

	LR	LR+Lasso	NN
Origin dataset	\$AUC [1] 0.9249787 \$threshold [1] 0.41 \$total_accuracy [1] 0.9134394 \$confusion_matrix actual_response pred_class 0 1 0 7088 473 1 240 436 \$positive_precision [1] 0.6449704 \$positive_recall [1] 0.479648 \$negative_precision [1] 0.9374421 \$negative_recall [1] 0.9672489	\$AUC [1] 0.925106 \$threshold [1] 0.03 \$total_accuracy [1] 0.9111327 \$confusion_matrix actual_response pred_class 0 1 0 7161 565 1 167 344 \$positive_precision [1] 0.6731898 \$positive_recall [1] 0.3784378 \$negative_precision [1] 0.9268703 \$negative_recall [1] 0.9772107	Test AUC: 0.9379 Test Accuracy: threshold 0.510000 accuracy 0.916474 accuracy_check 0.916474 positive_precision 0.519252 positive_recall 0.652835 negative_precision 0.965748 negative_recall 0.941842
	Call: glm(formula = y ~ ., family = "binomial", data = full_train_bccp) Deviance Residuals: Min 1Q Median 3Q Max -5.9036 -0.3290 -0.1967 -0.1417 3.3040 Coefficients: (1 not defined because of singularities) Estimate Std. Error z value Pr(> z) (Intercept) -3.338880 0.148093 -22.546 < 2e-16 *** jobentrepreneur -0.141085 0.133287 -1.059 0.289823 jobhousemaid 0.135082 0.157316 0.859 0.390525	\$opt_lambda_lasso [1] 0.0004023168 \$opt_lasso_coefficients 38 x 1 sparse Matrix of class "dgCMatrix" s0 (Intercept) -3.240085766 jobentrepreneur -0.101759551 jobhousemaid 0.068790534 jobmanagement -0.024443403 jobretired 0.454686692 jobself.employed -0.034214446	

jobmanagement	-0.055452	0.092834	-0.597	0.550292	jobservices	.
jobretired	0.463713	0.113165	4.098	4.17e-05 ***	jobstudent	0.433259487
jobself.employed	-0.075421	0.130231	-0.579	0.562498	jobtechnician	0.139490293
jobservices	0.002575	0.088177	0.029	0.976702	jobunemployed	0.106934733
jobstudent	0.452538	0.119725	3.780	0.000157 ***	jobunknown	0.125840293
jobtechnician	0.149176	0.072499	2.058	0.039626 *	maritalmarried	.
jobunemployed	0.143216	0.137031	1.045	0.295959	maritalsingle	0.085591888
jobunknown	0.200656	0.257623	0.779	0.436052	maritalunknown	-0.004234320
maritalmarried	0.008157	0.075180	0.109	0.913596	educationbasic.6y	-0.012405746
maritalsingle	0.102266	0.085710	1.193	0.232805	educationbasic.9y	-0.132058899
maritalunknown	-0.111660	0.461173	-0.242	0.808686	educationhigh.school	.
educationbasic.6y	0.025368	0.131846	0.192	0.847424	educationilliterate	0.599202544
educationbasic.9y	-0.081533	0.103902	-0.785	0.432622	educationprofessional.course	0.090555664
educationhigh.school	0.072666	0.095145	0.764	0.445020	educationuniversity.degree	0.263813436
educationilliterate	0.959618	1.019522	0.941	0.346580	educationunknown	0.085553613
educationprofessional.course	0.168434	0.107780	1.563		defaultunknown	-0.405447497
0.118110					defaultyes	.
educationuniversity.degree	0.345672	0.091723	3.769		housingunknown	-0.052778568
0.000164 ***					housingyes	.
educationunknown	0.166480	0.130350	1.277	0.201540	loanunknown	-0.007608956
defaultunknown	-0.418613	0.074174	-5.644	1.66e-08	loanyes	-0.014321283
***					contacttelephone	-0.856569815
defaultyes	-7.221093	139.129448	-0.052	0.958607	poutcomenonexistent	0.510027798
housingunknown	-0.110378	0.153478	-0.719	0.472030	poutcomesuccess	0.834466317
housingyes	-0.007147	0.045451	-0.157	0.875057	age	0.007057162
loanunknown	NA	NA	NA	NA	duration	1.162522308
loanyes	-0.034646	0.063139	-0.549	0.583197	campaign	-0.077666398
contacttelephone	-0.893036	0.067400	-13.250	< 2e-16	pdays	-0.200885129
***					previous	.
poutcomenonexistent	0.539427	0.104257	5.174	2.29e-07	emp.var.rate	-1.395424537
***					cons.price.idx	0.738130891
poutcomesuccess	0.839071	0.235215	3.567	0.000361	cons.conf.idx	0.239016972

age	0.017887	0.027918	0.641	0.521720		
duration	1.175000	0.020813	56.454	< 2e-16 ***		
campaign	-0.091665	0.034473	-2.659	0.007837 **		
pdays	-0.205432	0.045110	-4.554	5.26e-06 ***		
previous	-0.007138	0.032525	-0.219	0.826291		
emp.var.rate	-1.422511	0.036811	-38.643	< 2e-16		

	cons.price.idx 0.767954 0.033985 22.597 < 2e-16 *** cons.conf.idx 0.244791 0.019960 12.264 < 2e-16 *** --- Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1 (Dispersion parameter for binomial family taken to be 1) Null deviance: 23277 on 32950 degrees of freedom Residual deviance: 14170 on 32914 degrees of freedom AIC: 14244 Number of Fisher Scoring iterations: 10		
Balanced dataset	\$AUC [1] 0.9293573 \$threshold [1] 0.36 \$total_accuracy [1] 0.8592077 \$confusion_matrix actual_response pred_class 0 1 0 752 75 1 188 853 \$positive_precision [1] 0.8194044 \$positive_recall [1] 0.919181 \$negative_precision [1] 0.9093108 \$negative_recall	\$AUC [1] 0.9289177 \$threshold [1] 0.01 \$total_accuracy [1] 0.8543897 \$confusion_matrix actual_response pred_class 0 1 0 803 135 1 137 793 \$positive_precision [1] 0.8526882 \$positive_recall [1] 0.8545259 \$negative_precision [1] 0.8560768 \$negative_recall	Test AUC: 0.9310 Test Accuracy: threshold 0.790000 accuracy 0.860278 accuracy_check 0.860278 positive_precision 0.934267 positive_recall 0.812559 negative_precision 0.787234 negative_recall 0.923845

	[1] 0.8	[1] 0.8542553	
	Call: glm(formula = y ~ ., family = "binomial", data = full_train_bccp)	\$opt_lambda_lasso [1] 0.004632514	
	Deviance Residuals: Min 1Q Median 3Q Max -6.8752 -0.4458 -0.1277 0.5131 2.7552	\$opt_lasso_coefficients 37 x 1 sparse Matrix of class "dgCMatrix"	
	Coefficients: (1 not defined because of singularities) Estimate Std. Error z value Pr(> z)	s0	
	(Intercept) -0.17538 0.23607 -0.743 0.457536	(Intercept) -0.067331652	
	jobentrepreneur -0.39157 0.19858 -1.972 0.048632 *	jobentrepreneur -0.197161296	
	jobhousemaid 0.18316 0.24455 0.749 0.453867	jobhousemaid .	
	jobmanagement -0.10380 0.14555 -0.713 0.475742	jobmanagement .	
	jobretired 0.64833 0.18548 3.495 0.000473 ***	jobretired 0.571581476	
	jobself.employed -0.28134 0.20239 -1.390 0.164503	jobself.employed -0.011048088	
	jobservices -0.06247 0.13485 -0.463 0.643180	jobservices -0.010062958	
	jobstudent 0.75815 0.20250 3.744 0.000181 ***	jobstudent 0.533445121	
	jobtechnician 0.06757 0.11449 0.590 0.555046	jobtechnician .	
	jobunemployed 0.41322 0.21308 1.939 0.052474 .	jobunemployed 0.164930969	
	jobunknown 0.47174 0.38912 1.212 0.225383	jobunknown 0.005647422	
	maritalmarried -0.02954 0.11780 -0.251 0.801960	maritalmarried .	
	maritalsingle 0.06276 0.13306 0.472 0.637184	maritalsingle 0.020126688	
	maritalunknown -0.44135 0.68199 -0.647 0.517530	maritalunknown .	
	educationbasic.6y -0.26317 0.21054 -1.250 0.211315	educationbasic.6y -0.121088274	
	educationbasic.9y -0.21769 0.16384 -1.329 0.183947	educationbasic.9y -0.156574395	
	educationhigh.school -0.10896 0.15484 -0.704 0.481639	educationhigh.school -0.050947411	
	educationilliterate 0.27144 1.23402 0.220 0.825900	educationilliterate .	
	educationprofessional.course 0.05143 0.17474 0.294 0.768522	educationprofessional.course .	
	educationuniversity.degree 0.43965 0.14991 2.933 0.003359 **	educationuniversity.degree 0.327212118	
	educationunknown 0.05622 0.21172 0.266 0.790592	educationunknown .	
	defaultunknown -0.51112 0.11380 -4.491 7.07e-06 ***	defaultunknown -0.401166734	
	housingunknown -0.01354 0.24238 -0.056 0.955439	housingunknown .	
	housingyes 0.04075 0.07135 0.571 0.567887	housingyes .	
	loanunknown NA NA NA NA	loanunknown .	
	loanyes -0.14344 0.09881 -1.452 0.146572	loanyes -0.010780495	
	contacttelephone -0.72783 0.10035 -7.253 4.08e-13 ***	contacttelephone -0.512798760	
		poutcomenonexistent 0.243256196	
		poutcomesuccess 0.901413653	
		age .	
		duration 2.137849593	
		campaign -0.086350607	

	<p>poutcomenonexistent 0.46234 0.17932 2.578 0.009928 **</p> <p>poutcomesuccess 0.97433 0.45326 2.150 0.031585 *</p> <p>age 0.04560 0.04974 0.917 0.359284</p> <p>duration 2.34788 0.06332 37.081 < 2e-16 ***</p> <p>campaign -0.14020 0.04343 -3.228 0.001246 **</p> <p>pdays -0.33381 0.14300 -2.334 0.019577 *</p> <p>previous -0.08312 0.08640 -0.962 0.336024</p> <p>emp.var.rate -1.76088 0.06559 -26.848 < 2e-16 ***</p> <p>cons.price.idx 0.74841 0.06169 12.132 < 2e-16 ***</p> <p>cons.conf.idx 0.28328 0.03686 7.686 1.52e-14 ***</p> <p>---</p> <p>Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1</p> <p>(Dispersion parameter for binomial family taken to be 1)</p> <p>Null deviance: 10360.8 on 7473 degrees of freedom</p> <p>Residual deviance: 5276.4 on 7438 degrees of freedom</p> <p>AIC: 5348.4</p> <p>Number of Fisher Scoring iterations: 6</p>	<p>pdays -0.245739352</p> <p>previous .</p> <p>emp.var.rate -1.506759145</p> <p>cons.price.idx 0.513634200</p> <p>cons.conf.idx 0.221030902</p>	
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