

BCB570 HW5

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Solution 1

Bagging

Bagging is an ensemble method to improve stability of learning algorithms by reducing its variance. In general, given a dataset D with n datapoints (in any k -dimensional space), bagging generates m new datasets each of size n . It does so by doing sampling-with-replacement and samples n datapoints m times. If n is large enough, the 63.2% unique datapoints are expected to be in the m samples. A variation of bagging can have another parameter n' which is the dataset size of the sampled datasets. Here, bagging will generate m new datasets each of size n' by doing sampling-with-replacement.

```
# sample 100 datapoints
n_data_100 <- rnorm(100, mean = 2, sd = sqrt(8))
cat("Mean is", mean(n_data_100))
```

```
## Mean is 2.024138
```

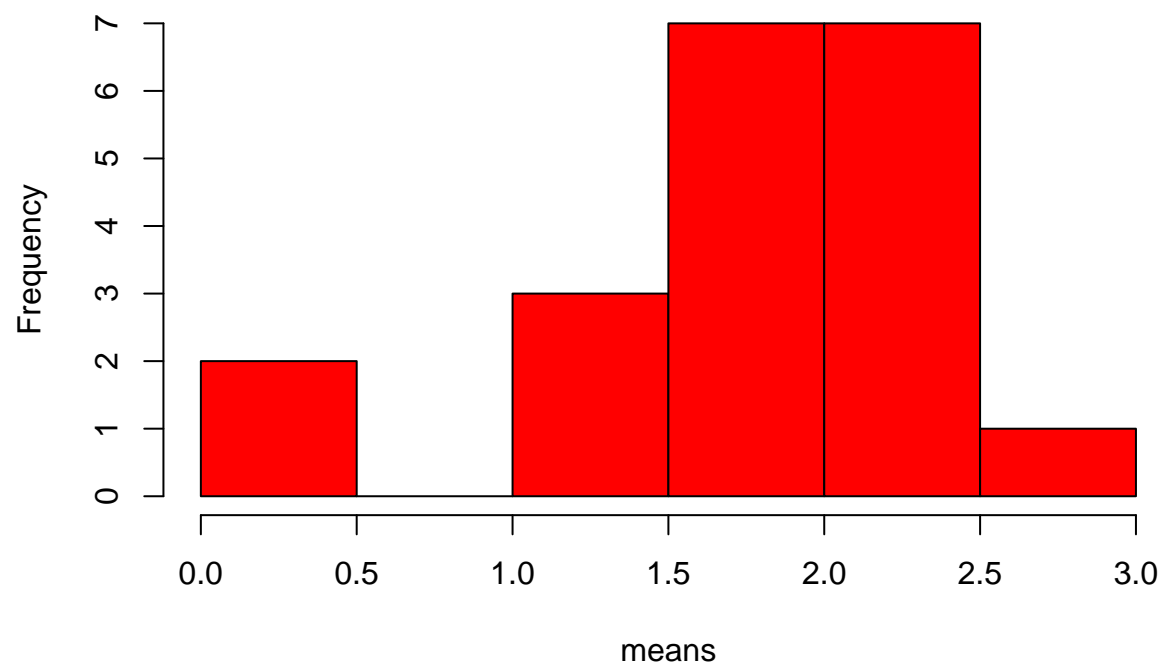
```
cat("Var is", var(n_data_100))
```

```
## Var is 8.714753
```

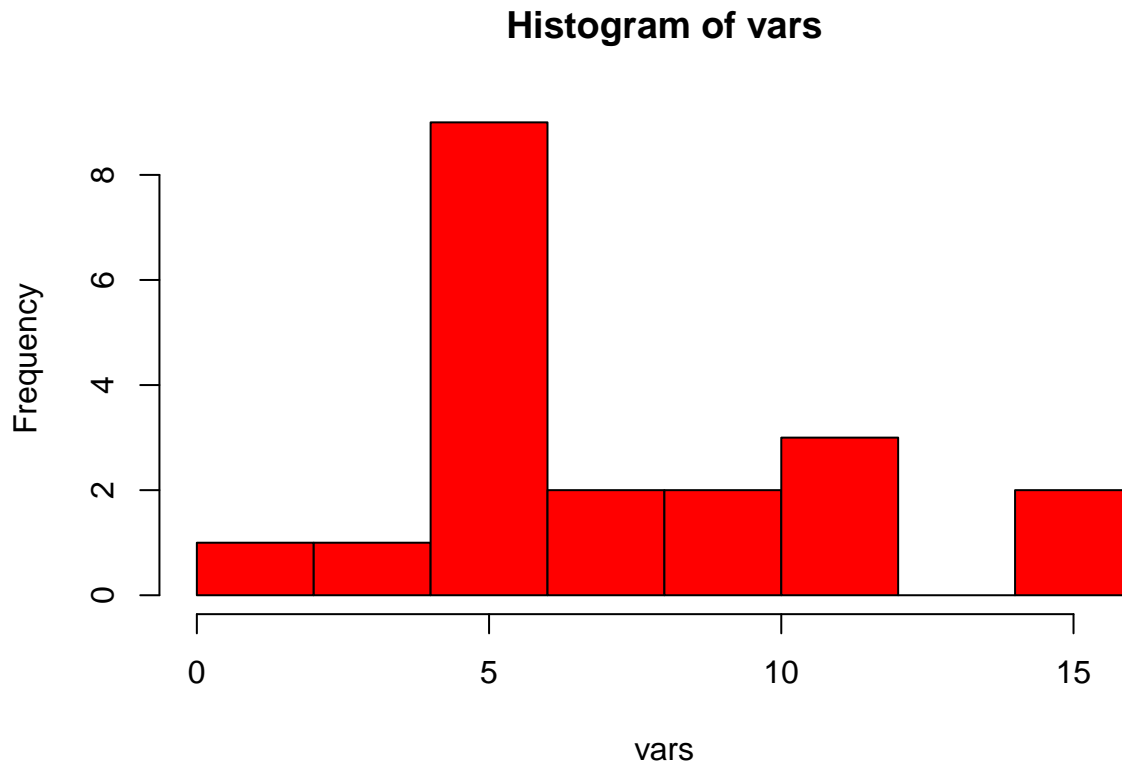
```
means <- c()
vars <- c()
N <- 20
k <- 10
for (i in 1:N) {
  this <- sample(n_data_100, k, replace = T)
  # print(this)
  means <- c(means, mean(this))
  vars <- c(vars, var(this))
}

hist(means, col = c("Red"))
```

Histogram of means



```
hist(vars, col = c("Red"))
```



```
e_mean <- mean(means)
e_var <- mean(vars)
```

From the histograms it is hard to estimate true parameters as we sampled very few datasets. From the sub samples the mean of means is 1.7435098 and mean of variances is 7.4471093. These are our best estimates from our sub-samples.

Solution 2

Creating association networks

To create any association network we need to go through following steps.

1. First we have to define a way that can “associate” datapoints or variables. If two points are associated we must connect them through an edge and in this way we can create a network. We have to define a mathematical function that can quantify association between two datapoints e.g. we can compute the correlation between two variables if it is greater than zero, we say the variables are associated.
2. We may choose an association function, depending on the type of network we want i.e directed, undirected, weighted or unweighted. E.g. if we want directed association graph our association function should not be always symmetrical.
3. After defining a function that can quantify association, we need to set a threshold such that we create edges between only the strongly associated nodes. E.g. if we define distance between two points to be our measure of association then we need to set a cut-off such that if the distance is more than a value, we do not count that as true association.

- 3a Depending on the data, sometimes it may be useful to scale all the association metrics computed to make all the association comparable with in the graph

A novel association method

A method for creating association networks, not discussed in class, is using partial correlations. This method is not completely novel as it has been applied to study phenotypic networks (Chu et. al. 2014) but i am describing this in context of my data which may be novel. I am working with human transcript expression data and it would be interesting to create an association network using partial correlations. Partial correlation method finds the correlation between two variable while conditional all other variables. We can easily threshold this kind of network by calculating p-values as under the null, where all variables are independent, Hotelling is the null distribution. Using this we can test for significant partial correlations and filter the network. I can compare results of networks obtained from using correlation and partial correlation networks. A network based on partial correlation is expected to be more dense and may reveal interesting hidden links. So I will first select the transcript which are highly likely to interact using mutual information metric and filter some of the weekly interacting components. Then, on the remaining transcripts i'll use partial correlation measure to build an association network.

Solution 4

```
library(GENIE3)
library(readr)
library("doParallel", lib.loc = "~/R/win-library/3.4")
library("doRNG", lib.loc = "~/R/win-library/3.4")
library(plyr)
library(PRRoc)

# read true labels
trueEcoli1 <- read.csv("NIHW in silico data/NIHW in silico data/Size100/DREAM3 gold standards/DREAM3GoldStandardsEcoli1.csv",
  sep = "\t", header = F)
trueEcoli2 <- read.csv("NIHW in silico data/NIHW in silico data/Size100/DREAM3 gold standards/DREAM3GoldStandardsEcoli2.csv",
  sep = "\t", header = F)
# do for ecolli data 1
adata <- read.csv("NIHW in silico data/NIHW in silico data/Size100/DREAM3 data/InSilicoSize100-Ecoli1-t.csv",
  sep = "\t")
adata <- t(adata)
expmat <- as.matrix(adata[2:101, 1:966])
rownames(expmat) <- paste("G", 1:100, sep = "")
colnames(expmat) <- paste("Samp", 1:966, sep = "")

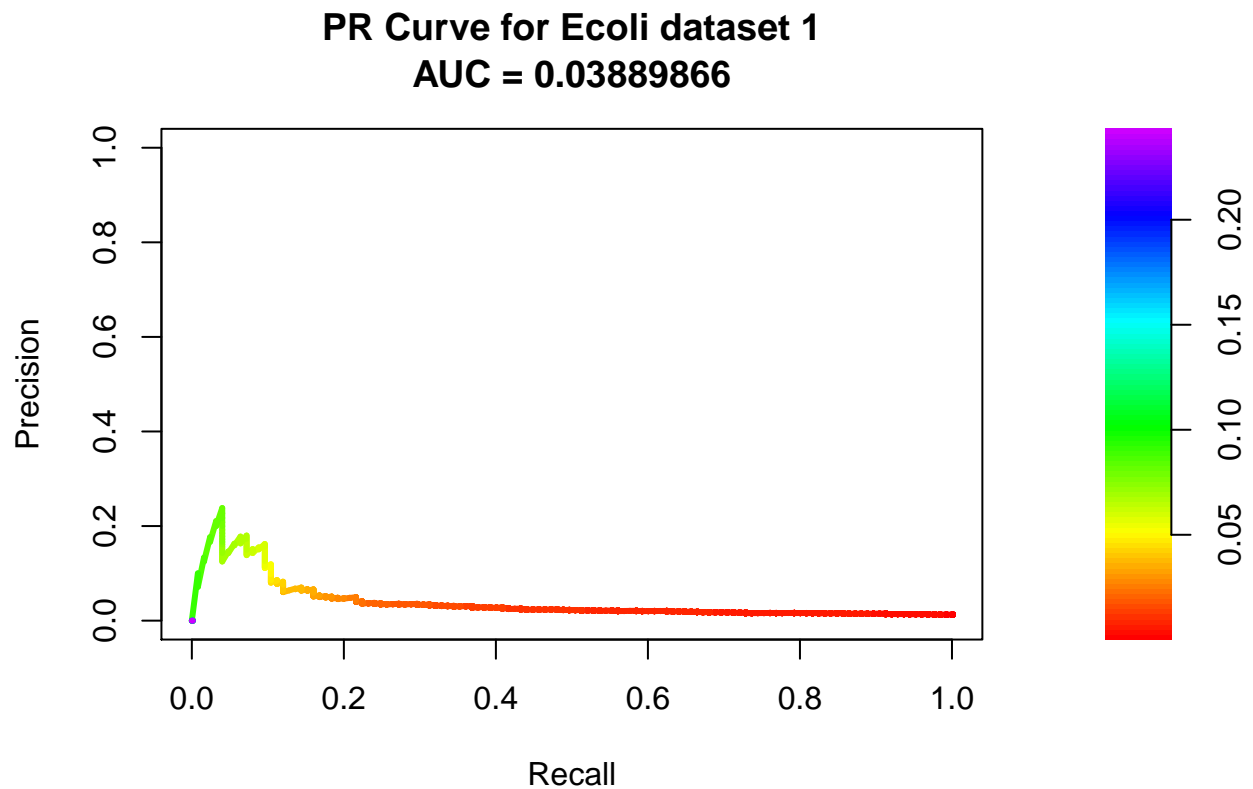
res1 <- GENIE3(expmat, nCores = 6, nTrees = 100)
linkList <- getLinkList(res1)
names(linkList) <- c("RG", "TG", "Weight")
names(trueEcoli1) <- c("RG", "TG", "Label")

q4thresh <- quantile(linkList$Weight, 0.75)
# q4thresh<-0.1
linkList_q4 <- linkList[linkList$Weight >= q4thresh, ]
names(linkList_q4) <- c("RG", "TG", "Weight")
joined_data <- plyr::join(linkList, trueEcoli1)
joined_data_q4 <- plyr::join(linkList_q4, trueEcoli1)
pr1 <- pr.curve(scores.class0 = joined_data$Weight, weights.class0 = joined_data$Label,
```

```

curve = T)
plot(pr1, main = "PR Curve for Ecoli dataset 1")

```

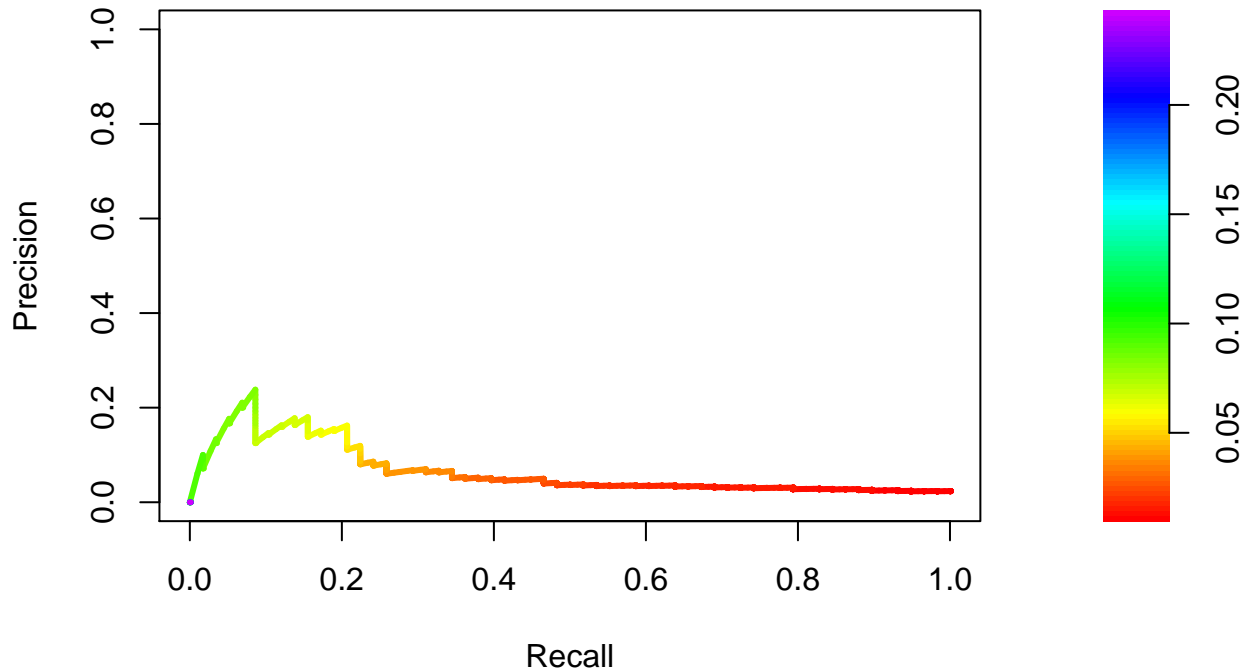


```

pr1_1 <- pr.curve(scores.class0 = joined_data_q4$Weight, weights.class0 = joined_data_q4$Label,
  curve = T)
plot(pr1_1, main = "PR Curve for Ecoli dataset 1 (only top quartile edges)")

```

PR Curve for Ecoli dataset 1 (only top quartile edges)
AUC = 0.0636776

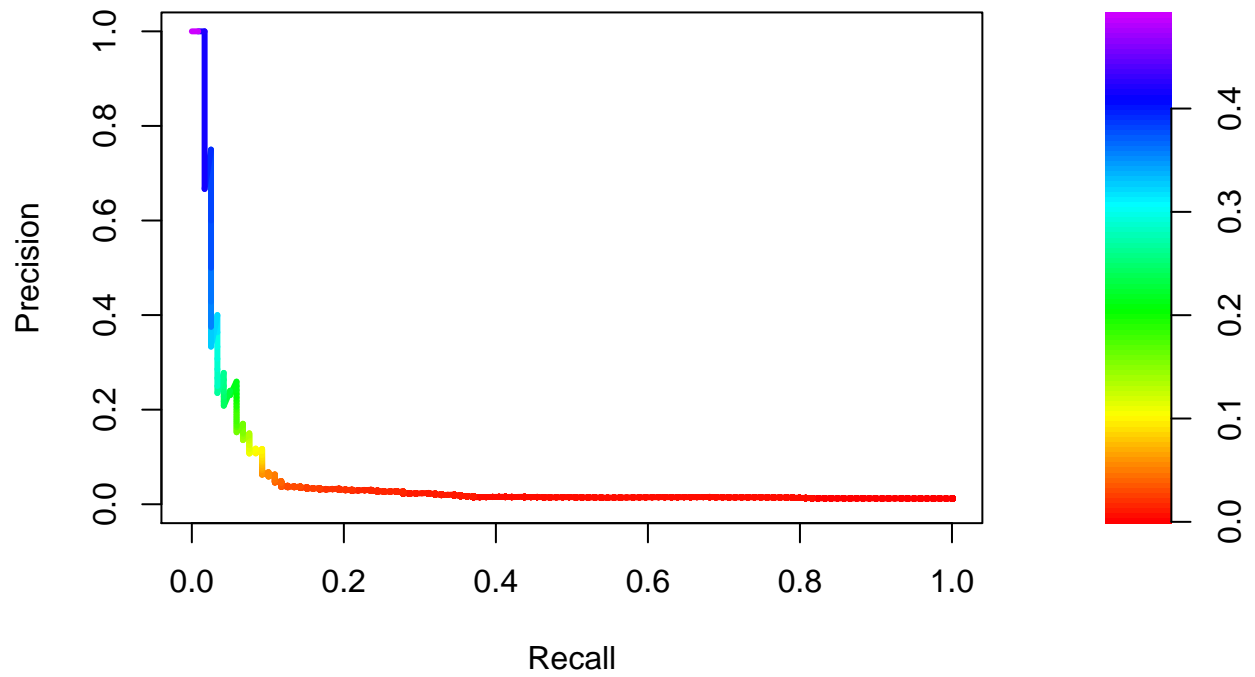


```
# do for ecoli data 2
adata2 <- read.csv("NIHW in silico data/NIHW in silico data/Size100/DREAM3 data/InSilicoSize100-Ecoli2-
  sep = "\t")
adata2 <- t(adata2)
expmat2 <- as.matrix(adata2[2:101, 1:966])
rownames(expmat2) <- paste("G", 1:100, sep = "")
colnames(expmat2) <- paste("Samp", 1:966, sep = "")

res2 <- GENIE3(expmat2, nCores = 6, K = "all", nTrees = 100)
linkList2 <- getLinkList(res2)
names(linkList2) <- c("RG", "TG", "Weight")

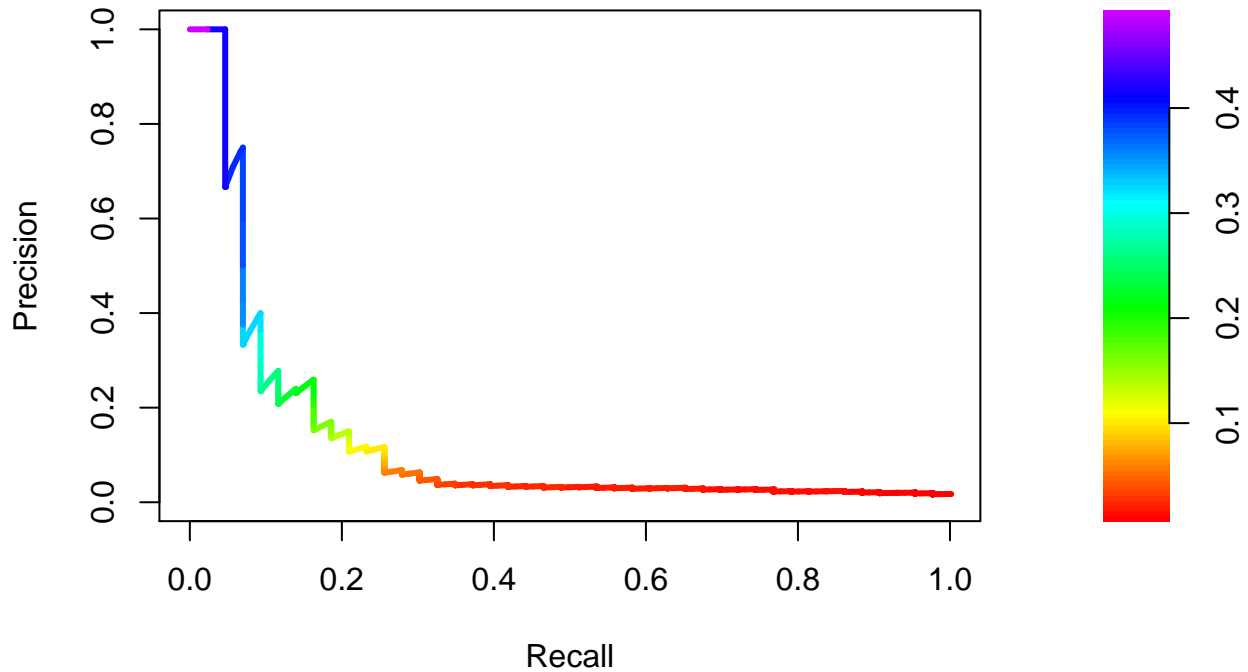
names(trueEcoli2) <- c("RG", "TG", "Label")
q4thresh2 <- quantile(linkList2$Weight, 0.75)
linkList2_q4 <- linkList2[linkList2$Weight >= q4thresh2, ]
names(linkList2_q4) <- c("RG", "TG", "Weight")
joined_data2 <- plyr::join(linkList2, trueEcoli2)
joined_data2_q4 <- plyr::join(linkList2_q4, trueEcoli2)
pr2 <- pr.curve(scores.class0 = joined_data2$Weight, weights.class0 = joined_data2$Label,
  curve = T)
plot(pr2, main = "PR Curve for Ecoli dataset 2")
```

PR Curve for Ecoli dataset 2
AUC = 0.05363801



```
pr2_1 <- pr.curve(scores.class0 = joined_data2_q4$Weight, weights.class0 = joined_data2_q4$Label,  
  curve = T)  
plot(pr2_1, main = "PR Curve for Ecoli dataset 2 (only top quartile edges)")
```

PR Curve for Ecoli dataset 2 (only top quartile edges)
AUC = 0.123601



```
# save results to file
write_tsv(linkList, "Ecoli1_results_all.tsv")
write_tsv(linkList_q4, "Ecoli1_results_topq.tsv")
write_tsv(linkList2, "Ecoli2_results_all.tsv")
write_tsv(linkList2_q4, "Ecoli2_results_topq.tsv")

# find best estimate intersection from linkList_q4 and linkList2_q4
temp_linkList_q4 <- linkList_q4
names(temp_linkList_q4) <- c("RG", "TG", "ecoli1")
temp_linkList2_q4 <- linkList2_q4
names(temp_linkList2_q4) <- c("RG", "TG", "ecoli2")
# common edges
common <- plyr::join(temp_linkList_q4, temp_linkList2_q4, type = "inner")
# save to file
write_tsv(common, "Ecoli_best_estimate.tsv")
```

a

GENIE3 is an algorithm for inferring gene regulatory networks from expression data. GENIE3 decomposes the prediction of a regulatory network between p genes into p different regression problems. In each of the regression problems, the expression pattern of one of the genes (target gene) is predicted from the expression patterns of all the other genes (input genes), using tree-based ensemble methods.

I ran GENIE3 with default parameters except I changed number of trees to 100 (to be faster). To threshold the network I took the top quartile values and selected the edges.

b

I ran GENEI3 for the two ecoli data sets on the trajectories file. Then I plotted the PR curve as shown in the attached figures. Then I filtered my networks by keeping only the edges which have weights in the top quartile. To come up with my best estimate about Ecoli network, I took the intersection of the two results i.e. if edge is present in the two networks. I saved these results to file “Ecoli_best_estimate.tsv”.

c Best estimates for ecoli data sets

common				
##	RG	TG	ecoli1	ecoli2
## 1	G66	G68	0.09550842	0.033417604
## 2	G18	G16	0.08563683	0.011648596
## 3	G31	G30	0.08346623	0.040799943
## 4	G68	G66	0.08264686	0.016296007
## 5	G37	G93	0.07636189	0.064046040
## 6	G28	G45	0.07032582	0.015785393
## 7	G63	G60	0.06979220	0.022103001
## 8	G18	G32	0.06965476	0.016845274
## 9	G16	G18	0.06771008	0.010163507
## 10	G14	G93	0.06759043	0.020017658
## 11	G41	G45	0.06751179	0.009864705
## 12	G18	G46	0.06656073	0.024081322
## 13	G99	G5	0.06433153	0.010759347
## 14	G82	G43	0.06365858	0.019729160
## 15	G45	G28	0.06185529	0.019163994
## 16	G30	G31	0.06135637	0.030019798
## 17	G40	G5	0.06094871	0.016461434
## 18	G93	G37	0.06074766	0.098612802
## 19	G31	G36	0.06066831	0.011496761
## 20	G97	G39	0.05814637	0.009990772
## 21	G83	G38	0.05740563	0.036520452
## 22	G32	G18	0.05688047	0.012798603
## 23	G90	G60	0.05641278	0.010744575
## 24	G88	G36	0.05596317	0.009386102
## 25	G99	G2	0.05566076	0.041797108
## 26	G33	G20	0.05548770	0.033741731
## 27	G23	G81	0.05509880	0.016668972
## 28	G40	G17	0.05333692	0.030176309
## 29	G36	G34	0.05256332	0.009601923
## 30	G23	G21	0.05243834	0.011945607
## 31	G16	G9	0.05236794	0.040137635
## 32	G5	G49	0.05168572	0.022957553
## 33	G17	G5	0.05078428	0.009215007
## 34	G18	G59	0.04987122	0.009471102
## 35	G51	G87	0.04965906	0.011863316
## 36	G63	G64	0.04945444	0.171017427
## 37	G16	G80	0.04938229	0.019806393
## 38	G28	G98	0.04801344	0.010306426
## 39	G82	G83	0.04671165	0.417102096
## 40	G47	G50	0.04428631	0.009279675
## 41	G51	G8	0.04387378	0.044331969

## 42	G61	G59	0.04380656	0.009579306
## 43	G61	G64	0.04361619	0.019268455
## 44	G39	G38	0.04357242	0.015530803
## 45	G32	G59	0.04262978	0.018944100
## 46	G68	G55	0.04177970	0.011411278
## 47	G50	G47	0.04176139	0.010402571
## 48	G43	G97	0.04149473	0.014574427
## 49	G5	G85	0.04105565	0.009799214
## 50	G64	G63	0.04093342	0.105197477
## 51	G14	G45	0.04081945	0.011576487
## 52	G75	G3	0.04020607	0.042214776
## 53	G51	G40	0.04012734	0.019246157
## 54	G68	G70	0.04011311	0.013675525
## 55	G49	G5	0.03998728	0.012354642
## 56	G41	G2	0.03975259	0.016928342
## 57	G39	G93	0.03972749	0.011024929
## 58	G97	G2	0.03968363	0.041735961
## 59	G16	G20	0.03955842	0.011896984
## 60	G17	G8	0.03920944	0.010039812
## 61	G41	G16	0.03914060	0.010129431
## 62	G35	G26	0.03912696	0.016311448
## 63	G80	G9	0.03889456	0.011233662
## 64	G21	G79	0.03876957	0.011618636
## 65	G9	G23	0.03849022	0.010621985
## 66	G49	G31	0.03808134	0.012153970
## 67	G64	G41	0.03797340	0.023641587
## 68	G34	G36	0.03792750	0.010155900
## 69	G60	G57	0.03790503	0.095877730
## 70	G40	G80	0.03786658	0.034194037
## 71	G82	G12	0.03776867	0.009234299
## 72	G99	G63	0.03732218	0.010507367
## 73	G21	G23	0.03703495	0.010749170
## 74	G41	G47	0.03692703	0.030236280
## 75	G16	G32	0.03689561	0.016269992
## 76	G49	G96	0.03677504	0.009815624
## 77	G23	G9	0.03674412	0.031960401
## 78	G28	G30	0.03673785	0.017631801
## 79	G37	G18	0.03618648	0.013782284
## 80	G81	G86	0.03616645	0.018829778
## 81	G23	G59	0.03605691	0.010950341
## 82	G93	G41	0.03603287	0.061618257
## 83	G17	G6	0.03581804	0.020602758
## 84	G93	G40	0.03560491	0.117911092
## 85	G80	G40	0.03547144	0.011042922
## 86	G64	G59	0.03542134	0.009930045
## 87	G96	G49	0.03534876	0.030223350
## 88	G14	G86	0.03530097	0.011460183
## 89	G18	G37	0.03527731	0.010415026
## 90	G34	G19	0.03491836	0.020757611
## 91	G55	G44	0.03477841	0.010134641
## 92	G3	G75	0.03469260	0.024129155
## 93	G88	G86	0.03431062	0.014521874
## 94	G17	G40	0.03421996	0.077210727
## 95	G46	G18	0.03365283	0.016662045

## 96	G32	G38	0.03359961	0.023757966
## 97	G42	G30	0.03359512	0.026147607
## 98	G2	G95	0.03353169	0.022873042
## 99	G47	G92	0.03340733	0.014101340
## 100	G64	G61	0.03318244	0.024101959
## 101	G92	G44	0.03308052	0.013338764
## 102	G6	G94	0.03297438	0.019151756
## 103	G72	G73	0.03273757	0.023934662
## 104	G81	G23	0.03215453	0.036361903
## 105	G37	G61	0.03204300	0.053511280
## 106	G81	G2	0.03199372	0.016084251
## 107	G55	G89	0.03194773	0.014502308
## 108	G79	G5	0.03165494	0.010691902
## 109	G42	G99	0.03151784	0.060846343
## 110	G40	G6	0.03151063	0.084998087
## 111	G17	G42	0.03147548	0.280873963
## 112	G52	G76	0.03142608	0.013721138
## 113	G13	G42	0.03112532	0.011471305
## 114	G18	G88	0.03067207	0.016998158
## 115	G46	G79	0.03061646	0.009568428
## 116	G79	G88	0.03058667	0.014618820
## 117	G16	G79	0.03045550	0.037622101
## 118	G37	G2	0.03045396	0.009204104
## 119	G25	G26	0.03039506	0.016177434
## 120	G5	G79	0.03018807	0.010922002
## 121	G8	G46	0.03013462	0.016400182
## 122	G39	G97	0.02995462	0.020425977
## 123	G37	G41	0.02990516	0.011930989
## 124	G51	G13	0.02988311	0.017857681
## 125	G11	G29	0.02977565	0.009491852
## 126	G80	G60	0.02961760	0.019747560
## 127	G18	G97	0.02949718	0.032383359
## 128	G58	G13	0.02949252	0.009288736
## 129	G43	G24	0.02948547	0.018347530
## 130	G96	G44	0.02946444	0.010797919
## 131	G83	G82	0.02941753	0.418905835
## 132	G40	G41	0.02913226	0.010459206
## 133	G37	G90	0.02890624	0.011492933
## 134	G42	G20	0.02849433	0.018953389
## 135	G80	G49	0.02846486	0.016948099
## 136	G41	G93	0.02839658	0.038047106
## 137	G24	G5	0.02818025	0.019503890
## 138	G18	G80	0.02813341	0.011304061
## 139	G37	G48	0.02812036	0.017863032
## 140	G16	G23	0.02796637	0.009208941
## 141	G69	G78	0.02790849	0.015470520
## 142	G86	G8	0.02783405	0.010243700
## 143	G58	G45	0.02773012	0.014003509
## 144	G32	G4	0.02772748	0.012981682
## 145	G39	G2	0.02757343	0.009158128
## 146	G76	G46	0.02743362	0.057231123
## 147	G99	G42	0.02743241	0.010492830
## 148	G60	G90	0.02740925	0.014537633
## 149	G18	G41	0.02740297	0.012614910

##	150	G40	G51	0.02727947	0.012082339
##	151	G86	G90	0.02725303	0.011844617
##	152	G62	G57	0.02722220	0.019749474
##	153	G79	G16	0.02698078	0.031840512
##	154	G15	G3	0.02684842	0.012039795
##	155	G40	G24	0.02673968	0.043076456
##	156	G86	G46	0.02666464	0.013304635
##	157	G9	G27	0.02665807	0.023369549
##	158	G23	G20	0.02658056	0.020575826
##	159	G67	G22	0.02654312	0.014760255
##	160	G37	G96	0.02650491	0.062020710
##	161	G55	G68	0.02626418	0.023233419
##	162	G97	G4	0.02620511	0.022084775
##	163	G38	G10	0.02618078	0.104777401
##	164	G49	G94	0.02610108	0.056201351
##	165	G6	G46	0.02598914	0.009401715
##	166	G99	G11	0.02595309	0.157582516
##	167	G38	G1	0.02584902	0.015432025
##	168	G8	G51	0.02582904	0.010972166
##	169	G6	G40	0.02578767	0.082968102
##	170	G57	G44	0.02565458	0.009182807
##	171	G23	G30	0.02564940	0.012160931
##	172	G81	G14	0.02553622	0.028191590
##	173	G88	G16	0.02545079	0.024796099
##	174	G99	G61	0.02542064	0.253702689
##	175	G87	G85	0.02541927	0.023492212
##	176	G88	G53	0.02538553	0.009763460
##	177	G57	G68	0.02526282	0.010101693
##	178	G32	G16	0.02524275	0.019783422
##	179	G46	G89	0.02521256	0.056506941
##	180	G95	G76	0.02503465	0.016012124
##	181	G77	G44	0.02499024	0.017505185
##	182	G14	G97	0.02492174	0.018670561
##	183	G24	G7	0.02487888	0.024935121
##	184	G100	G35	0.02467739	0.018547887
##	185	G39	G53	0.02455833	0.027538762
##	186	G47	G83	0.02442998	0.016769833
##	187	G62	G15	0.02439291	0.009463404
##	188	G81	G85	0.02435790	0.034181180
##	189	G61	G37	0.02434209	0.119170130
##	190	G20	G84	0.02432673	0.047375161
##	191	G83	G24	0.02425830	0.017046953
##	192	G40	G90	0.02416804	0.020720708
##	193	G59	G32	0.02413861	0.014546624
##	194	G42	G18	0.02411510	0.010119378
##	195	G42	G17	0.02401998	0.026920027
##	196	G69	G55	0.02401403	0.010038204
##	197	G43	G30	0.02382991	0.019871617
##	198	G76	G8	0.02374562	0.010659387
##	199	G4	G21	0.02366010	0.021214416
##	200	G88	G99	0.02340800	0.009097908
##	201	G94	G27	0.02328728	0.009612790
##	202	G61	G96	0.02315969	0.011370099
##	203	G90	G37	0.02297018	0.010455004

##	204	G19	G1	0.02292810	0.009300770
##	205	G86	G81	0.02291787	0.027262685
##	206	G22	G98	0.02273926	0.019581942
##	207	G32	G97	0.02266802	0.012611230
##	208	G5	G2	0.02263099	0.022866569
##	209	G32	G14	0.02251653	0.009154635
##	210	G12	G82	0.02241730	0.018755391
##	211	G79	G9	0.02241616	0.010028960
##	212	G100	G26	0.02236407	0.010134460
##	213	G19	G34	0.02222290	0.011776330
##	214	G20	G33	0.02207739	0.043474912
##	215	G98	G22	0.02195363	0.012131420
##	216	G4	G97	0.02182152	0.015194895
##	217	G13	G58	0.02165284	0.013805584
##	218	G72	G30	0.02162316	0.019717225
##	219	G17	G9	0.02154795	0.012761838
##	220	G87	G26	0.02144717	0.011536872
##	221	G65	G99	0.02142142	0.009535085
##	222	G16	G21	0.02138926	0.016677566
##	223	G77	G90	0.02130657	0.011046409
##	224	G14	G4	0.02124634	0.010807627
##	225	G22	G33	0.02115101	0.021117472
##	226	G39	G82	0.02105678	0.012661378
##	227	G64	G88	0.02105601	0.024356623
##	228	G57	G61	0.02104253	0.067656961
##	229	G30	G45	0.02098128	0.009771188
##	230	G16	G82	0.02096476	0.012933254
##	231	G49	G86	0.02095358	0.009139450
##	232	G39	G30	0.02093603	0.009215716
##	233	G9	G16	0.02081892	0.040178600
##	234	G16	G88	0.02066553	0.011588401
##	235	G65	G68	0.02065377	0.049096120
##	236	G86	G30	0.02063648	0.069965405
##	237	G16	G38	0.02060223	0.028007603
##	238	G6	G17	0.02055446	0.021048772
##	239	G16	G36	0.02047100	0.010885681
##	240	G24	G83	0.02044581	0.016858198
##	241	G49	G85	0.02038770	0.012256362
##	242	G93	G90	0.02038068	0.017469963
##	243	G40	G47	0.02034198	0.141261045
##	244	G47	G56	0.02026722	0.177109075
##	245	G85	G35	0.02018549	0.011140317
##	246	G40	G93	0.02010592	0.177495622
##	247	G38	G97	0.02009619	0.036925234
##	248	G31	G92	0.02008554	0.012237884
##	249	G46	G82	0.02008386	0.015844668
##	250	G15	G62	0.02001203	0.009740097
##	251	G84	G78	0.02000378	0.012310049
##	252	G5	G65	0.01979031	0.009137950
##	253	G65	G3	0.01974739	0.027304906
##	254	G75	G10	0.01970993	0.013818826
##	255	G96	G37	0.01961407	0.022752283
##	256	G28	G58	0.01945867	0.053877396
##	257	G45	G19	0.01943275	0.013252903

##	258	G88	G21	0.01940122	0.018776731
##	259	G35	G25	0.01939413	0.014079186
##	260	G67	G34	0.01936812	0.010128516
##	261	G28	G95	0.01936608	0.012192983
##	262	G55	G22	0.01923017	0.009782407
##	263	G37	G92	0.01912600	0.044517556
##	264	G67	G10	0.01910882	0.010617525
##	265	G51	G6	0.01906539	0.010624452
##	266	G68	G33	0.01883160	0.009982445
##	267	G99	G45	0.01880047	0.017958120
##	268	G44	G54	0.01871646	0.013526382
##	269	G61	G75	0.01863092	0.012499654
##	270	G71	G91	0.01857641	0.041809929
##	271	G68	G94	0.01850769	0.013428033
##	272	G45	G30	0.01840904	0.010668446
##	273	G38	G36	0.01840693	0.023432195
##	274	G99	G100	0.01840239	0.048980318
##	275	G22	G3	0.01837965	0.010503218
##	276	G60	G62	0.01836910	0.490542659
##	277	G83	G40	0.01834571	0.011823535
##	278	G47	G5	0.01828994	0.019559381
##	279	G51	G57	0.01822507	0.099632801
##	280	G39	G24	0.01814695	0.023800626
##	281	G97	G38	0.01813050	0.015277927
##	282	G42	G64	0.01806440	0.027929008
##	283	G32	G43	0.01803007	0.009973288
##	284	G88	G33	0.01798038	0.010835917
##	285	G17	G29	0.01786064	0.097107139
##	286	G86	G58	0.01785409	0.025114301
##	287	G31	G49	0.01783950	0.019322154
##	288	G16	G47	0.01770143	0.019120368
##	289	G31	G85	0.01768730	0.025037921
##	290	G24	G40	0.01768725	0.031242118
##	291	G96	G36	0.01767627	0.009622829
##	292	G42	G70	0.01760042	0.011465829
##	293	G28	G44	0.01753973	0.047949126
##	294	G90	G48	0.01751897	0.016806665
##	295	G59	G95	0.01748908	0.013600371
##	296	G67	G11	0.01741636	0.017026702
##	297	G26	G32	0.01740608	0.016375619
##	298	G83	G11	0.01740359	0.014736847
##	299	G94	G22	0.01737641	0.009509881
##	300	G69	G98	0.01733688	0.016012438
##	301	G43	G19	0.01727403	0.018306991
##	302	G38	G45	0.01727314	0.009411491
##	303	G46	G5	0.01726559	0.012882083
##	304	G27	G94	0.01725614	0.011002777
##	305	G83	G92	0.01723145	0.020738453
##	306	G71	G94	0.01721411	0.047587003
##	307	G18	G61	0.01718906	0.011624822
##	308	G5	G4	0.01717228	0.009567965
##	309	G94	G32	0.01716044	0.017950325
##	310	G18	G36	0.01709799	0.036432856
##	311	G56	G45	0.01708312	0.010860853

##	312	G35	G11	0.01704876	0.011520940
##	313	G48	G10	0.01703618	0.010406531
##	314	G29	G100	0.01703143	0.010128672
##	315	G82	G38	0.01701001	0.012442141
##	316	G48	G90	0.01700668	0.009955744
##	317	G53	G39	0.01693261	0.014702219
##	318	G41	G8	0.01678702	0.022236942
##	319	G41	G18	0.01675363	0.013116313
##	320	G3	G15	0.01671301	0.010452184
##	321	G5	G20	0.01663503	0.011419601
##	322	G42	G9	0.01660603	0.010819614
##	323	G5	G97	0.01656814	0.020171757
##	324	G31	G35	0.01649774	0.019609281
##	325	G61	G66	0.01649005	0.018533396
##	326	G38	G16	0.01647612	0.013182628
##	327	G68	G18	0.01641125	0.010855976
##	328	G88	G79	0.01638413	0.020097724
##	329	G93	G78	0.01636164	0.011032707
##	330	G85	G31	0.01635956	0.017563262
##	331	G29	G7	0.01634800	0.012765010
##	332	G24	G43	0.01631628	0.024979444
##	333	G23	G85	0.01627341	0.013681186
##	334	G21	G12	0.01626646	0.023751985
##	335	G100	G19	0.01626360	0.010517150
##	336	G96	G79	0.01625237	0.009406822
##	337	G42	G79	0.01623063	0.013327288
##	338	G56	G3	0.01617687	0.012521781
##	339	G8	G43	0.01617093	0.011894870
##	340	G11	G99	0.01617035	0.015549491
##	341	G75	G4	0.01616822	0.012450542
##	342	G91	G71	0.01616065	0.016870285
##	343	G16	G5	0.01614715	0.009189204
##	344	G10	G4	0.01613059	0.015743323
##	345	G94	G23	0.01613015	0.017069224
##	346	G30	G72	0.01611574	0.053427944
##	347	G53	G8	0.01607181	0.011073219
##	348	G6	G99	0.01601490	0.011450286
##	349	G28	G55	0.01598590	0.181649580
##	350	G50	G73	0.01598370	0.010864380
##	351	G40	G83	0.01597973	0.013505534
##	352	G99	G14	0.01595113	0.049069144
##	353	G86	G48	0.01589628	0.028648161
##	354	G20	G42	0.01580886	0.035178734
##	355	G68	G65	0.01577425	0.041717799
##	356	G15	G2	0.01576859	0.017831599
##	357	G9	G41	0.01574439	0.010188184
##	358	G85	G7	0.01570192	0.023189832
##	359	G40	G2	0.01569180	0.013858437
##	360	G41	G69	0.01565974	0.013018431
##	361	G77	G37	0.01562581	0.012535747
##	362	G51	G52	0.01561638	0.010140027
##	363	G5	G37	0.01560347	0.014625046
##	364	G46	G31	0.01559687	0.029798328
##	365	G23	G94	0.01556769	0.020246974

##	366	G84	G9	0.01553838	0.010527548
##	367	G13	G51	0.01551827	0.011081165
##	368	G85	G5	0.01549825	0.015431244
##	369	G12	G35	0.01545983	0.013404229
##	370	G37	G47	0.01545250	0.045107315
##	371	G7	G55	0.01543208	0.010240296
##	372	G45	G1	0.01542566	0.009488704
##	373	G86	G49	0.01530564	0.033525334
##	374	G87	G100	0.01524991	0.030413984
##	375	G1	G26	0.01523001	0.009098830
##	376	G93	G88	0.01522897	0.015783968
##	377	G38	G7	0.01518032	0.033588212
##	378	G2	G97	0.01510503	0.023785365
##	379	G37	G33	0.01507438	0.018372747
##	380	G79	G96	0.01503952	0.011882336
##	381	G22	G95	0.01501729	0.009441269
##	382	G41	G86	0.01500967	0.019368219
##	383	G39	G18	0.01496589	0.011467202
##	384	G93	G54	0.01495959	0.040548213
##	385	G81	G33	0.01494285	0.010115001
##	386	G11	G94	0.01485882	0.014269840
##	387	G90	G21	0.01484396	0.017527438
##	388	G75	G74	0.01483815	0.014888281
##	389	G21	G44	0.01483484	0.011354653
##	390	G61	G99	0.01481441	0.111345677
##	391	G47	G45	0.01475194	0.013071939
##	392	G42	G6	0.01474275	0.016596194
##	393	G98	G9	0.01473430	0.052290652
##	394	G42	G49	0.01473216	0.023939878
##	395	G12	G29	0.01472886	0.055399138
##	396	G52	G98	0.01470417	0.012008822
##	397	G19	G77	0.01470156	0.013076185
##	398	G25	G40	0.01469556	0.118798319
##	399	G26	G75	0.01466622	0.009254634
##	400	G83	G6	0.01465083	0.022116737
##	401	G80	G96	0.01460888	0.056622411
##	402	G49	G87	0.01460478	0.010889459
##	403	G96	G85	0.01459812	0.114418078
##	404	G28	G35	0.01453037	0.029399528
##	405	G18	G73	0.01452774	0.009093985
##	406	G95	G54	0.01449030	0.009172992
##	407	G46	G22	0.01446725	0.010699880
##	408	G81	G93	0.01446226	0.011179217
##	409	G17	G98	0.01444894	0.021129048
##	410	G43	G84	0.01444864	0.061568161
##	411	G100	G87	0.01442141	0.028906691
##	412	G81	G36	0.01442040	0.016381010
##	413	G10	G38	0.01438529	0.113558351
##	414	G61	G57	0.01434348	0.133824163
##	415	G19	G12	0.01427479	0.009295462
##	416	G55	G28	0.01426799	0.159091007
##	417	G70	G95	0.01425885	0.016175884
##	418	G11	G67	0.01419555	0.016403284
##	419	G90	G7	0.01417693	0.010390860

##	420	G41	G95	0.01414471	0.015615795
##	421	G21	G55	0.01414185	0.014689030
##	422	G32	G94	0.01413783	0.025243073
##	423	G57	G53	0.01407677	0.018279446
##	424	G82	G35	0.01407298	0.011527240
##	425	G2	G65	0.01406751	0.009445141
##	426	G50	G32	0.01405308	0.011462763
##	427	G72	G95	0.01403188	0.010007607
##	428	G8	G41	0.01402477	0.014535423
##	429	G99	G6	0.01402297	0.024165197
##	430	G16	G49	0.01399440	0.011428542
##	431	G26	G3	0.01394183	0.018866989
##	432	G94	G38	0.01393746	0.009729380
##	433	G45	G14	0.01386169	0.009625025
##	434	G84	G43	0.01377998	0.125070162
##	435	G43	G90	0.01377740	0.010233739
##	436	G95	G22	0.01375573	0.012812126
##	437	G36	G73	0.01374495	0.011500078
##	438	G32	G26	0.01368723	0.023958688
##	439	G40	G29	0.01367715	0.014896975
##	440	G43	G55	0.01367136	0.010550320
##	441	G60	G19	0.01366761	0.009368599
##	442	G59	G68	0.01365038	0.012601427
##	443	G88	G18	0.01364628	0.016671784
##	444	G87	G57	0.01361795	0.015294108
##	445	G81	G57	0.01345910	0.011521366
##	446	G11	G75	0.01343301	0.034559411
##	447	G80	G30	0.01338782	0.010222176
##	448	G60	G30	0.01338250	0.012412447
##	449	G59	G15	0.01338158	0.010740322
##	450	G32	G20	0.01338003	0.012928830
##	451	G61	G7	0.01334976	0.012792435
##	452	G58	G75	0.01333593	0.009302470
##	453	G14	G99	0.01328851	0.021292707
##	454	G21	G16	0.01322350	0.011323120
##	455	G47	G41	0.01321562	0.036365825
##	456	G86	G88	0.01321036	0.027742974
##	457	G25	G35	0.01317762	0.014967649
##	458	G13	G77	0.01316635	0.038919606
##	459	G97	G9	0.01315402	0.009210860
##	460	G87	G33	0.01313611	0.018300044
##	461	G17	G25	0.01311419	0.325484980
##	462	G26	G74	0.01308281	0.009638134
##	463	G64	G6	0.01304450	0.021379716
##	464	G44	G86	0.01303411	0.010007935
##	465	G87	G88	0.01303317	0.016424158
##	466	G21	G4	0.01299501	0.010667449
##	467	G18	G68	0.01294368	0.014234630
##	468	G34	G65	0.01293528	0.024850227
##	469	G50	G60	0.01293454	0.072773834
##	470	G100	G29	0.01292581	0.020570792
##	471	G64	G90	0.01291201	0.011138635
##	472	G97	G32	0.01290020	0.009282077
##	473	G41	G99	0.01289375	0.020234334

##	474	G54	G69	0.01285808	0.011157360
##	475	G78	G75	0.01282698	0.024284713
##	476	G96	G5	0.01281734	0.053446627
##	477	G86	G7	0.01281718	0.014462333
##	478	G76	G95	0.01280227	0.024591993
##	479	G35	G100	0.01280102	0.018013066
##	480	G25	G86	0.01278824	0.018993770
##	481	G3	G98	0.01277875	0.016666734
##	482	G5	G96	0.01276398	0.010420387
##	483	G73	G22	0.01273447	0.049359132
##	484	G24	G8	0.01268931	0.011001013
##	485	G49	G54	0.01268750	0.011084513
##	486	G35	G76	0.01267817	0.010945475
##	487	G42	G56	0.01266410	0.122406268
##	488	G23	G98	0.01265348	0.009966886
##	489	G97	G60	0.01264488	0.015374102
##	490	G5	G19	0.01262581	0.010131271
##	491	G27	G9	0.01262388	0.021090441
##	492	G53	G72	0.01260598	0.010837525
##	493	G55	G7	0.01257086	0.011098057
##	494	G22	G77	0.01256433	0.009695280
##	495	G24	G55	0.01255827	0.018003745
##	496	G65	G26	0.01255357	0.018274407
##	497	G6	G59	0.01255183	0.020819578
##	498	G71	G97	0.01250104	0.010743000
##	499	G63	G4	0.01246245	0.011394948
##	500	G85	G1	0.01243102	0.013259278
##	501	G87	G22	0.01242173	0.021061409
##	502	G94	G11	0.01238433	0.012199998
##	503	G46	G19	0.01238305	0.016645259
##	504	G13	G78	0.01238154	0.021196908
##	505	G22	G35	0.01235661	0.010876488
##	506	G15	G65	0.01234151	0.043810183
##	507	G9	G49	0.01230266	0.016856883
##	508	G98	G77	0.01229857	0.011133065
##	509	G78	G76	0.01229598	0.017456808
##	510	G99	G18	0.01228614	0.011249321
##	511	G42	G71	0.01224624	0.024090593
##	512	G68	G19	0.01223817	0.016180475
##	513	G47	G53	0.01221548	0.027245233
##	514	G5	G36	0.01218814	0.018602665
##	515	G8	G74	0.01218394	0.155015810
##	516	G96	G80	0.01216915	0.048234874
##	517	G65	G34	0.01212646	0.017640514
##	518	G16	G26	0.01209769	0.009709399
##	519	G63	G61	0.01209747	0.025714881
##	520	G13	G52	0.01208915	0.032781430
##	521	G38	G87	0.01208392	0.010649580
##	522	G54	G32	0.01207869	0.018169013
##	523	G19	G95	0.01204624	0.014157374
##	524	G4	G32	0.01203124	0.026639951
##	525	G28	G33	0.01199002	0.011286573
##	526	G72	G55	0.01197957	0.032077539
##	527	G57	G90	0.01197922	0.013947796

##	528	G9	G97	0.01196357	0.016172742
##	529	G30	G42	0.01196186	0.014151602
##	530	G65	G67	0.01195916	0.022287479
##	531	G1	G25	0.01195531	0.018248083
##	532	G9	G80	0.01195208	0.031935122
##	533	G12	G68	0.01194153	0.010683526
##	534	G46	G78	0.01194090	0.013394220
##	535	G9	G79	0.01193852	0.014879857
##	536	G51	G44	0.01190971	0.010365370
##	537	G52	G69	0.01190916	0.046502161
##	538	G58	G28	0.01190019	0.036167119
##	539	G4	G34	0.01189736	0.017816258
##	540	G14	G79	0.01188854	0.011826651
##	541	G57	G60	0.01188579	0.178531193
##	542	G54	G52	0.01186704	0.030196421
##	543	G59	G76	0.01186342	0.011738710
##	544	G71	G74	0.01186058	0.202565284
##	545	G67	G65	0.01185456	0.055414176
##	546	G5	G30	0.01184310	0.009084817
##	547	G42	G98	0.01183430	0.009586116
##	548	G24	G51	0.01178964	0.094530260
##	549	G7	G12	0.01178885	0.013392963
##	550	G25	G29	0.01178818	0.153889994
##	551	G1	G95	0.01177757	0.011857955
##	552	G67	G72	0.01177642	0.122266908
##	553	G27	G38	0.01177181	0.009919759
##	554	G29	G25	0.01174107	0.075111003
##	555	G65	G89	0.01174094	0.030125698
##	556	G56	G27	0.01166361	0.011876149
##	557	G17	G81	0.01165990	0.054889798
##	558	G98	G33	0.01165531	0.011394872
##	559	G45	G99	0.01165213	0.015756204
##	560	G69	G33	0.01163671	0.010043605
##	561	G69	G12	0.01163168	0.017258974
##	562	G29	G9	0.01162484	0.011359451
##	563	G90	G20	0.01160649	0.009430369
##	564	G25	G2	0.01160560	0.013343176
##	565	G57	G29	0.01159388	0.013165255
##	566	G50	G6	0.01158999	0.009595716
##	567	G44	G28	0.01158147	0.031678435
##	568	G22	G13	0.01154797	0.014924112
##	569	G85	G20	0.01154639	0.017283133
##	570	G98	G48	0.01154057	0.025002217
##	571	G93	G30	0.01152947	0.009394704
##	572	G81	G97	0.01151711	0.013524157
##	573	G15	G66	0.01150310	0.015762558
##	574	G31	G70	0.01148153	0.015493325
##	575	G5	G88	0.01147999	0.016077248
##	576	G56	G7	0.01147874	0.009616466
##	577	G57	G54	0.01147032	0.009470275
##	578	G83	G58	0.01145307	0.027721954
##	579	G90	G56	0.01141404	0.010330381
##	580	G77	G19	0.01141376	0.031585800
##	581	G34	G33	0.01141181	0.023626918

##	582	G85	G87	0.01140138	0.009911849
##	583	G98	G19	0.01139896	0.027799950
##	584	G85	G49	0.01139823	0.018906174
##	585	G6	G22	0.01139788	0.012500515
##	586	G56	G74	0.01137856	0.045079057
##	587	G81	G58	0.01137696	0.024295348
##	588	G26	G67	0.01137115	0.011478655
##	589	G32	G72	0.01135161	0.012024176
##	590	G21	G76	0.01134108	0.012540520
##	591	G29	G12	0.01133755	0.069859768
##	592	G10	G34	0.01133539	0.020846839
##	593	G67	G33	0.01130666	0.070128155
##	594	G98	G75	0.01130464	0.013704869
##	595	G96	G86	0.01129394	0.015259079
##	596	G42	G73	0.01128977	0.053787052
##	597	G40	G25	0.01128472	0.022844993
##	598	G55	G4	0.01127368	0.012096818
##	599	G74	G43	0.01126948	0.022642514
##	600	G80	G16	0.01126417	0.030648737
##	601	G57	G51	0.01125322	0.046957801
##	602	G42	G61	0.01122055	0.053521878
##	603	G44	G7	0.01120900	0.013840058
##	604	G56	G92	0.01120039	0.009130645
##	605	G35	G72	0.01119410	0.010141225
##	606	G76	G78	0.01118380	0.052332750
##	607	G50	G100	0.01118351	0.018616532
##	608	G82	G3	0.01115564	0.010954743
##	609	G47	G40	0.01113746	0.019852665
##	610	G44	G85	0.01113050	0.012978914
##	611	G79	G18	0.01112511	0.009838404
##	612	G31	G71	0.01111351	0.033590976
##	613	G25	G68	0.01108526	0.009981104
##	614	G8	G86	0.01107934	0.024254381
##	615	G53	G19	0.01104490	0.015081495
##	616	G52	G54	0.01103989	0.209010834
##	617	G76	G28	0.01102084	0.219401489
##	618	G13	G22	0.01101281	0.020753609
##	619	G55	G95	0.01101129	0.017296392
##	620	G82	G61	0.01099485	0.009630219
##	621	G81	G79	0.01096874	0.014678699
##	622	G85	G44	0.01095537	0.018304924
##	623	G5	G90	0.01094671	0.011904048
##	624	G68	G77	0.01091604	0.015218679
##	625	G55	G62	0.01091348	0.014241262
##	626	G27	G16	0.01091042	0.038545102
##	627	G12	G6	0.01090995	0.044843712
##	628	G27	G33	0.01090853	0.011222186
##	629	G76	G73	0.01090572	0.017546840
##	630	G90	G100	0.01087339	0.011563191
##	631	G69	G6	0.01087143	0.083811480
##	632	G33	G17	0.01083549	0.012155122
##	633	G13	G98	0.01082936	0.009583639
##	634	G51	G9	0.01082820	0.009158000
##	635	G26	G10	0.01080367	0.015016506

##	636	G4	G75	0.01079443	0.022248927
##	637	G58	G27	0.01078080	0.013631046
##	638	G38	G27	0.01077737	0.010191054
##	639	G21	G88	0.01076828	0.119563008
##	640	G55	G26	0.01076706	0.010049733
##	641	G80	G26	0.01076270	0.009355551
##	642	G32	G1	0.01075956	0.102490372
##	643	G22	G4	0.01075588	0.009092789