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%TREATMENTEFFECTS Test whether the population distributions are normal then %calculates differential stats for comparing across tretament groups and %produces the figures to show differences in distributions.

Initial configuration

```
close all
clear all
path(path,'./support_scripts/')
%set filename and path to the file on your computer
[metaboliteFileName, otuFileName] = fileNameCheck('results2.txt', 'otu_table3.txt');
%separate variables
[mconditionStr, metaboliteName, metabolite] = separateMetaboliteVars(metaboliteFileName);
%add insulin:gluose ratio
for i=1:59
    metabolite(27,i)=metabolite(18,i)/metabolite(1,i);
%correct metabolite labels
metaboliteName{1} = 'glucose';
metaboliteName{18} = 'insulin';
metaboliteName{27}='insulin:glucose';
%compute averages and errors
[norm, maverages, mstderrors, mcategory] = metaboliteBasicstats(metabolite,mconditionStr);
%compute t-test p value between lean and high dose gram negative
% this could aslo be done with log values, however, the result should
% essentially be the same if not less significant
[hvalue, pvalue]=ttest2(mcategory\{1,2\}\{27,2\},mcategory\{4,2\}\{27,2\});
```

Plots to help see relation between glucose and insulin and ins:glu ratio

```
weight_loss_figure;
% figure
% groupcategory=[ones(10,1) ; ones(10,1)+1 ; ones(9,1)+2 ; ones(10,1)+3 ; ones(10,1)+4 ; one
```

```
% xax=metabolite(1,:)';
% yax=metabolite(18,:)';
% gscatter(xax,yax,groupcategory,'bgrcmk','*',15);
% figure
% xax=1:59
% plotyy(xax,metabolite(1,:),xax,metabolite(18,:));
% plotyy(xax,metabolite(18,:),xax,metabolite(27,:));
%n x p
\% set up p variable for delta weight and total GLP
weight = [weightchange(:,1); weightchange(:,2); weightchange(1:9,3); weightchange(:,6); weightchange(:,6);
aglp=metabolite(25,:);
tglp=metabolite(26,:);
X=[ones(length(weight),1) weight metabolite(1:26,:)'];
y=metabolite(27,:)';
metaboliteName1{1}='intersect';
for i=1:length(metaboliteName)
        metaboliteName1{i+1}=metaboliteName{i};
end
for i=1:length(y)
    if (isnan(y(i)))
    else
        y1(j)=y(i);
        X1(j,:)=X(i,:);
        j=j+1;
    end
end
% model insulin resistance
mdl2=stepwiselm(X1,y1,'PEnter',0.06,'ResponseVar','insulin resistence','PredictorVars',metal
%[b,se,pval,stats]=stepwiselm(X1,y1, 'display', 'on');
ans =
 Columns 1 through 7
                       40.3200
                                39.9900 40.5300 40.9700 40.8900
   26.8200
            40.6700
  Columns 8 through 9
   40.1600 40.4800
```

```
ans =
 Columns 1 through 7
  28.0400 41.4300 39.6333 39.3800 40.3222 36.7100 39.0333
 Columns 8 through 9
  38.9600 35.9000
ans =
 Columns 1 through 7
   3.2800 15.5900 15.3000 15.2100 15.4500 15.5400 15.6200
 Columns 8 through 9
  15.2700 15.2600
ans =
 Columns 1 through 7
   3.7400 15.9900 14.1444 13.9400 14.6778 11.6400 13.9667
 Columns 8 through 9
  14.3800 11.2100
mmn =
  -1
mmn =
  -1
BW p value:L
```

p =

```
0.0012
baseline corrected BW p value:L
p =
   0.9405
fat p value:L
p =
  1.4563e-04
baseline corrected fat p value:L
p =
  0.4344
BW p value:0
p =
   0.1204
baseline corrected BW p value:0
p =
   0.3466
fat p value:0
p =
   0.4193
baseline corrected fat p value:0
p =
   1.0000
```

BW p value:A_{500}

```
p =
   0.3122
baseline corrected BW p value:A_{500}
p =
   0.0150
fat p value:A_{500}
p =
   0.0806
baseline corrected fat p value:A_{500}
p =
   0.0226
BW p value:A_{150}
p =
   0.0989
baseline corrected BW p value:A_{150}
p =
  4.0022e-04
fat p value:A_{150}
p =
   0.0011
baseline corrected fat p value:A_{150}
p =
  1.5989e-04
```

```
BW p value:A_{50}
p =
   0.9633
baseline corrected BW p value:A_{50}
p =
   0.0361
fat p value:A_{50}
p =
   0.0957
baseline corrected fat p value:A_{50}
p =
   0.0151
BW p value:B_{500}
p =
  1.7575e-08
baseline corrected BW p value:B_{500}
p =
  1.9618e-09
fat p value:B_{500}
p =
  1.2767e-07
baseline corrected fat p value:B_{500}
p =
```

```
5.4573e-08
BW p value:B_{150}
p =
   0.0083
baseline corrected BW p value:B_{150}
p =
  3.6601e-04
fat p value:B_{150}
p =
   0.0023
baseline corrected fat p value:B_{150}
p =
  5.5893e-04
BW p value:B_{50}
p =
   0.1210
baseline corrected BW p value:B_{50}
p =
   0.0076
fat p value:B_{50}
p =
   0.2118
baseline corrected fat p value:B_{50}
```

```
p =
   0.0832
BW p value:D_{10%}
p =
  1.5054e-06
baseline corrected BW p value:D_{10%}
p =
   2.0597e-07
fat p value:D_{10%}
p =
  3.6407e-07
baseline corrected fat p value:D_{10%}
p =
  1.6156e-07
unpaired t-test for delta weight against obese control:A_{500}
h =
     0
p =
   0.0726
unpaired t-test for delta weight against obese control:A_{150}
h =
     1
```

```
p =
   0.0234
unpaired t-test for delta weight against obese control:A_{50}
h =
    0
p =
   0.2683
unpaired t-test for delta weight against obese control:B_{500}
h =
   1
p =
  8.1375e-09
unpaired t-test for delta weight against obese control:B_{150}
h =
   1
p =
   0.0016
unpaired t-test for delta weight against obese control:B_{50}
h =
   1
p =
```

```
0.0295
unpaired t-test for delta weight against obese control:D_{10%}
h =
     1
p =
   5.9830e-08
1. Adding Amylin, FStat = 406.3868, pValue = 1.807879e-24
2. Adding NEFA mEq/L, FStat = 160.5211, pValue = 1.914427e-16
3. Adding NEFA mEq/L:Amylin, FStat = 195.1909, pValue = 8.777328e-18
Generate stats and graphs between lean and obese controls
%graphs and stats for glucose (1) insulin (18) active GLP-1 (25) total
%GLP-1 (26) and insulin:glucose ratio (27). Matrix indexes in brackets.
variableIndex = [1 18 25 26 27];
logswitch = [0 0 0 0 0];
%treatment groups: Lean control (1) obese control (2) gram positive
%antibiotic (3) high dose gram negative anitbiotic (4) low dose gram
%negative antibiotic (5) olligofructosccharide supplement (6)
%groups = [1 2 3 4 5 6];
groups = [2 3 4 5 6];
%plot graphs horizontally or vertically
horizontal=0;
%normalisation tests
[h1, hs1] = normalisationTest(variableIndex, mcategory, norm, metaboliteName, logswitch, gro
%box plots and stats
[pvalues, string_answers, h2, hs2] = generateBoxPlotsAndAnovaPValue(variableIndex, mcategory
h1 =
```

3

'glucose Lean 0.87055'

```
'glucose Vehicle 0.61212'
```

^{&#}x27;glucose Vancomycin 0.79858'

^{&#}x27;glucose Ceftazadine_low 0.97961'

^{&#}x27;glucose Ceftazadine_high 0.87915'

^{&#}x27;insulin Lean 0.33031'

^{&#}x27;insulin Vehicle 0.90642'

^{&#}x27;insulin Vancomycin 0.61477'

^{&#}x27;insulin Ceftazadine_low 0.9891'

^{&#}x27;insulin Ceftazadine_high 0.88017'

^{&#}x27;active GLP-1 Lean 0.98174'

^{&#}x27;active GLP-1 Vehicle 0.88755'

^{&#}x27;active GLP-1 Vancomycin 0.6058'

^{&#}x27;active GLP-1 Ceftazadine_low 0.57942'

^{&#}x27;active GLP-1 Ceftazadine_high 0.73583'

^{&#}x27;total GLP-1 Lean 0.89778'

^{&#}x27;total GLP-1 Vehicle 0.40209'

^{&#}x27;total GLP-1 Vancomycin 0.95099'

^{&#}x27;total GLP-1 Ceftazadine_low 0.57479'

^{&#}x27;total GLP-1 Ceftazadine_high 0.5291'

^{&#}x27;insulin:glucose Lean 0.28108'

^{&#}x27;insulin:glucose Vehicle 0.90987'

^{&#}x27;insulin:glucose Vancomycin 0.27855'

^{&#}x27;insulin:glucose Ceftazadine_low 0.82367'

```
'insulin:glucose Ceftazadine_high 0.99903'
h2 =
   4
hs2 =
  1.0410e+03
   'glucose 0.0019484'
hs2 =
  1.0900e+03
   'insulin 0.038103'
hs2 =
  1.1390e+03
   'active GLP-1 1.0051e-08'
hs2 =
  1.1880e+03
   'total GLP-1 1.1242e-11'
hs2 =
  1.2370e+03
    'insulin:glucose 0.1594'
```

Final formatting and save to pdf file

```
figure(h1)
% axesHandles = get(gcf,'children');
% set(axesHandles, 'fontsize', 5)
% for i=1:length(axesHandles)
     title = get(axesHandles(i), 'title');
      set(title, 'fontsize', 7)
%
      ylabel(axesHandles(i),'Probability')
% end
%figuresize(15, 10, 'centimeters') %updated script using Matt's magic number!
saveas(gcf, 'pdf_figures/tretamenteffects_diabetic_markers_normalisation_test', 'pdf')
figure(h2)
figuresize(15, 10, 'centimeters')
axesHandles = get(gcf,'children');
%set(h2, 'fontsize', 7)
for i=1:length(axesHandles)
    title = get(axesHandles(i), 'title');
     set(title, 'fontsize', 8)
%
       ylabel(axesHandles(i),'Probability')
 end
%set(axesHandles,'fontsize', 10)
saveas(gcf, 'pdf_figures/treatment_effects_diabetic_markers', 'pdf')
mmn =
    -1
```