3	0.00622	30.58315	335.7248	11.53210		
## 4	0.00461	30.75681	310.6464	11.60919		
## 5	0.00393	31.26939	305.7453	11.56749		
## 6	0.00496	30.52540	330.9954	11.34674		
##	DAVE_cooc.H.ADC	DVAR_cooc.H.ADC	DENT_cooc.H.ADC	SAVE_cooc.H.ADC		
## 1	15.71847	162.7022	5.37436	59.91700		
## 2	15.39980	148.1637	5.34697	67.23440		
## 3	13.82367	148.1751	5.24052	61.16377		

## 4	12.67796	118.4962	5.12061	61.51110
## 5	15.22805	152.9835	5.34969	62.53624
## 6	12.68957	134.2114	5.12373	61.04826
##	SVAR_cooc.H.ADC	SENT_cooc.H.ADC	ASM_cooc.H.ADC	Contrast_cooc.H.ADC
## 1	834.2180	3.87272	0.00312	409.6931
## 2	866.0614	3.21841	0.00292	385.2396
## 3	1003.6953	3.81762	0.00296	339.1990
## 4	963.4178	3.73436	0.00290	279.1628
## 5	838.1762	3.61892	0.00291	384.8001
## 6	1028.8043	3.58842	0.00300	295.1723
##	Dissimilarity_cooc.H.ADC	Inv_diff_cooc.H.ADC	Inv_diff_norm_cooc.H.ADC	
## 1	15.71847	0.14449	0.82408	
## 2	15.39980	0.13871	0.82594	
## 3	13.82367	0.16711	0.84276	
## 4	12.67796	0.16941	0.85215	
## 5	15.22805	0.14798	0.82834	
## 6	12.68957	0.17461	0.85365	
##	IDM_cooc.H.ADC	IDM_norm_cooc.H.ADC	Inv_var_cooc.H.ADC	Correlation_cooc.H.ADC
## 1	0.07807	0.92422	0.08536	0.34381
## 2	0.06993	0.92757	0.07472	0.38679
## 3	0.09608	0.93697	0.09823	0.49736
## 4	0.09588	0.94673	0.09742	0.55321
## 5	0.07991	0.92793	0.08208	0.37325
## 6	0.09995	0.94500	0.10213	0.55665
##	Autocorrelation_cooc.H.ADC	Tendency_cooc.H.ADC	Shade_cooc.H.ADC	
## 1	1003.570	834.2180	4888.58538	
## 2	1250.239	866.0614	-4080.74039	
## 3	1101.301	1003.6953	7361.25628	
## 4	1116.892	963.4178	2723.56893	
## 5	1090.963	838.1762	-98.86912	
## 6	1115.056	1028.8043	509.16337	
##	Prominence_cooc.H.ADC	IC1_d.H.ADC	IC2_d.H.ADC	Coarseness_vdif.H.ADC
## 1	1518300	-0.15943	0.92667	0.02421
## 2	1589114	-0.05988	0.72703	0.01048
## 3	2077405	-0.06514	0.74687	0.00767
## 4	1824192	-0.05338	0.70043	0.00496
## 5	1538643	-0.05818	0.72034	0.00898
## 6	1971550	-0.09605	0.83415	0.00994
##	Contrast_vdif.H.ADC	Busyness_vdif.H.ADC	Complexity_vdif.H.ADC	
## 1	1.85757	0.03586	16806.66	
## 2	1.80534	0.09301	16186.56	
## 3	1.49359	0.14284	13464.93	
## 4	1.41213	0.29907	12641.54	
## 5	1.83534	0.11398	16384.39	
## 6	1.45238	0.09984	12914.39	
##	Strength_vdif.H.ADC	SRE_align.H.ADC	LRE_align.H.ADC	GLNU_align.H.ADC
## 1	29.66079	0.99220	1.04664	4.07230
## 2	10.90410	0.99123	1.04949	11.31108
## 3	7.03589	0.98442	1.08787	20.88959
## 4	3.31909	0.98263	1.08821	46.68109
## 5	8.83863	0.98826	1.06328	13.66324
## 6	10.10115	0.98601	1.07180	15.03108
##	RLNU_align.H.ADC	RP_align.H.ADC	LGRE_align.H.ADC	HGRE_align.H.ADC
## 1	246.9236	0.98876	0.02752	1363.457

## 2	687.6470	0.98755	0.02717	1357.005		
## 3	1249.7235	0.97718	0.02776	1343.165		
## 4	2786.7832	0.97588	0.02638	1359.587		
## 5	824.1350	0.98339	0.02668	1358.525		
## 6	904.6320	0.98077	0.02665	1361.936		
##	LGSRE_align.H.ADC	HGSRE_align.H.ADC	LGHRE_align.H.ADC	HGLRE_align.H.ADC		
## 1	0.02695	1349.190	0.02979	1430.871		
## 2	0.02648	1340.025	0.02994	1430.336		
## 3	0.02707	1310.372	0.03080	1516.790		
## 4	0.02500	1334.267	0.03403	1466.691		
## 5	0.02577	1338.937	0.03083	1444.863		
## 6	0.02539	1342.295	0.03304	1444.991		
##	GLNU_norm_align.H.ADC	RLNU_norm_align.H.ADC	GLVAR_align.H.ADC			
## 1	0.01859		0.97614	329.5023		
## 2	0.01850		0.97320	329.3505		
## 3	0.01848		0.95625	325.6524		
## 4	0.01843		0.95150	327.9251		
## 5	0.01850		0.96576	329.3047		
## 6	0.01845		0.96023	327.5799		
##	RLVAR_align.H.ADC	Entropy_align.H.ADC	SZSE.H.ADC	LZSE.H.ADC	LGLZE.H.ADC	
## 1	0.01753	6.01510	0.96829	1.15763	0.02871	
## 2	0.01839	6.04615	0.96505	1.15896	0.02661	
## 3	0.03446	6.10308	0.93628	1.65499	0.02502	
## 4	0.03209	6.13418	0.95168	1.26414	0.02388	
## 5	0.02330	6.06848	0.95866	1.24670	0.02517	
## 6	0.02577	6.08534	0.94459	1.30242	0.02141	
##	HGLZE.H.ADC	SZLGE.H.ADC	SZHGE.H.ADC	LZLGE.H.ADC	LZHGE.H.ADC	GLNU_area.H.ADC
## 1	1353.052	0.02838	1303.023	0.03004	1618.472	3.99028
## 2	1355.552	0.02483	1302.738	0.03376	1584.380	10.95282
## 3	1293.549	0.02152	1196.086	0.04888	2953.476	19.42358
## 4	1353.634	0.02049	1283.290	0.04521	1725.853	44.63370
## 5	1328.345	0.02373	1252.666	0.04477	1783.557	13.08842
## 6	1363.271	0.01864	1280.446	0.06832	1716.544	14.22012
##	ZSNU.H.ADC	ZSP.H.ADC	GLNU_norm.H.ADC	ZSNU_norm.H.ADC	GLVAR_area.H.ADC	
## 1	223.9086	0.95584	0.01881	0.91643	324.0822	
## 2	619.2862	0.95385	0.01854	0.90792	327.6186	
## 3	1007.9399	0.89316	0.01876	0.84458	305.6363	
## 4	2450.9039	0.93025	0.01848	0.87848	321.4979	
## 5	727.4123	0.93716	0.01859	0.89506	324.1160	
## 6	762.1457	0.92170	0.01856	0.86177	315.8327	
##	ZSVAR.H.ADC	Entropy_area.H.ADC	Max_cooc.W.ADC	Average_cooc.W.ADC		
## 1	0.05727	6.06723	0.00675	65.37977		
## 2	0.05401	6.18594	0.00382	118.60405		
## 3	0.39430	6.37088	0.00376	60.27417		
## 4	0.10225	6.32299	0.00302	117.52784		
## 5	0.10193	6.21756	0.00355	101.18139		
## 6	0.11881	6.31556	0.00343	130.61014		
##	Variance_cooc.W.ADC	DAVE_cooc.W.ADC	DVAR_cooc.W.ADC	DENT_cooc.W.ADC		
## 1	1010.0875	25.43812	706.5272	6.06338		
## 2	746.1691	23.15154	390.8192	5.94785		
## 3	1991.6618	28.49457	1018.7085	6.25261		
## 4	1181.5174	24.91785	487.4797	6.07963		
## 5	945.7911	26.38488	481.6157	6.14012		
## 6	2779.9243	36.21365	1103.2759	6.60187		

##	SAVE_cooc.W.ADC	SVAR_cooc.W.ADC	SENT_cooc.W.ADC	ASM_cooc.W.ADC
## 1	130.7570	2686.849	5.54316	0.00323
## 2	237.2056	2057.975	2.77584	0.00280
## 3	120.5458	6136.137	6.76239	0.00275
## 4	235.0531	3617.812	6.13864	0.00265
## 5	202.3602	2605.515	5.80987	0.00273
## 6	261.2178	8705.171	3.87339	0.00266
##	Contrast_cooc.W.ADC	Dissimilarity_cooc.W.ADC	Inv_diff_cooc.W.ADC	
## 1	1353.496	25.43812	0.12826	
## 2	926.696	23.15154	0.10420	
## 3	1830.505	28.49457	0.10990	
## 4	1108.253	24.91785	0.10456	
## 5	1177.644	26.38488	0.09861	
## 6	2414.521	36.21365	0.08344	
##	Inv_diff_norm_cooc.W.ADC	IDM_cooc.W.ADC	IDM_norm_cooc.W.ADC	
## 1	0.88720	0.06987	0.96438	
## 2	0.91342	0.04700	0.98505	
## 3	0.90097	0.05282	0.97202	
## 4	0.92684	0.04905	0.98983	
## 5	0.90880	0.04557	0.98327	
## 6	0.88642	0.03696	0.96879	
##	Inv_var_cooc.W.ADC	Correlation_cooc.W.ADC	Autocorrelation_cooc.W.ADC	
## 1	0.07218	0.33254	4607.525	
## 2	0.04790	0.38156	14349.142	
## 3	0.05640	0.54299	4709.081	
## 4	0.04962	0.53354	14439.590	
## 5	0.04452	0.37996	10594.131	
## 6	0.03757	0.56825	18631.013	
##	Tendency_cooc.W.ADC	Shade_cooc.W.ADC	Prominence_cooc.W.ADC	IC1_d.W.ADC
## 1	2686.849	154504.57	28492973	-0.20561
## 2	2057.975	-49857.50	17100002	-0.13210
## 3	6136.137	755229.72	202604689	-0.13981
## 4	3617.812	57995.75	38091821	-0.08828
## 5	2605.515	31890.26	23457384	-0.13836
## 6	8705.171	-113889.96	146542333	-0.23037
##	IC2_d.W.ADC	Coarseness_vdif.W.ADC	Contrast_vdif.W.ADC	Busyness_vdif.W.ADC
## 1	0.96152	0.01818	4.78265	0.01774
## 2	0.91270	0.01162	1.49489	0.00979
## 3	0.92904	0.00742	1.99390	0.02744
## 4	0.85241	0.00544	1.11708	0.01846
## 5	0.92596	0.01002	1.72379	0.01257
## 6	0.98684	0.00959	3.20701	0.00873
##	Complexity_vdif.W.ADC	Strength_vdif.W.ADC	SRE_align.W.ADC	LRE_align.W.ADC
## 1	94483.95	120.21874	0.99193	1.04495
## 2	123984.35	70.45906	0.99469	1.03484
## 3	322896.60	118.12334	0.99389	1.03917
## 4	270786.27	41.10745	0.99307	1.04143
## 5	183481.75	68.98942	0.99446	1.03681
## 6	408132.18	116.30778	0.99699	1.02540
##	GLNU_align.W.ADC	RLNU_align.W.ADC	RP_align.W.ADC	LGRE_align.W.ADC
## 1	4.26622	246.5777	0.98876	0.00683
## 2	8.60033	696.8829	0.99205	0.00418
## 3	13.91071	1298.3291	0.99080	0.00430
## 4	24.38419	2904.1988	0.98991	0.00579

## 5	8.43212	844.4260	0.99159	0.00400		
## 6	6.05624	944.0342	0.99512	0.00374		
##	HGRE_align.W.ADC	LGSRE_align.W.ADC	HGSRE_align.W.ADC	LGHRE_align.W.ADC		
## 1	5992.756	0.00683	5952.927	0.00685		
## 2	14395.425	0.00418	14281.115	0.00418		
## 3	5853.808	0.00429	5824.143	0.00434		
## 4	15776.936	0.00562	15649.652	0.00681		
## 5	11683.555	0.00400	11599.962	0.00400		
## 6	21008.240	0.00373	20894.393	0.00374		
##	HGLRE_align.W.ADC	GLNU_norm_align.W.ADC	RLNU_norm_align.W.ADC			
## 1	6152.074	0.01935	0.97502			
## 2	14868.922	0.01462	0.98198			
## 3	5983.117	0.01300	0.97996			
## 4	16293.667	0.01072	0.97773			
## 5	12044.998	0.01230	0.98150			
## 6	21478.153	0.00885	0.98802			
##	GLVAR_align.W.ADC	RLVAR_align.W.ADC	Entropy_align.W.ADC	SZSE.W.ADC	LZSE.W.ADC	
## 1	1139.4041	0.01629	6.94511	0.98460	1.07424	
## 2	842.8456	0.01345	6.67452	0.96527	1.11797	
## 3	1938.7178	0.01519	6.79621	0.98765	1.17872	
## 4	1327.6869	0.01562	7.20649	0.98060	1.10239	
## 5	1109.3728	0.01437	6.95074	0.97667	1.13245	
## 6	2767.6284	0.01027	7.49193	0.98323	1.08445	
##	LGLZE.W.ADC	HGLZE.W.ADC	SZLGE.W.ADC	SZHGE.W.ADC	LZLGE.W.ADC	LZHGE.W.ADC
## 1	0.00686	6055.150	0.00686	6018.454	0.00690	6201.935
## 2	0.00422	14407.506	0.00422	14026.413	0.00423	16054.013
## 3	0.00433	5883.686	0.00430	5711.245	0.00453	6674.638
## 4	0.00511	15809.845	0.00455	15506.485	0.00888	17172.910
## 5	0.00403	11663.603	0.00403	11366.888	0.00405	13231.943
## 6	0.00376	20996.110	0.00375	20573.429	0.00377	22707.428
##	GLNU_area.W.ADC	ZSNU.W.ADC	ZSP.W.ADC	GLNU_norm.W.ADC	ZSNU_norm.W.ADC	
## 1	4.13400	239.2894	0.97918	0.01899	0.95586	
## 2	8.37627	644.7370	0.95637	0.01461	0.93288	
## 3	13.11686	1165.7026	0.97268	0.02501	0.91537	
## 4	23.84726	2760.4129	0.97203	0.01069	0.94658	
## 5	8.14437	784.5973	0.96469	0.02526	0.93769	
## 6	5.93657	893.1791	0.97662	0.00884	0.95272	
##	GLVAR_area.W.ADC	ZSVAR.W.ADC	Entropy_area.W.ADC			
## 1	1145.1050	0.02586	6.28632			
## 2	847.5254	0.04153	6.77853			
## 3	1923.8571	0.07104	7.15685			
## 4	1329.9529	0.03848	7.29521			
## 5	1116.3867	0.05223	7.05149			
## 6	2743.2376	0.03055	7.54787			

Data description

```
str(radiomics[1:10])
```

```
## 'data.frame':  197 obs. of  10 variables:
## $ Institution      : chr  "A" "A" "A" "A" ...
## $ Failure.binary   : int   0 1 0 1 0 1 0 0 1 1 ...
## $ Failure          : num   49.3 12.6 79.8 17.9 39.6 ...
## $ Entropy_cooc.W.ADC : num   12.9 12.2 12.8 13.5 12.6 ...
## $ GLNU_align.H.PET  : num   46.3 27.5 90.2 325.6 89.6 ...
```

```
## $ Min_hist.PET          : num  6.25 11.01 2.78 6.3 3.58 ...
## $ Max_hist.PET          : num  17.83 26.47 6.88 22.03 7.92 ...
## $ Mean_hist.PET         : num   9.78 15.43 4.3 10.33 4.45 ...
## $ Variance_hist.PET     : num   6.814 12.932 0.923 6.65 0.572 ...
## $ Standard_Deviation_hist.PET: num   2.612 3.598 0.962 2.581 0.757 ...
```

Variables *Institution* and *Failure.binary* needs to change its datatype into factors. *Institution* was re-coded into dummy variables

```
newdf = dummy_cols(radiomics, select_columns = "Institution" )
newdf = newdf[,-1]
str(newdf[431:434])
```

```
## 'data.frame':   197 obs. of  4 variables:
## $ Institution_A: int   1 1 1 1 1 1 1 1 1 1 ...
## $ Institution_B: int   0 0 0 0 0 0 0 0 0 0 ...
## $ Institution_C: int   0 0 0 0 0 0 0 0 0 0 ...
## $ Institution_D: int   0 0 0 0 0 0 0 0 0 0 ...
```

```
newdf$Institution_A = as.factor(newdf$Institution_A)
newdf$Institution_B = as.factor(newdf$Institution_B)
newdf$Institution_C = as.factor(newdf$Institution_C)
newdf$Institution_D = as.factor(newdf$Institution_D)
newdf$Failure.binary = as.factor(newdf$Failure.binary)
str(newdf[1])
```

```
## 'data.frame':   197 obs. of  1 variable:
## $ Failure.binary: Factor w/ 2 levels "0","1": 1 2 1 2 1 2 1 1 2 2 ...
```

Null Values

```
sum(is.na(newdf))
```

```
## [1] 0
```

No null values found in the data.

Normality of the data

To check the normality of the data, we used Shapiro-wilk test.

SHAPIRO-WILK TEST OF NORMALITY H_0 = Data is normally distributed H_i = Data is not normally distributed

```
lshap <- lapply(newdf[2:430], shapiro.test) #applying shapiro-wilk test of normality to the data frame
l = 1:429
for (i in l) { #
  x = lshap[[i]]$p.value
  z= 1 +i
  if (lshap[[i]]$p.value >= 0.05){
    print(paste(x, "p-value for",colnames(newdf[z]) ))
  }
}
```

```
## [1] "0.135011298030283 p-value for Entropy_cooc.W.ADC"
```

Upon checking its normality, only the *Entropy_cooc.W.ADC* has p-value of 0.135 which tells that it follows normal distribution, the rest of the variables do not follow normal distribution. Transformation is required to the data to address the non-normality of the data. In this project we use *Ordered Quantile (ORQ) normalization transformation* to normalize the data.


```
newdf1 = newdf %>% select_if(is.numeric)
tempDF=apply(newdf1,2,orderNorm)
tempDF=lapply(tempDF, function(x) x$x.t)
tempDF=tempDF%>%as.data.frame()
norm_data = cbind(newdf[c('Failure.binary','Institution_A','Institution_B','Institution_C','Institution_D')],tempDF)
```

Testing the normality of the data using the normalize data using **Ordered Quantile (ORQ) normalization transformation**.

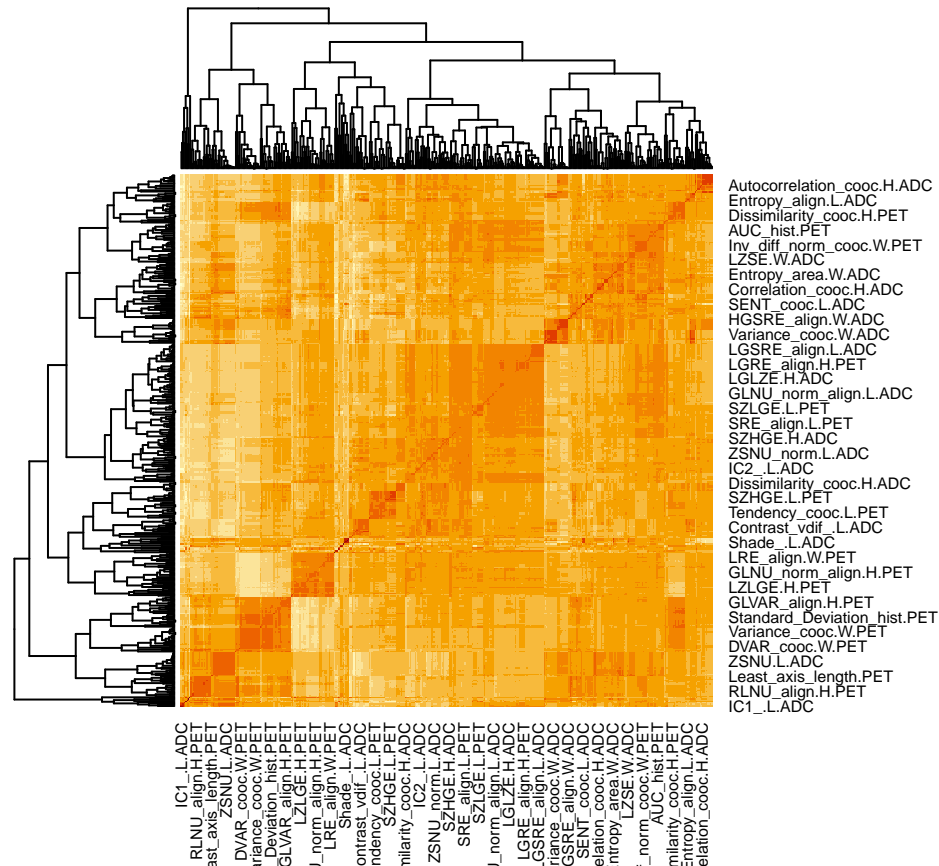
```
lshap <- lapply(tempDF, shapiro.test) #applying shapiro-wilk test of normality to the data frame
l = 1:429
for (i in l) { #
  x = lshap[[i]]$p.value
  z= 1 +i
  if (lshap[[i]]$p.value <= 0.05){
    print(paste(x, "p-value for",colnames(newdf[z]) ))
  }
}
```

Since all p-values are greater than 0.05 then we cannot reject the null hypothesis that the data by this time follows Normal distribution.

Correlation

Checking the correlation of the data. By assumption that highly correlated data does not do good in model making, thus by removing this will improve the performance of our model. But in this project, removing highly correlated columns is not considered.

```
heatmap(corMatrix)
```



```
#ggcorrplot(corr <= 0.40 ,colors = c("blue", "white", "red"), ggtheme=theme_bw, lab = FALSE, title = "C
```

Data Splitting

After pre-processing, we now split our data into 80% for training and 20% for testing. * **Training:** our training dataset has a 158 observations * **Testing:** our testing dataset has a 39 observations

```
set.seed(3333)
trainIndex <- createDataPartition(norm_data$Failure.binary, p = .80,
                                   list = FALSE,
                                   times = 1)
finaldata_train<- norm_data[ trainIndex,]
finaldata_test<- norm_data[-trainIndex,]
dim(finaldata_train) ; dim(finaldata_test)
```

```
## [1] 158 434
```

```
## [1] 39 434
```

DATA MODELING

Libraries Needed for Modeling

```
library(rpart)      # for fitting decision trees
library(ipred)      # for fitting bagged decision trees
library(doParallel) # for parallel backend to foreach
library(foreach)    # for parallel processing with for loops
```

Bagging Model

1. This model uses bagging() function from ipred package. In this model we used 100 iterations to create 100 bootstrapped samples.

```
bagging_1 <- bagging(  
  formula = Failure.binary ~ .,  
  data = finaldata_train,  
  nbagg = 100, # number of iteration to be included in the model.  
  coob = TRUE, # OOB error rate  
  control = rpart.control(minsplit = 2, cp = 0)  
)  
  
bagging_1
```

```
##  
## Bagging classification trees with 100 bootstrap replications  
##  
## Call: bagging.data.frame(formula = Failure.binary ~ ., data = finaldata_train,  
##      nbagg = 100, coob = TRUE, control = rpart.control(minsplit = 2,  
##      cp = 0))  
##  
## Out-of-bag estimate of misclassification error: 0.1203
```

From the output above, we obtain the Out-of-bag estimate of misclassification error of 0.1203. This means that model bagging_1 has a tendency of 12% misclassifying error in classifying the Failure.binary variable.

2. This model applied bagging from caret package with 10-fold cross-validation. The model has an accuracy of 0.875 which is good enough.

```
bagging_2 <- train(  
  Failure.binary ~ .,  
  data = finaldata_train,  
  method = "treebag",  
  trControl = trainControl(method = "cv", number = 10),  
  nbagg = 200,  
  control = rpart.control(minsplit = 2, cp = 0)  
)  
  
bagging_2
```

```
## Bagged CART  
##  
## 158 samples  
## 433 predictors  
## 2 classes: '0', '1'  
##  
## No pre-processing  
## Resampling: Cross-Validated (10 fold)  
## Summary of sample sizes: 141, 141, 143, 142, 142, 143, ...  
## Resampling results:  
##  
## Accuracy Kappa  
## 0.8800735 0.7309916
```

The model has accuracy of 88% in classifying the Failure.binary variable. To choose the best model we try to test the models performance on test set.

```

library(performance)
library(ROCR)

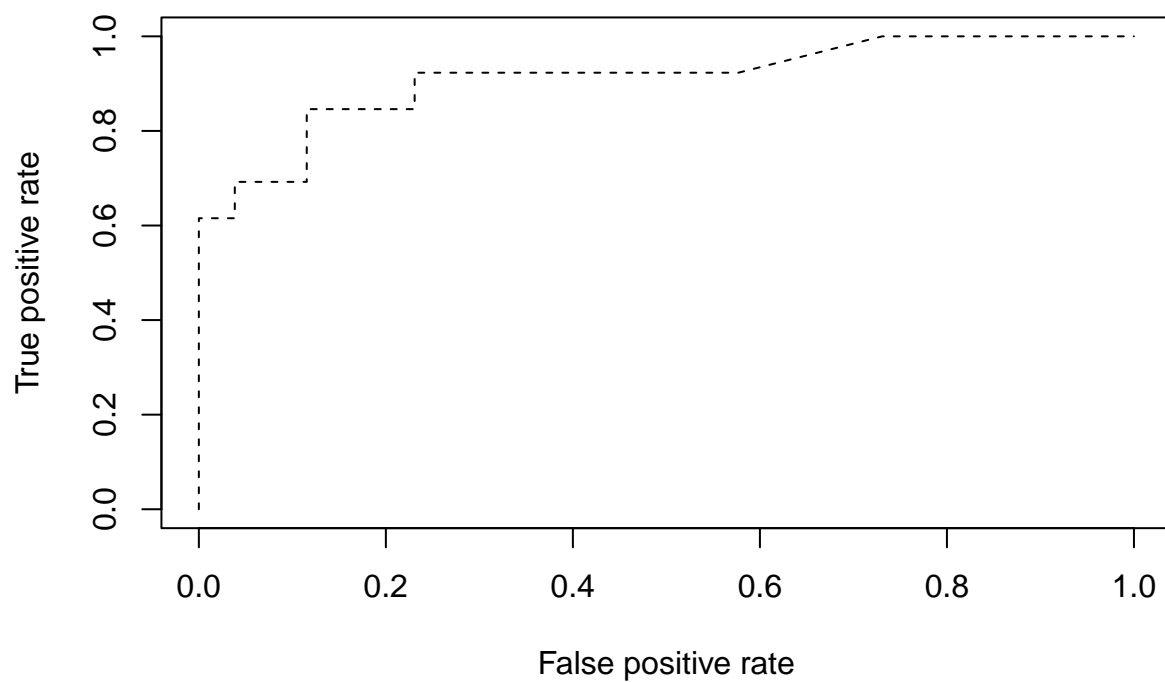
##
## Attaching package: 'ROCR'
## The following object is masked from 'package:performance':
##
##      performance
library(pROC)

## Type 'citation("pROC")' for a citation.
##
## Attaching package: 'pROC'
## The following objects are masked from 'package:stats':
##
##      cov, smooth, var
# Compute predicted probabilities on training data
m1_prob <- predict(bagging_1, finaldata_test, type = "prob")[,2]

# Compute AUC metrics for cv_model1,2 and 3
perf1 <- prediction(m1_prob,finaldata_test$Failure.binary) %>%
  performance(measure = "tpr", x.measure = "fpr")

# Plot ROC curves for cv_model1,2 and 3
plot(perf1, col = "black", lty = 2)

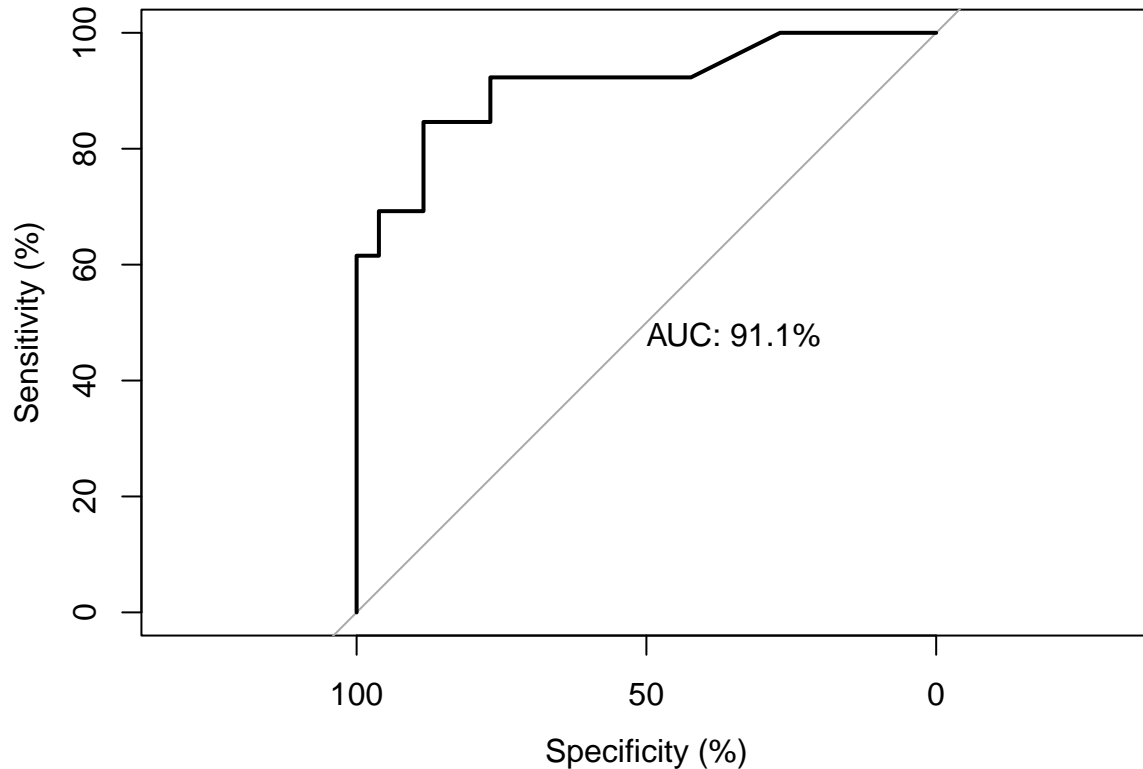
```



```
# ROC plot for training data
roc( finaldata_test$Failure.binary ~ m1_prob, plot=TRUE, legacy.axes=FALSE,
      percent=TRUE, col="black", lwd=2, print.auc=TRUE)
```

```
## Setting levels: control = 0, case = 1
```

```
## Setting direction: controls < cases
```



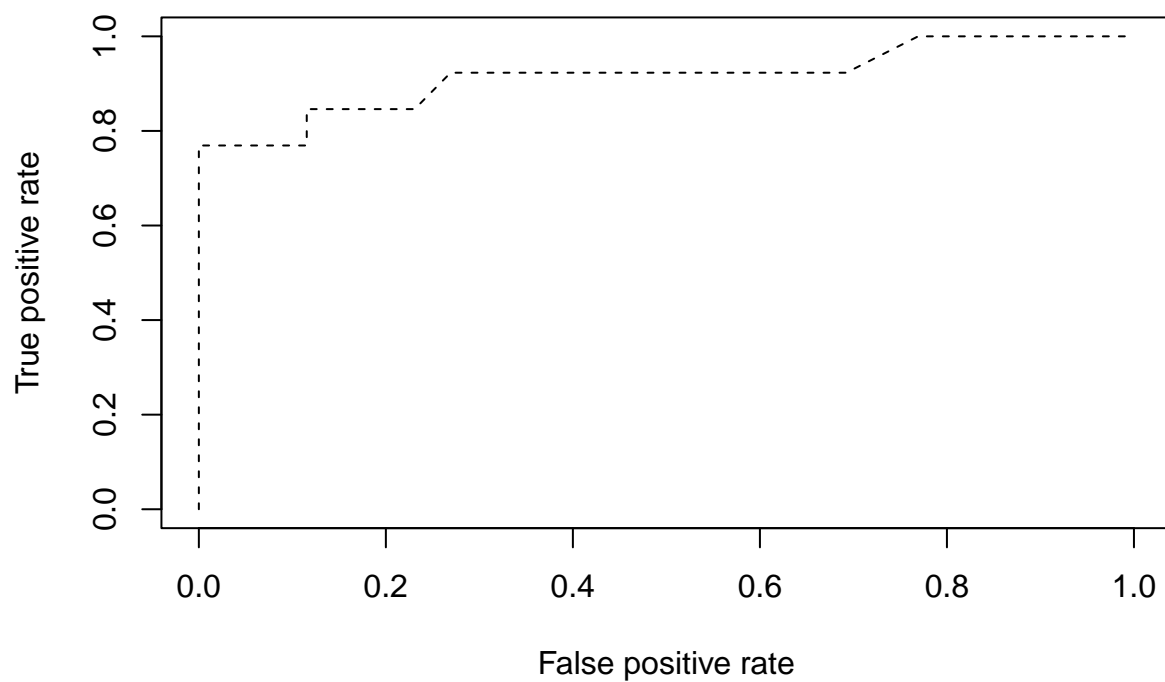
```
##
## Call:
## roc.formula(formula = finaldata_test$Failure.binary ~ m1_prob,      plot = TRUE, legacy.axes = FALSE,
##
## Data: m1_prob in 26 controls (finaldata_test$Failure.binary 0) < 13 cases (finaldata_test$Failure.bi
## Area under the curve: 91.12%
```

Testing our model `bagging_1` to the testing set is helpful to determine the performance of our model and its sensitivity and specificity. The model has an AUC of 91.1%, which means it is a better model with 91.1% of the time the model will correctly classify a random case.

```
# Compute predicted probabilities on training data
m1_prob <- predict(bagging_2, finaldata_test, type = "prob")[,2]

# Compute AUC metrics for cv_model1,2 and 3
perf1 <- prediction(m1_prob,finaldata_test$Failure.binary) %>%
  performance(measure = "tpr", x.measure = "fpr")

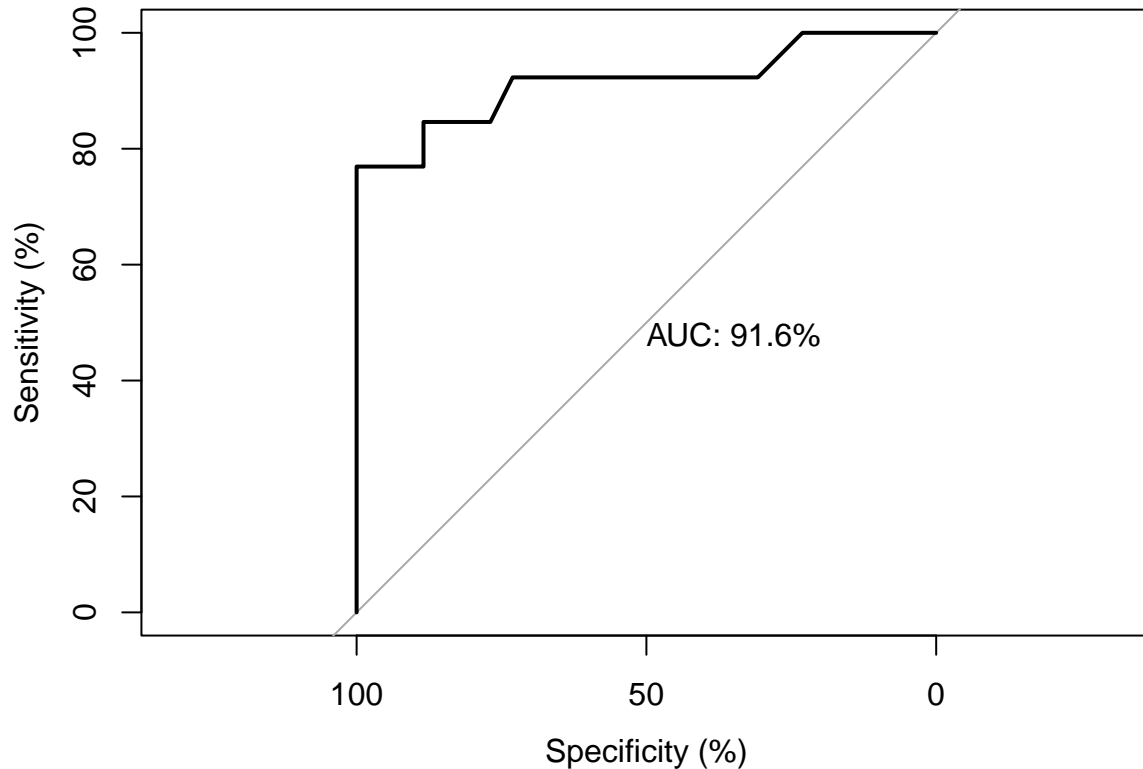
# Plot ROC curves for cv_model1,2 and 3
plot(perf1, col = "black", lty = 2)
```



```
# ROC plot for training data
roc( finaldata_test$Failure.binary ~ m1_prob, plot=TRUE, legacy.axes=FALSE,
      percent=TRUE, col="black", lwd=2, print.auc=TRUE)
```

```
## Setting levels: control = 0, case = 1
```

```
## Setting direction: controls < cases
```



```
##
## Call:
## roc.formula(formula = finaldata_test$Failure.binary ~ m1_prob,      plot = TRUE, legacy.axes = FALSE,
##
## Data: m1_prob in 26 controls (finaldata_test$Failure.binary 0) < 13 cases (finaldata_test$Failure.bi
## Area under the curve: 91.57%
```

Plotting the performance of the model bagging_2 will help us to tell how our model performed in predicting the target variable (Failure.binary). We test the performance of the model on the training set and found out that the model performed well on our testing set with 0.5% higher than the AUC of previous model, thus we can conclude that model bagging_2 is our final model in bagging.

Checking it on confusion matrix,

```
predictions = predict(bagging_2, newdata = finaldata_test)
confusionMatrix(data = finaldata_test$Failure.binary, predictions )
```

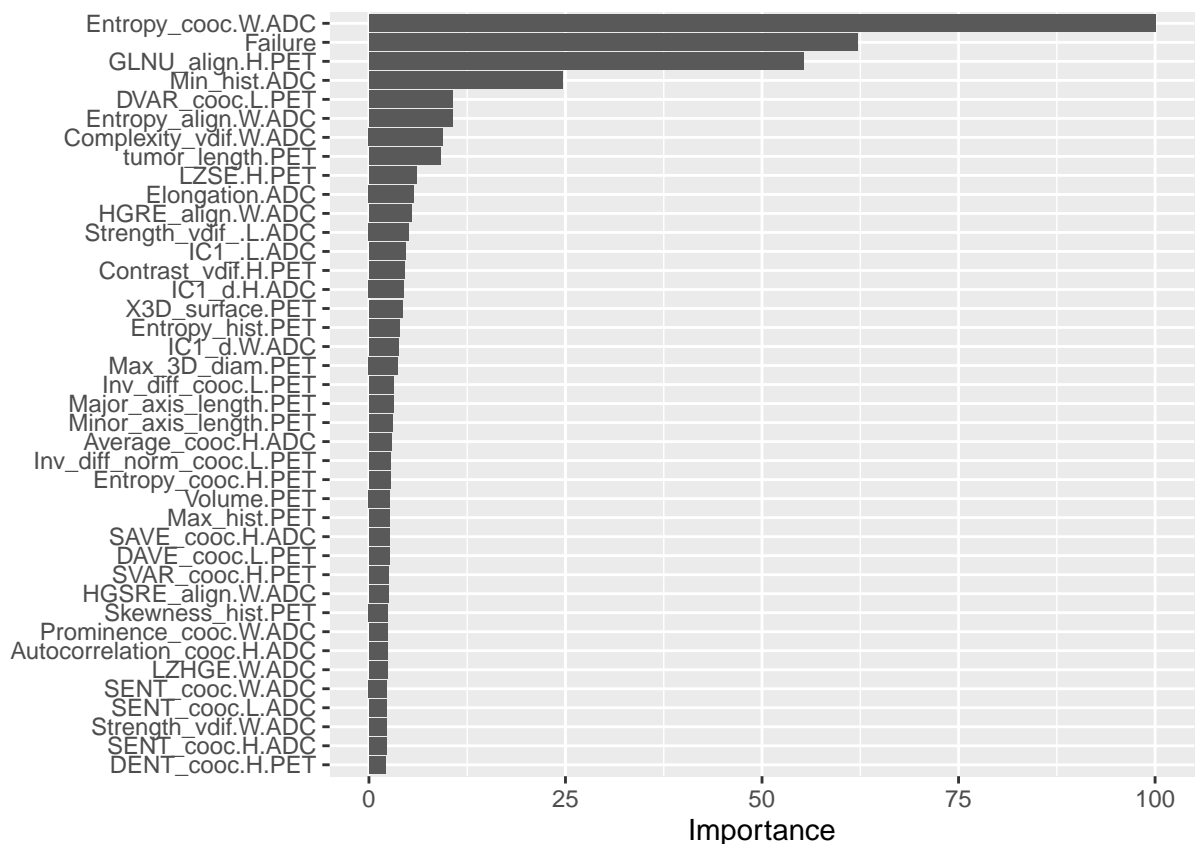
```
## Confusion Matrix and Statistics
##
##           Reference
## Prediction  0   1
##           0 23   3
##           1   2 11
##
##               Accuracy : 0.8718
##               95% CI   : (0.7257, 0.957)
##           No Information Rate : 0.641
##           P-Value [Acc > NIR] : 0.001224
```



```
##
##          Kappa : 0.717
##
## Mcnemar's Test P-Value : 1.000000
##
##          Sensitivity : 0.9200
##          Specificity : 0.7857
##          Pos Pred Value : 0.8846
##          Neg Pred Value : 0.8462
##          Prevalence : 0.6410
##          Detection Rate : 0.5897
##          Detection Prevalence : 0.6667
##          Balanced Accuracy : 0.8529
##
##          'Positive' Class : 0
##
```

According to confusion matrix, 23 of the predicted 0 is correctly classified by the model and 11 of the predicted 1 are classified correctly by the model only 5 were misclassified.

```
vip::vip(bagging_2, num_features = 40)
```



Based on model bagging_2, Entropy_cooc.W.ADC, Failure and GLNU_align.H.PET are the top important variables that helps the model to make an accurate prediction on the classification of any random case inputted in the model.

```
p1 <- pdp::partial(
  bagging_2,
```

```

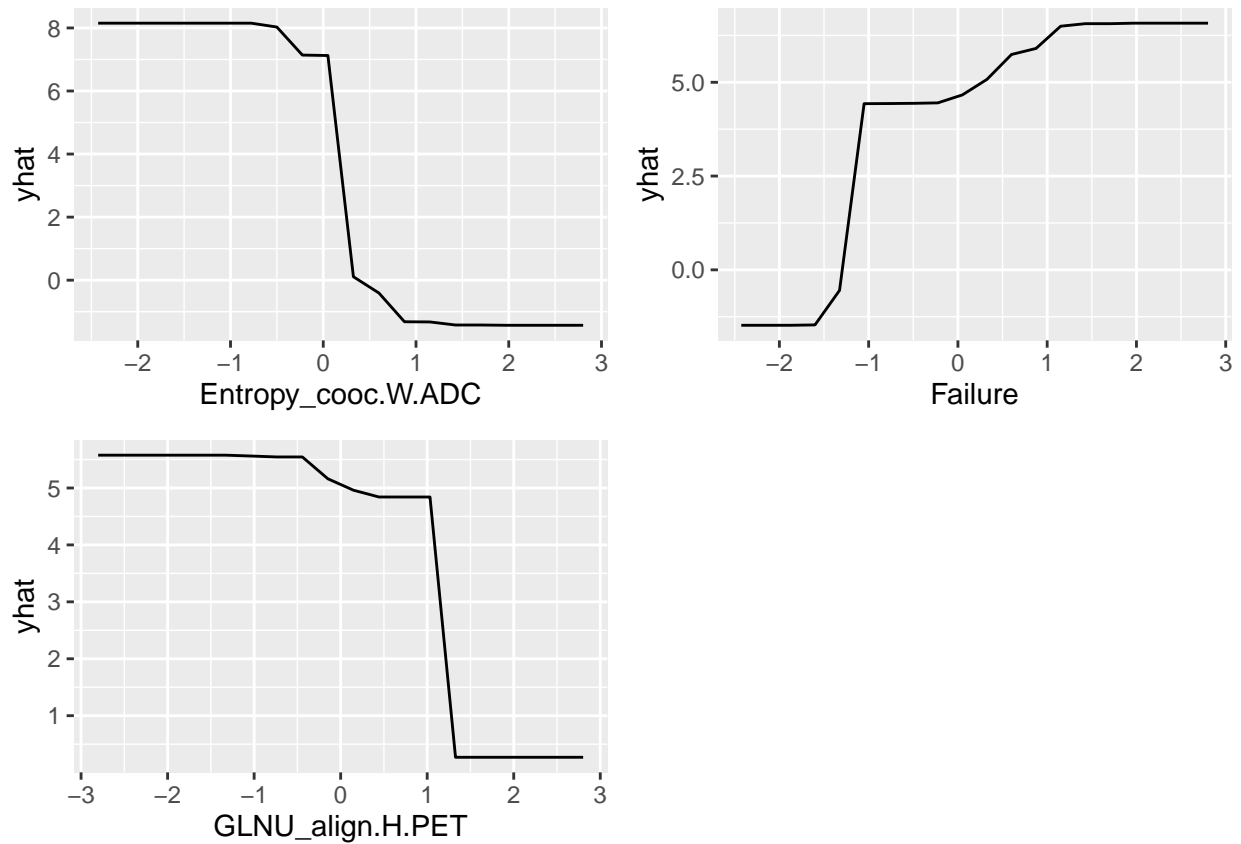
pred.var = "Entropy_cooc.W.ADC",
grid.resolution = 20
) %>%
autoplot()

p2 <- pdp::partial(
  bagging_2,
  pred.var = "Failure",
  grid.resolution = 20
) %>%
autoplot()

p3 = pdp::partial(
  bagging_2,
  pred.var = "GLNU_align.H.PET",
  grid.resolution = 20
) %>%
autoplot()

gridExtra::grid.arrange(p1, p2, p3, nrow = 2)

```



The partial dependence plot shows the dependence between the Failure.binary response/target variable and these top 3 feature variables (i.e. Failure, GLNU_align.H.PET and Entropy-cooc.W.ADC. of interest. As shown in PDP above, lower Entropy.cooc.W.ADC implies that Failure.binary is more likely to be 1, the

higher the Entropy_cooc.W.ADC the more likely the `Failure.binary` to be 0, and the lower the failure the more likely the `Final.binary` to be 0.