Fit Qin-lab natural isolates using bootstraps

h qin 2017 April 24 - June 11

```
rm(list=ls())
host = "Applejack" #"Ridgeside"
if (host == "AppleJack") {
    setwd("/Users/hqin/github/bmc_netwk_aging_manuscript/R1/0.nat.rls.fitting");
}
if (host == "Ridgeside") {
    setwd("/home/hqin/github/bmc_netwk_aging_manuscript/R1/0.nat.rls.fitting");
}
library('flexsurv')

## Loading required package: survival
source("../lifespan.r")

Set seed for consistency for publication purpose.
set.seed(20170101) #remove to test for different bootstraps
```

parse the strains from files

```
RUNS = 100; #bootstrap runs
files = list.files(path="../qinlab_rls/", pattern="rls.tab")
tmp1 = gsub("\\d{6}.", "", files)
redundant_strains = gsub(".rls.tab", "", tmp1)
strains = sort( unique( redundant_strains ))
strains
## [1] "101S"
                         "BY4716"
                                          "BY4741"
                                                           "BY4742"
## [5] "BY4743"
                        "JSBY4741"
                                          "M1-2"
                                                           "M13"
## [9] "M14"
                                          "M22"
                                                           "M32"
                        "M2-8"
## [13] "M34"
                        "M5"
                                          "M8"
                                                           "RM112N"
## [17] "S288c"
                                          "sir2D.4741a"
                         "SGU57"
                                                           "sir2D.4742"
                                                           "YPS128"
                                         "W303"
## [21] "sir2DSIR2.4742" "SK1"
## [25] "YPS163"
```

Take files from natural isolates

```
my.strains=c("101S", "M1-2", "M13", "M14", "M2-8", "M22", "M32", "M34", "M5", "M8", "RM112N", "S288c", "SGU57", "Y. files2=c();
for( i in 1:length(my.strains)){
   files2 = c( files2, files[grep(my.strains[i], files)]);
}
```

```
report = data.frame(cbind(my.strains))
report$samplesize = NA; report$R=NA; report$t0=NA; report$n=NA; report$G=NA; report$longfilename=NA;
files = files2;
strains = my.strains;
```

Now, fit all RLS data sets by strains

```
for( BootstrapCount in 1:RUNS ) {#!!!!!!
for( i in 1:length(report[,1])){
#for( i in 3:4){
 my.files = files[grep(strains[i], files)]
  report$longfilename[i] = paste(my.files, collapse = "::");
  tb = read.table( paste("../qinlab_rls/",my.files[1],sep=''), sep="\t")
  if( length(my.files)> 1){
   for( fi in 2:length(my.files)) {
      tmp.tb = read.table( paste("../qinlab_rls/",my.files[fi],sep=''), sep="\t")
     tb = rbind( tb, tmp.tb)
   }
  }
  report$samplesize[i] = length(tb[,1])
  #boostrap
  tb[,1] = sample(tb[,1], replace=TRUE); # BOOTSTRAP HERE
  GompFlex = flexsurvreg(formula = Surv(tb[,1]) ~ 1, dist = 'gompertz')
  WeibFlex = flexsurvreg(formula = Surv(tb[,1]) ~ 1, dist = 'weibull')
  report$avgLS[i] = mean(tb[,1])
  report$stdLS[i] = sd(tb[,1])
  report$CV[i] = report$stdLS[i] / report$avgLS[i]
  report$GompGFlex[i] = GompFlex$res[1,1]
  report$GompRFlex[i] = GompFlex$res[2,1]
  report$GompLogLikFlex[i] = round(GompFlex$loglik, 1)
  report$GompAICFlex[i] = round(GompFlex$AIC)
  report$WeibShapeFlex[i] = WeibFlex$res[1,1]
  report$WeibRateFlex[i] = WeibFlex$res[2,1]
  report$WeibLogLikFlex[i] = round(WeibFlex$loglik, 1)
  report$WeibAICFlex[i] = round(WeibFlex$AIC)
  #set initial values
  Rhat = report$GompRFlex[i]; # 'i' was missing. a bug costed HQ a whole afternoon.
  Ghat = report$GompGFlex[i];
  nhat = 6;
  t0= (nhat-1)/Ghat;
  fitBinom = optim (c(Rhat, t0, nhat), llh.binomialMortality.single.run,
                     lifespan=tb[,1],
                     #method='SANN') #SANN needs control
```

summarize boostrap results.

Pick row-col value from every file into a buffer, then mean and stddev.

```
BootstrapMean = report;
BootstrapStd = report;
rownames = names(report);
for( col in 2:length(report[1,])) {
  for ( row in 1:length(report[,1])) {
   buffer = c();
   for( BootstrapCount in 1:RUNS ) {
      filename = paste("bootstrap/", BootstrapCount, ".csv", sep='');
     tb = read.csv(filename)
      if( rownames[col] == "BestModel" ) {
       buffer = as.character( c(buffer, as.character(tb[row,col])) );
       buffer = c(buffer, tb[row, col]);
   }
    if( rownames[col] == "BestModel" ) {
      tmp = table( buffer );
      BootstrapMean[row,col] = paste(names(tmp), tmp, sep="=", collapse = ":");
     BootstrapMean[row,col] = mean(buffer);
     BootstrapStd[row,col] = sqrt(var(buffer));
 }
}
```

Means

BootstrapMean

```
my.strains samplesize
##
                                              t0
## 1
                         85 0.002505451 35.99604 6.686567 0.16043672
            101S
## 2
           M1-2
                         54 0.003389155 40.47858 7.077706 0.15126428
## 3
            M13
                        70 0.003418251 40.80201 7.364152 0.15665075
## 4
            M14
                         60 0.003484255 55.06556 6.558044 0.10172215
           M2-8
                        105 0.003409035 42.35209 7.998702 0.16660562
## 5
                         60 0.003311774 46.31594 6.873676 0.12749105
## 6
            M22
                         60 0.002698475 34.32367 7.039921 0.17693307
## 7
            M32
```

```
58 0.002762630 31.44145 6.677334 0.18155728
## 8
             M34
## 9
              M5
                         166 0.003434030 73.99028 7.761917 0.09198913
                          60 0.001828786 30.65496 6.077315 0.16688372
## 10
              M8
## 11
          RM112N
                          59 0.002527789 55.44863 6.202258 0.09464606
## 12
           S288c
                          41 0.005077696 56.77646 7.885613 0.12478848
## 13
                          58 0.006519542 58.03551 7.879940 0.12195836
           SGU57
## 14
          YPS128
                          69 0.002613473 41.90105 6.475758 0.13165058
                         130 0.002296450 37.25939 6.404451 0.14547179
## 15
          YPS163
##
      longfilename
                       avgLS
                                 stdLS
                                               CV
                                                   GompGFlex
                                                                 GompRFlex
## 1
                              7.414184 0.2360298 0.14183796 0.0011886014
                 15 31.45824
##
                 9 27.89889
                              9.088393 0.3271839 0.12535089 0.0025636829
## 3
                   26.59871
                              9.150567 0.3452559 0.12325312 0.0030381156
## 4
                 10 36.32417 12.787126 0.3531547 0.09198143 0.0021270344
## 5
                  4 24.76743 8.056250 0.3255815 0.11812658 0.0043352445
## 6
                11 32.07050 10.140110 0.3171061 0.11150913 0.0020152844
## 7
                 6 28.14783
                              6.664404 0.2368370 0.14566819 0.0015725113
## 8
                              8.062000 0.2993340 0.16048962 0.0013442040
                 1 27.02103
## 9
                 5 36.85289 12.990472 0.3527247 0.06803908 0.0040096710
## 10
                             6.781575 0.1951294 0.16453160 0.0003590377
                 8 34.78883
## 11
                 12 44.13085 12.975729 0.2948886 0.09121338 0.0010544096
## 12
                 14 26.36390 10.201881 0.3879796 0.09101058 0.0062074368
## 13
                 13 23.89845 10.516315 0.4414499 0.09131664 0.0077070530
## 14
                 2 35.08014
                              9.586650 0.2738812 0.12013655 0.0011318327
## 15
                  3 34.41446 8.563412 0.2490014 0.13442802 0.0008118923
##
      GompLogLikFlex GompAICFlex WeibShapeFlex WeibRateFlex WeibLogLikFlex
## 1
            -292.604
                           589.20
                                       4.941140
                                                     34.24623
                                                                     -289.332
## 2
            -192.909
                           389.76
                                        3.635913
                                                     30.95623
                                                                     -194.592
## 3
            -250.000
                           504.00
                                        3.162652
                                                     29.31791
                                                                     -258.419
## 4
                                                     40.53819
                                                                     -236.454
            -233.435
                           470.91
                                       3.369193
## 5
            -372.001
                           748.01
                                       3.417174
                                                     27.54976
                                                                     -367.165
## 6
            -221.673
                           447.31
                                       3.713709
                                                     35.57107
                                                                     -222.881
## 7
            -203.847
                           411.68
                                       4.597658
                                                     30.76086
                                                                     -199.428
## 8
            -196.896
                           397.81
                                       4.124638
                                                     29.73866
                                                                     -202.132
## 9
            -669.603
                          1343.21
                                       3.120367
                                                     41.22948
                                                                     -658.682
## 10
            -200.250
                           404.45
                                       6.028250
                                                     37.50598
                                                                     -198.654
## 11
            -231.399
                           466.79
                                       4.104189
                                                     48.67951
                                                                     -233.248
## 12
            -153.402
                           310.80
                                        2.856687
                                                     29.48898
                                                                     -152.826
## 13
            -215.925
                           435.81
                                        2.487367
                                                     26.87244
                                                                     -217.746
## 14
            -251.437
                           506.89
                                        4.383940
                                                     38.57445
                                                                     -251.798
## 15
            -459.429
                           922.83
                                        4.836988
                                                                     -459.809
                                                     37.60504
                               BestModel
##
      WeibAICFlex
## 1
           582.72 <2=10:Gomp=16:Weib=74
## 2
           393.19
                    <2=25:Gomp=69:Weib=6
## 3
           520.85
                     <2=5:Gomp=88:Weib=7
## 4
           476.90
                     <2=7:Gomp=91:Weib=2
## 5
           738.35
                     <2=5:Gomp=2:Weib=93
## 6
           449.76 <2=32:Gomp=51:Weib=17
## 7
           402.89
                            <2=1:Weib=99
## 8
           408.27
                            <2=7:Gomp=93
## 9
          1321.34
                                Weib=100
## 10
           401.29
                    <2=22:Gomp=6:Weib=72
## 11
           470.48
                   <2=23:Gomp=70:Weib=7
## 12
           309.61 <2=31:Gomp=26:Weib=43
## 13
           439.51 <2=15:Gomp=73:Weib=12
```

```
## 14 507.62 <2=38:Gomp=40:Weib=22
## 15 923.64 <2=21:Gomp=44:Weib=35
```

StdDev

BootstrapStd

```
G
##
      my.strains samplesize
                                        R
                                                  t0
## 1
            101S
                           0 0.0009287405
                                           5.479713 0.6974831 0.022866994
## 2
            M1-2
                           0 0.0009980579
                                           5.276955 0.7602881 0.017968586
## 3
             M13
                           0 0.0009459944
                                           4.195255 0.6396622 0.015279534
## 4
             M14
                           0 0.0010014401
                                           6.604037 0.6372519 0.012057134
## 5
            M2 - 8
                                           3.989513 0.1091360 0.014718183
                           0 0.0006344560
## 6
             M22
                           0 0.0010404733 11.837460 1.5075316 0.015948835
## 7
                                           2.586144 0.7143603 0.024292330
             M32
                           0 0.0011194964
## 8
             M34
                            0.0007362679
                                           3.544114 0.6159757 0.018066724
## 9
              M5
                           0 0.0005916447
                                           7.219866 0.3813192 0.007859491
## 10
                           0 0.0003484946
                                          2.919832 0.2071864 0.014669925
              M8
## 11
          RM112N
                           0 0.0006463488 5.986260 0.3746111 0.009729632
## 12
           S288c
                           0 0.0016711154 12.922973 1.3414738 0.023097209
## 13
                           0 0.0021172202 19.651135 1.5046710 0.017847788
           SGU57
## 14
          YPS128
                           0 0.0007754037
                                           3.888046 0.4863120 0.015269929
  15
##
          YPS163
                           0 0.0004804929
                                           2.624209 0.4280389 0.011935765
##
      longfilename
                        avgLS
                                  stdLS
                                                 CV
                                                     GompGFlex
                                                                   GompRFlex
                 0 0.8145652 0.6935133 0.02504680 0.02380020 0.0006263407
## 1
## 2
                 0 1.2904696 0.8190490 0.03951375 0.01707491 0.0011734232
## 3
                 0 1.0640702 0.8726972 0.04234235 0.01207602 0.0010961058
##
                 0 1.6213699 0.8832416 0.03409446 0.01068784 0.0009291640
## 5
                 0 0.7334054 0.5917975 0.02604108 0.01095486 0.0010385633
##
                 0 1.3592553 0.7678484 0.03111031 0.01055153 0.0007328389
  6
##
  7
                 0 0.9733216 0.6534801 0.02281649 0.01109063 0.0004621282
##
                 0 0.9969735 0.8140843 0.03715807 0.01798446 0.0006980044
  8
## 9
                 0 0.9422285 0.5958022 0.01861178 0.00679012 0.0008521899
## 10
                 0 0.9690498 0.6401819 0.01971980 0.01594485 0.0001833472
## 11
                 0 1.7506204 1.1804215 0.03295971 0.01003443 0.0004660643
##
  12
                 0 1.5006540 1.1151878 0.04538956 0.01579110 0.0020188426
##
  13
                 0 1.3190327 0.8063968 0.04257855 0.01148507 0.0023363275
##
  14
                 0 1.1245440 0.8631905 0.02934451 0.01138383 0.0004510274
                 0 0.6989248 0.4736352 0.01567590 0.01000082 0.0002297927
##
   15
      GompLogLikFlex GompAICFlex WeibShapeFlex WeibRateFlex WeibLogLikFlex
##
## 1
           10.040795
                        20.103269
                                      0.6118343
                                                    0.7991631
                                                                    8.453998
                        10.111619
                                                                     5.264550
## 2
            5.052674
                                      0.5445300
                                                    1.2478190
## 3
            5.107392
                        10.197048
                                      0.5175619
                                                    1.1412121
                                                                    9.055014
## 4
            4.422243
                         8.859276
                                      0.4008239
                                                    1.6226350
                                                                     4.283122
## 5
            6.815764
                        13.635585
                                      0.2868792
                                                    0.7848509
                                                                    7.468214
## 6
            4.054075
                         8.143529
                                      0.4353666
                                                    1.3545984
                                                                    4.936643
## 7
            4.313953
                                      0.3661096
                                                    1.0628450
                                                                    5.162590
                         8.673517
## 8
            5.446413
                        10.926471
                                      0.6325299
                                                    0.9340973
                                                                    7.057482
## 9
            8.619410
                        17.228878
                                      0.1810164
                                                    1.0062292
                                                                    7.133282
## 10
            5.314465
                        10.603387
                                      0.5951988
                                                    0.9761665
                                                                    5.354530
## 11
                        10.270380
                                                                    5.692623
            5.138201
                                      0.5215999
                                                    1.7111870
## 12
            4.135825
                         8.298716
                                      0.4090500
                                                                    4.664818
                                                    1.6389505
## 13
            3.544191
                        7.107678
                                      0.2945314
                                                    1.4349958
                                                                    4.403819
## 14
            5.665947
                        11.381267
                                      0.5023200
                                                    1.0705033
                                                                    6.280722
## 15
            7.528428
                        15.066891
                                      0.3347178
                                                    0.6817798
                                                                    7.470658
```

```
##
      WeibAICFlex BestModel
## 1
        16.946189
                        Weib
## 2
        10.509923
                        Gomp
## 3
        18.131604
                          <2
## 4
         8.552275
                        Gomp
## 5
        14.919565
                        Weib
## 6
         9.867936
                        Weib
## 7
        10.321004
                        Weib
## 8
        14.105558
                        Gomp
## 9
        14.319740
                        Weib
## 10
        10.710417
                          <2
## 11
        11.353351
                        Gomp
## 12
         9.314803
                        Weib
         8.840329
## 13
                        Gomp
## 14
        12.550327
                        Gomp
## 15
        14.965631
                          <2
Merge the two tables
BootstrapMean$Rstd = BootstrapStd$R
BootstrapMean$t0std = BootstrapStd$t0
BootstrapMean$nstd = BootstrapStd$n
BootstrapMean$Gstd = BootstrapStd$G
names(BootstrapMean)
                                            "R"
                                                              "t0"
##
    [1] "my.strains"
                          "samplesize"
    [5] "n"
                          "G"
                                                              "avgLS"
##
                                            "longfilename"
   [9] "stdLS"
                          "CV"
##
                                            "GompGFlex"
                                                              "GompRFlex"
## [13] "GompLogLikFlex" "GompAICFlex"
                                            "WeibShapeFlex"
                                                              "WeibRateFlex"
## [17] "WeibLogLikFlex" "WeibAICFlex"
                                            "BestModel"
                                                              "Rstd"
## [21] "t0std"
                          "nstd"
                                            "Gstd"
Reorganized the columns
BootstrapMean = BootstrapMean[, c( "my.strains", "samplesize", "R", "Rstd", "t0",
                                                                                         "t0std",
                                    "Gstd", "avgLS",
                                                       "stdLS",
                                                                  "BestModel",
                                    "longfilename",
                                                       "CV", "GompGFlex", "GompRFlex", "GompLogLikFlex
 "GompAICFlex",
                                    "WeibRateFlex",
                                                     "WeibLogLikFlex", "WeibAICFlex" )];
                 "WeibShapeFlex",
{\tt BootstrapMean}
##
      my.strains samplesize
                                       R
                                                  Rstd
                                                                     t0std
## 1
            101S
                          85 0.002505451 0.0009287405 35.99604
                                                                 5.479713
## 2
            M1-2
                          54 0.003389155 0.0009980579 40.47858
                                                                 5.276955
## 3
             M13
                          70 0.003418251 0.0009459944 40.80201
                                                                 4.195255
## 4
             M14
                          60 0.003484255 0.0010014401 55.06556
                                                                 6.604037
                         105 0.003409035 0.0006344560 42.35209 3.989513
## 5
            M2 - 8
## 6
             M22
                          60 0.003311774 0.0010404733 46.31594 11.837460
## 7
                          60 0.002698475 0.0011194964 34.32367
             M32
                                                                 2.586144
                          58 0.002762630 0.0007362679 31.44145
## 8
             M34
                                                                 3.544114
## 9
              M5
                         166 0.003434030 0.0005916447 73.99028
                                                                 7.219866
                          60 0.001828786 0.0003484946 30.65496
## 10
              M8
## 11
          RM112N
                          59 0.002527789 0.0006463488 55.44863
                                                                 5.986260
```

69 0.002613473 0.0007754037 41.90105

41 0.005077696 0.0016711154 56.77646 12.922973 58 0.006519542 0.0021172202 58.03551 19.651135

130 0.002296450 0.0004804929 37.25939 2.624209

3.888046

12

13

14

15

S288c

SGU57

YPS128

YPS163

```
##
                                   G
                                            Gstd
                                                     avgLS
                                                               stdLS
                    nstd
             n
     6.686567 0.6974831 0.16043672 0.022866994 31.45824
## 1
                                                            7.414184
     7.077706 0.7602881 0.15126428 0.017968586 27.89889
                                                            9.088393
     7.364152 0.6396622 0.15665075 0.015279534 26.59871
                                                            9.150567
      6.558044 0.6372519 0.10172215 0.012057134 36.32417 12.787126
## 5
     7.998702 0.1091360 0.16660562 0.014718183 24.76743
                                                           8.056250
     6.873676 1.5075316 0.12749105 0.015948835 32.07050 10.140110
      7.039921 0.7143603 0.17693307 0.024292330 28.14783
                                                            6.664404
      6.677334 0.6159757 0.18155728 0.018066724 27.02103
                                                            8.062000
      7.761917 0.3813192 0.09198913 0.007859491 36.85289 12.990472
## 10 6.077315 0.2071864 0.16688372 0.014669925 34.78883
## 11 6.202258 0.3746111 0.09464606 0.009729632 44.13085 12.975729
  12 7.885613 1.3414738 0.12478848 0.023097209 26.36390 10.201881
## 13 7.879940 1.5046710 0.12195836 0.017847788 23.89845 10.516315
## 14 6.475758 0.4863120 0.13165058 0.015269929 35.08014 9.586650
## 15 6.404451 0.4280389 0.14547179 0.011935765 34.41446
                                                           8.563412
##
                  BestModel longfilename
                                                  CV GompGFlex
                                                                    GompRFlex
##
      <2=10:Gomp=16:Weib=74
                                       15 0.2360298 0.14183796 0.0011886014
  1
                                        9 0.3271839 0.12535089 0.0025636829
##
  2
       <2=25:Gomp=69:Weib=6
##
  3
        <2=5:Gomp=88:Weib=7
                                        7 0.3452559 0.12325312 0.0030381156
## 4
        <2=7:Gomp=91:Weib=2
                                       10 0.3531547 0.09198143 0.0021270344
## 5
                                        4 0.3255815 0.11812658 0.0043352445
        <2=5:Gomp=2:Weib=93
                                       11 0.3171061 0.11150913 0.0020152844
## 6
      <2=32:Gomp=51:Weib=17
                                        6 0.2368370 0.14566819 0.0015725113
## 7
               <2=1:Weib=99
## 8
               <2=7:Gomp=93
                                        1 0.2993340 0.16048962 0.0013442040
## 9
                   Weib=100
                                        5 0.3527247 0.06803908 0.0040096710
       <2=22:Gomp=6:Weib=72
                                        8 0.1951294 0.16453160 0.0003590377
## 10
## 11
       <2=23:Gomp=70:Weib=7
                                       12 0.2948886 0.09121338 0.0010544096
                                       14 0.3879796 0.09101058 0.0062074368
  12 <2=31:Gomp=26:Weib=43
## 13 <2=15:Gomp=73:Weib=12
                                       13 0.4414499 0.09131664 0.0077070530
  14 <2=38:Gomp=40:Weib=22
                                        2 0.2738812 0.12013655 0.0011318327
##
   15 <2=21:Gomp=44:Weib=35
                                        3 0.2490014 0.13442802 0.0008118923
##
      GompLogLikFlex GompAICFlex WeibShapeFlex WeibRateFlex WeibLogLikFlex
## 1
            -292.604
                          589.20
                                       4.941140
                                                     34.24623
                                                                    -289.332
  2
##
            -192.909
                           389.76
                                       3.635913
                                                     30.95623
                                                                    -194.592
## 3
            -250.000
                          504.00
                                       3.162652
                                                     29.31791
                                                                    -258.419
## 4
            -233.435
                           470.91
                                       3.369193
                                                     40.53819
                                                                    -236.454
## 5
            -372.001
                          748.01
                                                     27.54976
                                                                    -367.165
                                       3.417174
## 6
                           447.31
                                                                    -222.881
            -221.673
                                       3.713709
                                                     35.57107
## 7
            -203.847
                           411.68
                                                     30.76086
                                                                    -199.428
                                       4.597658
## 8
            -196.896
                          397.81
                                       4.124638
                                                     29.73866
                                                                    -202.132
## 9
                                                                    -658.682
            -669.603
                          1343.21
                                       3.120367
                                                     41.22948
## 10
            -200.250
                           404.45
                                       6.028250
                                                     37.50598
                                                                    -198.654
## 11
                                                     48.67951
                                                                    -233.248
            -231.399
                           466.79
                                       4.104189
## 12
            -153.402
                           310.80
                                       2.856687
                                                     29.48898
                                                                    -152.826
## 13
                           435.81
                                                                    -217.746
            -215.925
                                       2.487367
                                                     26.87244
##
  14
            -251.437
                           506.89
                                       4.383940
                                                     38.57445
                                                                    -251.798
##
  15
            -459.429
                           922.83
                                       4.836988
                                                     37.60504
                                                                    -459.809
##
      WeibAICFlex
## 1
           582.72
## 2
           393.19
## 3
           520.85
## 4
           476.90
## 5
           738.35
```

```
## 6
            449.76
## 7
           402.89
           408.27
## 8
## 9
          1321.34
## 10
           401.29
## 11
           470.48
## 12
            309.61
## 13
           439.51
## 14
           507.62
## 15
           923.64
```

Merge mean and std for publication

```
BootstrapMeanPublishing = data.frame( BootstrapMean[, c("my.strains")] )
BootstrapMeanPublishing$RwithStd = as.vector( paste(round(BootstrapMean$R, 4), round(BootstrapMean$Rstd BootstrapMeanPublishing$t0withStd = as.vector( paste(round(BootstrapMean$t0, 1), round(BootstrapMean$t0 BootstrapMeanPublishing$nwithStd = as.vector( paste(round(BootstrapMean$n, 1), round(BootstrapMean$nstd BootstrapMeanPublishing$GwithStd = as.vector( paste(round(BootstrapMean$G, 2), round(BootstrapMean$Gstd
```

BootstrapMeanPublishing

```
##
                                                RwithStd
                                                             t0withStd
      BootstrapMean...c..my.strains...
## 1
                                  101S 0.0025 +/- 9e-04
                                                            36 +/- 5.5
## 2
                                  M1-2 \quad 0.0034 +/- 0.001
                                                          40.5 +/- 5.3
## 3
                                        0.0034 +/- 9e-04
                                                          40.8 +/- 4.2
                                   M13
## 4
                                   M14
                                        0.0035 +/- 0.001
                                                          55.1 +/- 6.6
## 5
                                  M2-8 0.0034 +/- 6e-04
                                                             42.4 + / - 4
## 6
                                   M22 0.0033 +/- 0.001 46.3 +/- 11.8
## 7
                                   M32 0.0027 +/- 0.0011
                                                          34.3 +/- 2.6
                                   M34 0.0028 +/- 7e-04
                                                          31.4 +/- 3.5
## 8
                                                            74 +/- 7.2
## 9
                                       0.0034 +/- 6e-04
                                    M8 0.0018 +/- 3e-04
                                                          30.7 +/- 2.9
## 10
                                RM112N 0.0025 +/- 6e-04
## 11
                                                            55.4 +/- 6
                                 S288c 0.0051 +/- 0.0017 56.8 +/- 12.9
## 12
                                                           58 +/- 19.7
## 13
                                 SGU57 0.0065 +/- 0.0021
## 14
                                YPS128
                                       0.0026 +/- 8e-04 41.9 +/- 3.9
                                YPS163 0.0023 +/- 5e-04 37.3 +/- 2.6
## 15
           nwithStd
##
                          GwithStd
## 1 6.7 +/- 0.697 0.16 +/- 0.023
      7.1 +/- 0.76 0.15 +/- 0.018
## 3
      7.4 +/- 0.64 0.16 +/- 0.015
## 4
     6.6 +/- 0.637 0.1 +/- 0.012
        8 +/- 0.109 0.17 +/- 0.015
     6.9 +/- 1.508 0.13 +/- 0.016
## 6
        7 +/- 0.714 0.18 +/- 0.024
## 8 6.7 +/- 0.616 0.18 +/- 0.018
## 9 7.8 +/- 0.381 0.09 +/- 0.008
## 10 6.1 +/- 0.207 0.17 +/- 0.015
## 11 6.2 +/- 0.375 0.09 +/- 0.01
## 12 7.9 +/- 1.341 0.12 +/- 0.023
## 13 7.9 +/- 1.505 0.12 +/- 0.018
## 14 6.5 +/- 0.486 0.13 +/- 0.015
## 15 6.4 +/- 0.428 0.15 +/- 0.012
```

output

```
write.csv(BootstrapMean, file="sandbox/Bootstrap_summary.csv", row.names = FALSE)
write.csv(BootstrapMeanPublishing, file="sandbox/Bootstrap_summary_for_publication.csv", row.names = FA
Calculate lambda based on t0 = (1-p)/(p \text{ Lambda}) So, 1/\text{lambda} = t0 * p / (1-p)
p = 0.7
BootstrapMean$One.over.lambdaP07 = BootstrapMean$t0 * p/ (1-p)
BootstrapMean[, c("t0", "One.over.lambdaP07")]
##
            t0 One.over.lambdaP07
## 1
      35.99604
                          83.99076
                          94.45003
## 2
      40.47858
      40.80201
                          95.20470
## 3
## 4
      55.06556
                         128.48632
## 5
      42.35209
                          98.82154
                         108.07053
## 6
      46.31594
## 7
      34.32367
                          80.08855
## 8
      31.44145
                          73.36337
      73.99028
                         172.64399
## 10 30.65496
                          71.52823
## 11 55.44863
                         129.38014
## 12 56.77646
                         132.47841
## 13 58.03551
                         135.41619
## 14 41.90105
                          97.76911
```

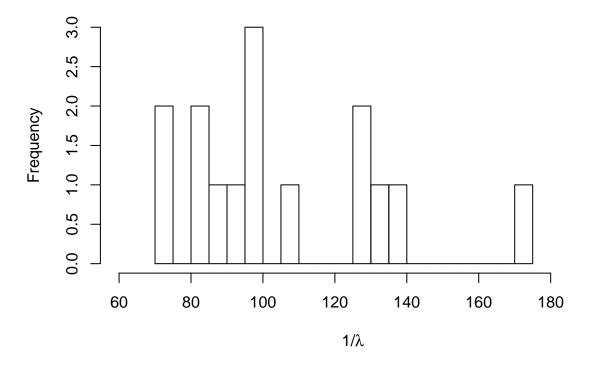
Histogram of 1/lambda for p=0.7

86.93857

15 37.25939

hist(BootstrapMean\$One.over.lambdaP07, br=20, xlim=c(60,190), xlab=expression(paste("1/",lambda)), main

Histogram of $1/\lambda$ when p=0.7



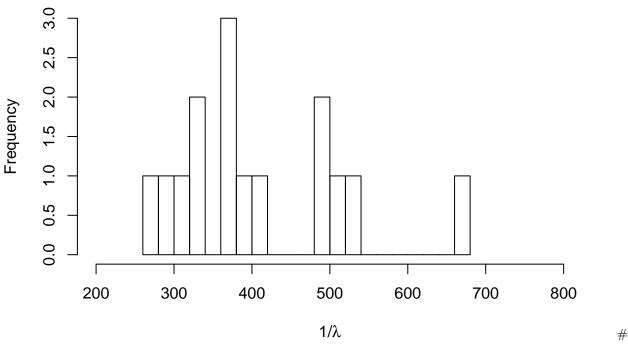
```
p = 0.9
BootstrapMean$One.over.lambdaP09 = BootstrapMean$t0 * p/ (1-p)
BootstrapMean[, c("t0", "One.over.lambdaP09")]
```

```
##
            t0 One.over.lambdaP09
## 1
      35.99604
                          323.9643
## 2
      40.47858
                          364.3073
## 3
      40.80201
                          367.2181
      55.06556
                          495.5901
      42.35209
                          381.1688
## 5
## 6
      46.31594
                          416.8435
## 7
      34.32367
                          308.9130
## 8
      31.44145
                          282.9730
      73.99028
                          665.9125
## 9
## 10 30.65496
                          275.8946
## 11 55.44863
                          499.0377
## 12 56.77646
                          510.9881
## 13 58.03551
                          522.3196
## 14 41.90105
                          377.1094
## 15 37.25939
                          335.3345
```

Histogram 1/lambda, p=0.9

hist(BootstrapMean\$One.over.lambdaP09, br=20, xlim=c(200,800), xlab=expression(paste("1/",lambda)), ma

Histogram of $1/\lambda$ when p=0.9



Strehler-Mildvan correlation in using bootstrap results

```
# BootstrapMean
summary(lm(log10(BootstrapMean$GompRFlex) ~ BootstrapMean$GompGFlex))
```

##

Call:

```
## lm(formula = log10(BootstrapMean$GompRFlex) ~ BootstrapMean$GompGFlex)
##
## Residuals:
##
                1Q Median
                                 3Q
       Min
                                         Max
## -0.49942 -0.20689 -0.02195 0.20883 0.36533
##
## Coefficients:
                         Estimate Std. Error t value Pr(>|t|)
##
## (Intercept)
                          -1.7110
                                  0.3204 -5.340 0.000134 ***
                                     2.6337 -3.191 0.007091 **
## BootstrapMean$GompGFlex -8.4039
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.2766 on 13 degrees of freedom
## Multiple R-squared: 0.4392, Adjusted R-squared: 0.3961
## F-statistic: 10.18 on 1 and 13 DF, p-value: 0.007091
summary(lm(log10(BootstrapMean$R) ~ BootstrapMean$GompGFlex))
##
## Call:
## lm(formula = log10(BootstrapMean$R) ~ BootstrapMean$GompGFlex)
## Residuals:
                        Median
        Min
                  1Q
                                     30
                                              Max
## -0.177513 -0.079147 0.002811 0.054411 0.234284
##
## Coefficients:
                         Estimate Std. Error t value Pr(>|t|)
##
## (Intercept)
                          -2.1366
                                    0.1251 -17.080 2.75e-10 ***
## BootstrapMean$GompGFlex -3.1043
                                     1.0283 -3.019 0.00987 **
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.108 on 13 degrees of freedom
## Multiple R-squared: 0.4121, Adjusted R-squared: 0.3669
## F-statistic: 9.114 on 1 and 13 DF, p-value: 0.009871
##
## Call:
## lm(formula = log10(BootstrapMean$R) ~ BootstrapMean$G)
##
## Residuals:
##
        Min
                  1Q
                        Median
                                     3Q
## -0.188886 -0.077273 -0.003401 0.059639 0.289279
## Coefficients:
                 Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                  -2.2745 0.1718 -13.238 6.38e-09 ***
## BootstrapMean$G -1.6447
                             1.2030 -1.367
                                              0.195
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
```

```
## Residual standard error: 0.1317 on 13 degrees of freedom
## Multiple R-squared: 0.1257, Adjusted R-squared: 0.05846
## F-statistic: 1.869 on 1 and 13 DF, p-value: 0.1947
summary(lm(log10(BootstrapMean$GompGFlex) ~ BootstrapMean$t0)) #G from t0 and n
##
## Call:
## lm(formula = log10(BootstrapMean$GompGFlex) ~ BootstrapMean$t0)
## Residuals:
##
        Min
                   1Q
                         Median
                                       30
## -0.016469 -0.012257 -0.005318  0.008555  0.024096
##
## Coefficients:
##
                     Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                   -0.5370301 0.0150244 -35.74 2.29e-14 ***
## BootstrapMean$t0 -0.0088343 0.0003205 -27.56 6.43e-13 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.01454 on 13 degrees of freedom
## Multiple R-squared: 0.9832, Adjusted R-squared: 0.9819
## F-statistic: 759.7 on 1 and 13 DF, p-value: 6.43e-13
summary(lm(log10(BootstrapMean$R) ~ BootstrapMean$n)) #G from t0 and n
##
## Call:
## lm(formula = log10(BootstrapMean$R) ~ BootstrapMean$n)
## Residuals:
        Min
                         Median
                   10
                                       30
## -0.132065 -0.057526  0.000259  0.043924  0.169621
##
## Coefficients:
##
                  Estimate Std. Error t value Pr(>|t|)
                  -3.68903
                              0.24459 -15.083 1.29e-09 ***
## (Intercept)
## BootstrapMean$n 0.16924
                              0.03482
                                       4.861 0.000311 ***
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.0839 on 13 degrees of freedom
## Multiple R-squared: 0.6451, Adjusted R-squared: 0.6178
## F-statistic: 23.63 on 1 and 13 DF, p-value: 0.0003109
Does to mediate GFlex \sim \log(RFlex)? Yes.
```

```
##
## Call:
## lm(formula = BootstrapMean$GompGFlex ~ BootstrapMean$t0)
##
```

summary(lm(BootstrapMean\$GompGFlex ~ BootstrapMean\$t0))

```
## Residuals:
                         Median
##
        Min
                   10
                                       30
                                                Max
## -0.007271 -0.004972 -0.002378 0.002244 0.013517
## Coefficients:
                     Estimate Std. Error t value Pr(>|t|)
##
                    0.2202732 0.0075774
                                          29.07 3.25e-13 ***
## (Intercept)
## BootstrapMean$t0 -0.0022402 0.0001616 -13.86 3.65e-09 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.007334 on 13 degrees of freedom
## Multiple R-squared: 0.9366, Adjusted R-squared: 0.9317
## F-statistic: 192.1 on 1 and 13 DF, p-value: 3.65e-09
summary(lm(log10(BootstrapMean$GompRFlex) ~ BootstrapMean$t0))
##
## Call:
## lm(formula = log10(BootstrapMean$GompRFlex) ~ BootstrapMean$t0)
## Residuals:
##
       Min
                 10
                      Median
                                   30
## -0.46908 -0.19237 -0.00486 0.23982 0.39995
## Coefficients:
                    Estimate Std. Error t value Pr(>|t|)
##
## (Intercept)
                   -3.533595
                               0.299440 -11.801 2.55e-08 ***
## BootstrapMean$t0 0.018197
                               0.006388
                                          2.849
                                                  0.0137 *
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.2898 on 13 degrees of freedom
## Multiple R-squared: 0.3843, Adjusted R-squared: 0.337
## F-statistic: 8.115 on 1 and 13 DF, p-value: 0.01369
summary(lm(log10(BootstrapMean$GompRFlex) ~ BootstrapMean$GompGFlex ))
##
## lm(formula = log10(BootstrapMean$GompRFlex) ~ BootstrapMean$GompGFlex)
##
## Residuals:
##
       Min
                 1Q
                      Median
                                   30
                                           Max
## -0.49942 -0.20689 -0.02195 0.20883 0.36533
## Coefficients:
                          Estimate Std. Error t value Pr(>|t|)
                                       0.3204 -5.340 0.000134 ***
## (Intercept)
                           -1.7110
## BootstrapMean$GompGFlex -8.4039
                                       2.6337 -3.191 0.007091 **
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.2766 on 13 degrees of freedom
## Multiple R-squared: 0.4392, Adjusted R-squared: 0.3961
```

```
## F-statistic: 10.18 on 1 and 13 DF, p-value: 0.007091
summary(lm(log10(BootstrapMean$GompGFlex) ~ BootstrapMean$t0 + log10(BootstrapMean$GompRFlex)))
##
## Call:
## lm(formula = log10(BootstrapMean$GompGFlex) ~ BootstrapMean$t0 +
##
       log10(BootstrapMean$GompRFlex))
##
## Residuals:
##
        Min
                   1Q
                          Median
## -0.017722 -0.010619 -0.005119 0.010194 0.021283
## Coefficients:
                                    Estimate Std. Error t value Pr(>|t|)
##
## (Intercept)
                                  -0.5734526  0.0523756  -10.949  1.33e-07 ***
## BootstrapMean$t0
                                  -0.0086468  0.0004161  -20.781  8.91e-11 ***
## log10(BootstrapMean$GompRFlex) -0.0103075 0.0141753 -0.727
                                                                  0.481
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.01481 on 12 degrees of freedom
## Multiple R-squared: 0.9839, Adjusted R-squared: 0.9812
## F-statistic: 366.4 on 2 and 12 DF, p-value: 1.75e-11
summary(lm(log10(BootstrapMean$GompRFlex) ~ BootstrapMean$t0 + log10(BootstrapMean$GompGFlex))) #not si
##
## Call:
## lm(formula = log10(BootstrapMean$GompRFlex) ~ BootstrapMean$t0 +
       log10(BootstrapMean$GompGFlex))
##
## Residuals:
                 1Q Median
##
       Min
                                    3Q
                                            Max
## -0.50584 -0.21093 -0.01843 0.20102 0.40655
##
## Coefficients:
##
                                  Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                                  -5.73236
                                             3.03919 -1.886
                                                                0.0837
## BootstrapMean$t0
                                  -0.01797
                                              0.05017
                                                       -0.358
                                                                0.7264
## log10(BootstrapMean$GompGFlex) -4.09431
                                             5.63067
                                                      -0.727
                                                                0.4811
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.2952 on 12 degrees of freedom
## Multiple R-squared: 0.4103, Adjusted R-squared: 0.312
## F-statistic: 4.175 on 2 and 12 DF, p-value: 0.04205
```