Extended evaluation of differential chromatin interaction detection analysis using simulated Hi-C data

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Introduction

To evaluate the performance of different normalization methods on the detection of chromatin interaction differences, controlled changes were used. To better control for existing differences in the real Hi-C data, simulated Hi-C datasets were used. The data was simulated using the hic_simulate function. For the simulated matrices, the default values were used (see ?hic_simulate) except for changing the size to be a 250 x 250 matrix. The effect of varying fold changes (1.5 by default) and varying number of controlled changes (500 by default) is investigated.

Fold changes are applied to one of the datasets by up-regulating the selected IF if the difference between the datasets is positive. If the difference between the datasets at that point is negative the IF is down-regulated by the specified fold change. This method of making changes ensures that the fold change specified is actually realized on the MD plot.

The effect of fold change

1.5 Fold change

	loess	chromoR	ice	kr	scn	ma
true positive	119	25	93	96	61	112
false positive	1050	1420	1080	1100	1160	1060
true negative	24800	29400	23900	24800	24700	24800
false negative	381	475	394	404	439	388
Total	26400	31400	25400	26400	26400	26400
$ ext{TPR}$	0.238	0.05	0.191	0.192	0.122	0.224
${\bf SPC}$	0.959	0.954	0.957	0.957	0.955	0.959
$\mathbf{F1}$	0.972	0.969	0.97	0.97	0.969	0.972
\mathbf{AUC}	0.728	0.513	0.641	0.642	0.589	0.714
m AUC~20%	0.0723	0.0212	0.0579	0.0588	0.0412	0.069
\mathbf{FDR}	0.898	0.983	0.921	0.92	0.95	0.904
Accuracy	0.946	0.939	0.942	0.943	0.94	0.945
Precision	0.102	0.0172	0.0793	0.0799	0.0502	0.0956
\mathbf{FPR}	0.0406	0.0462	0.0433	0.0427	0.0447	0.041
\mathbf{FNR}	0.762	0.95	0.809	0.808	0.878	0.776
FOR	0.0151	0.0159	0.0162	0.0161	0.0175	0.0154
NPV	0.985	0.984	0.984	0.984	0.983	0.985

2.0 Fold change

	loess	$\operatorname{chromoR}$	ice	kr	scn	ma
true positive	267	31	196	201	112	239
false positive	903	1430	980	1020	1100	926
true negative	25100	29400	24100	24900	24900	25000
false negative	233	469	291	299	388	261
Total	26500	31400	25600	26500	26500	26500
$ ext{TPR}$	0.534	0.062	0.402	0.402	0.224	0.478
\mathbf{SPC}	0.965	0.954	0.961	0.961	0.958	0.964
$\mathbf{F1}$	0.978	0.969	0.974	0.974	0.971	0.977
\mathbf{AUC}	0.903	0.503	0.784	0.786	0.703	0.873
m AUC~20%	0.138	0.0236	0.103	0.104	0.067	0.129
FDR	0.772	0.979	0.833	0.835	0.908	0.795
Accuracy	0.957	0.939	0.95	0.95	0.944	0.955
Precision	0.228	0.0212	0.167	0.165	0.0923	0.205
\mathbf{FPR}	0.0348	0.0464	0.0391	0.0392	0.0424	0.0357
\mathbf{FNR}	0.466	0.938	0.598	0.598	0.776	0.522
FOR	0.00921	0.0157	0.0119	0.0118	0.0154	0.0103
\mathbf{NPV}	0.991	0.984	0.988	0.988	0.985	0.99

4.0 Fold change

	loess	$\operatorname{chromoR}$	ice	kr	scn	ma
true positive	474	55	408	421	293	461
false positive	716	1400	774	791	917	712
true negative	25200	29500	24600	25100	25000	25200
false negative	26	445	85	79	207	39
Total	26400	31400	25800	26400	26400	26400
TPR	0.948	0.11	0.828	0.842	0.586	0.922
\mathbf{SPC}	0.972	0.955	0.969	0.969	0.965	0.973
$\mathbf{F1}$	0.985	0.97	0.983	0.983	0.978	0.985
\mathbf{AUC}	0.993	0.514	0.958	0.959	0.906	0.989
m AUC~20%	0.194	0.0311	0.174	0.175	0.141	0.19
FDR	0.602	0.962	0.655	0.653	0.758	0.607
Accuracy	0.972	0.941	0.967	0.967	0.957	0.972
Precision	0.398	0.0377	0.345	0.347	0.242	0.393
\mathbf{FPR}	0.0276	0.0454	0.0305	0.0305	0.0354	0.0275
\mathbf{FNR}	0.052	0.89	0.172	0.158	0.414	0.078
FOR	0.00103	0.0149	0.00345	0.00314	0.00821	0.00155
\mathbf{NPV}	0.999	0.985	0.997	0.997	0.992	0.998

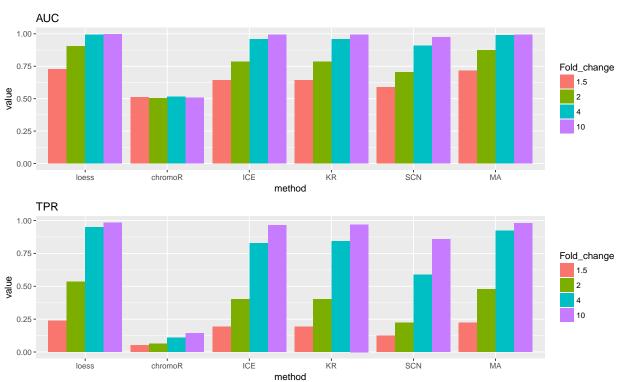
10.0 fold change

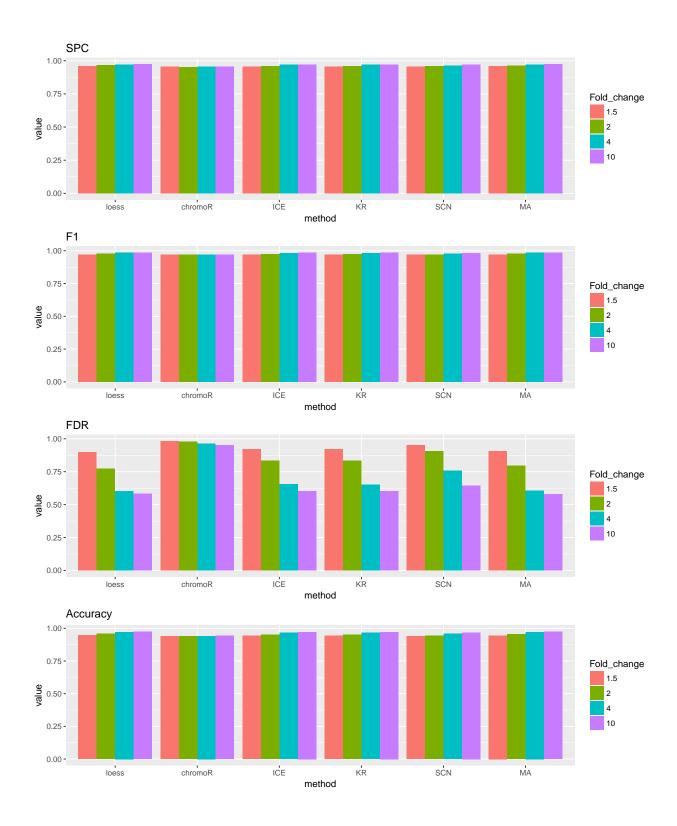
	loess	chromoR	ice	kr	scn	ma
true positive	492	71	472	485	429	490
false positive	685	1390	712	733	782	676
true negative	25200	29500	24500	25200	25100	25200
false negative	8	429	17	15	71	10

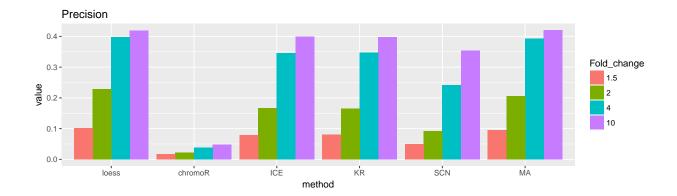
	loess	chromoR	ice	kr	scn	ma
Total	26400	31400	25700	26400	26400	26400
\mathbf{TPR}	0.984	0.142	0.965	0.97	0.858	0.98
\mathbf{SPC}	0.974	0.955	0.972	0.972	0.97	0.974
$\mathbf{F1}$	0.986	0.97	0.985	0.985	0.983	0.987
\mathbf{AUC}	0.996	0.506	0.994	0.994	0.975	0.994
m AUC~20%	0.196	0.0303	0.194	0.194	0.181	0.195
FDR	0.582	0.952	0.601	0.602	0.646	0.58
Accuracy	0.974	0.942	0.972	0.972	0.968	0.974
Precision	0.418	0.0485	0.399	0.398	0.354	0.42
\mathbf{FPR}	0.0265	0.0452	0.0283	0.0283	0.0302	0.0261
\mathbf{FNR}	0.016	0.858	0.0348	0.03	0.142	0.02
\mathbf{FOR}	0.000317	0.0143	0.000695	0.000596	0.00282	0.000396
NPV	1	0.986	0.999	0.999	0.997	1

Bar plots

The bar plots show comparisons of the effect of different fold changes using fixed numbers of controlled changes on different performance metrics.







Summary

The tables show that the most true differences are detected after joint loess normalization compared to the other normalization techniques. loess also has the lowest number of false positives among the normalization techniques. With loess, differences at smaller fold changes (1.5 and 2) are able to be detected more reliably compared to the other methods and this superiority continues to the higher fold changes though the individual normalization methods tend to make up some ground.MA normalization performed almost as well as loess at lower fold changes and was about equivalent at the higher fold changes. ChromoR performed the worst of the normalization techniques tested while KR and SCN tended to perform better.

The effect of introduced number of changes

A specified number of interaction frequencies were increased or decreased to produce a 1.5-fold change.

1 change

	loess	$\operatorname{chromoR}$	ice	kr	scn	$_{\mathrm{ma}}$
true positive	0	0	0	0	0	0
false positive	1180	1440	1170	1210	1210	1160
true negative	25200	29900	24500	25200	25200	25200
false negative	1	1	1	1	1	1
Total	26400	31400	25700	26400	26400	26400
$ ext{TPR}$	0	0	0	0	0	0
\mathbf{SPC}	0.955	0.954	0.954	0.954	0.954	0.956
$\mathbf{F1}$	0.977	0.976	0.977	0.976	0.977	0.978
\mathbf{AUC}	0.51	0.685	0.701	0.687	0.977	0.943
$\mathbf{AUC}\ 20\%$	0	0	0	0	0.177	0.143
FDR	1	1	1	1	1	1
Accuracy	0.955	0.954	0.954	0.954	0.954	0.956
Precision	0	0	0	0	0	0
\mathbf{FPR}	0.0446	0.0461	0.0457	0.0459	0.0459	0.044
\mathbf{FNR}	1	1	1	1	1	1
FOR	3.96 e-05	3.34e-05	4.08e-05	3.97e-05	3.97e-05	3.96e-05
NPV	1	1	1	1	1	1

100 changes

	loess	chromoR	ice	kr	scn	ma
true positive	27	3	14	14	5	25
false positive	1140	1460	1160	1220	1220	1130
true negative	25300	29800	24400	25200	25200	25300
false negative	73	97	80	86	95	75
Total	26600	31400	25600	26600	26600	26600
TPR	0.27	0.03	0.149	0.14	0.05	0.25
${\bf SPC}$	0.957	0.953	0.954	0.954	0.954	0.957
$\mathbf{F1}$	0.977	0.975	0.975	0.975	0.975	0.977
\mathbf{AUC}	0.774	0.511	0.638	0.627	0.549	0.74
$\mathbf{AUC}\ 20\%$	0.0824	0.0191	0.0553	0.0546	0.0272	0.0745
FDR	0.977	0.998	0.988	0.989	0.996	0.978
Accuracy	0.954	0.95	0.951	0.951	0.951	0.955
Precision	0.0231	0.00206	0.0119	0.0114	0.00409	0.0216
\mathbf{FPR}	0.0431	0.0466	0.0456	0.046	0.046	0.0428
\mathbf{FNR}	0.73	0.97	0.851	0.86	0.95	0.75
FOR	0.00288	0.00324	0.00327	0.0034	0.00375	0.00295
NPV	0.997	0.997	0.997	0.997	0.996	0.997

200 changes

	loess	chromoR	ice	kr	scn	ma
true positive	53	8	33	35	17	47
false positive	1110	1450	1150	1180	1190	1110
true negative	25100	29700	24500	25000	25000	25100
false negative	147	192	164	165	183	153
Total	26400	31400	25800	26400	26400	26400
$ ext{TPR}$	0.265	0.04	0.168	0.175	0.085	0.235
SPC	0.958	0.953	0.955	0.955	0.954	0.958
$\mathbf{F1}$	0.976	0.973	0.974	0.974	0.973	0.975
\mathbf{AUC}	0.753	0.535	0.676	0.678	0.577	0.75
$\mathbf{AUC}\ 20\%$	0.0761	0.027	0.0545	0.0547	0.038	0.073
FDR	0.954	0.995	0.972	0.971	0.986	0.959
Accuracy	0.952	0.948	0.949	0.949	0.948	0.952
Precision	0.0457	0.00549	0.0279	0.0288	0.014	0.0407
\mathbf{FPR}	0.0423	0.0465	0.0449	0.0451	0.0456	0.0423
\mathbf{FNR}	0.735	0.96	0.832	0.825	0.915	0.765
FOR	0.00583	0.00642	0.00666	0.00656	0.00727	0.00607
\mathbf{NPV}	0.994	0.994	0.993	0.993	0.993	0.994

1000 changes

	loess	chromoR	ice	kr	scn	ma
true positive	216	47	164	171	98	213
false positive	970	1410	1000	1050	1120	972
true negative	24500	29000	23600	24400	24300	24500
false negative	784	953	799	829	902	787
Total	26400	31400	25600	26400	26400	26400
TPR	0.216	0.047	0.17	0.171	0.098	0.213

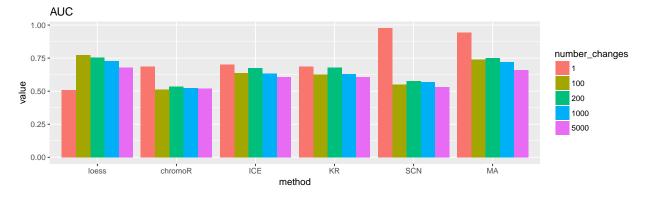
	loess	chromoR	ice	kr	scn	ma
SPC	0.962	0.954	0.959	0.959	0.956	0.962
$\mathbf{F1}$	0.965	0.961	0.963	0.963	0.96	0.965
\mathbf{AUC}	0.727	0.524	0.634	0.631	0.569	0.719
m AUC~20%	0.0726	0.0225	0.0546	0.0551	0.0338	0.0693
FDR	0.818	0.968	0.86	0.86	0.919	0.82
Accuracy	0.934	0.925	0.929	0.929	0.924	0.933
Precision	0.182	0.0322	0.14	0.14	0.0805	0.18
\mathbf{FPR}	0.0381	0.0465	0.0408	0.0413	0.044	0.0382
\mathbf{FNR}	0.784	0.953	0.83	0.829	0.902	0.787
FOR	0.031	0.0319	0.0328	0.0329	0.0357	0.0312
\mathbf{NPV}	0.969	0.968	0.967	0.967	0.964	0.969

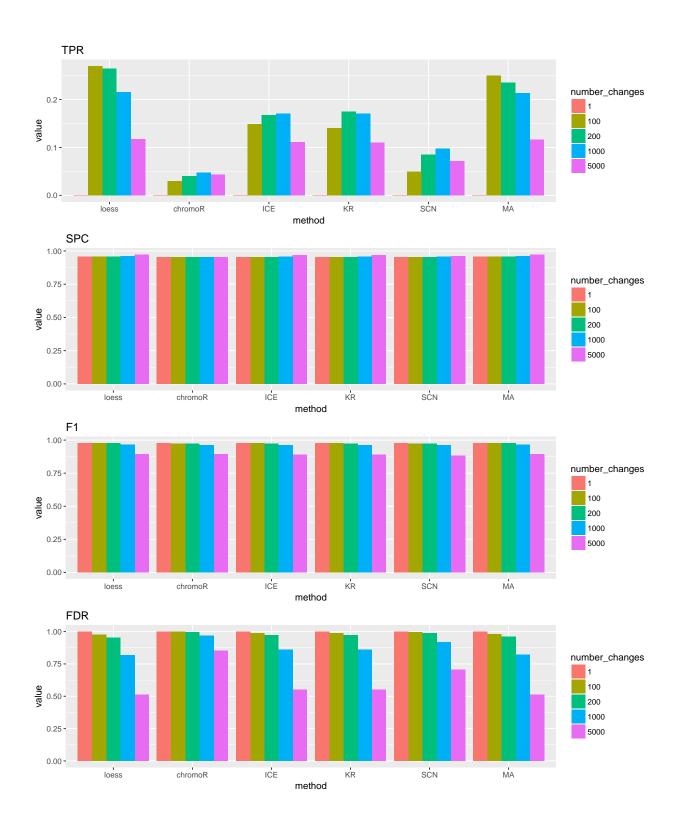
5000 changes

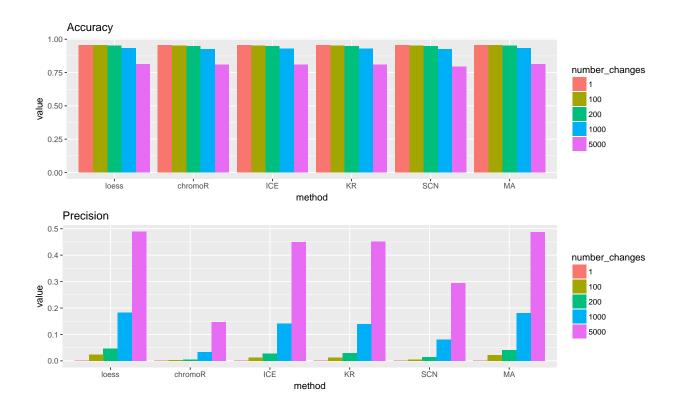
-	loess	chromoR	ice	kr	scn	ma
true positive	586	216	532	553	359	581
false positive	613	1250	650	673	861	610
true negative	21000	25100	20200	20900	20800	21000
false negative	4410	4780	4270	4450	4640	4420
Total	26600	31400	25700	26600	26600	26600
$ ext{TPR}$	0.117	0.0432	0.111	0.111	0.0718	0.116
SPC	0.972	0.953	0.969	0.969	0.96	0.972
$\mathbf{F1}$	0.893	0.893	0.891	0.891	0.883	0.893
\mathbf{AUC}	0.678	0.519	0.608	0.607	0.532	0.661
${\rm AUC}~20\%$	0.0553	0.0222	0.0457	0.0452	0.029	0.0528
FDR	0.511	0.853	0.55	0.549	0.706	0.512
Accuracy	0.811	0.808	0.808	0.808	0.793	0.811
Precision	0.489	0.147	0.45	0.451	0.294	0.488
\mathbf{FPR}	0.0284	0.0474	0.0311	0.0311	0.0398	0.0282
\mathbf{FNR}	0.883	0.957	0.889	0.889	0.928	0.884
FOR	0.174	0.16	0.174	0.175	0.183	0.174
\mathbf{NPV}	0.826	0.84	0.826	0.825	0.817	0.826

Bar plots

Below are bar plots showing comparisons of the different normalization methods over the varying numbers of changes at a fixed fold change for selected metrics.







Summary

When only a single true difference was introduced at a 1.5 fold change all methods failed to detect it. At all other numbers of introduced differences checked, loess was able to detect the most while also maintaining the lowest number of false positives. KR and SCN again performed better than ICE and ChromoR at detecting the true differences.