# Grouped one dimensional data comparison (violinpoint version 0.3.0)

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#### Abstract

This is a comparison of various methods for visualizing groups of 1-dimensional data with an emphasis on the **violinpoint** package.

Keywords: visualization, display, one dimensional, grouped, groups, violin, scatter, points, quasirandom, beeswarm, van der Corput, beanplot.

#### 1. Methods

There are several ways to plot grouped one-dimensional data combining points and density estimation:

- **pseudorandom** The kernel density is estimated then points are distributed uniform randomly within the density estimate for a given bin. Selection of an appropriate number of bins does not greatly affect appearance but coincidental clumpiness is common.
- alternating within bins The kernel density is estimated then points are distributed within the density estimate for a given bin evenly spaced with extreme values alternating from right to left e.g. max, 3rd max, ..., 4th max, 2nd max. If maximums are placed on the outside then these plots often form consecutive "smiley" patterns. If minimums are placed on the outside then "frowny" patterns are generated. Selection of the number of bins can have large effects on appearance important.
- beanplot The package beeswarm provides methods for generating a "beeswarm" plot where points are distibuted so that no points overlap. Kernel density is not calculated although the resulting plot does provide an approximate density estimate. Selection of an appropriate number of bins affects appearance and plot and point sizes must be known in advance.
- **quasirandom** The kernel density is estimated then points are distributed quasirandomly using the von der Corput sequence within the density estimate for a given bin. Selection of an appropriate number of bins does not greatly affect appearance and position does not depend on plotting parameters.

#### 2. Simulated data

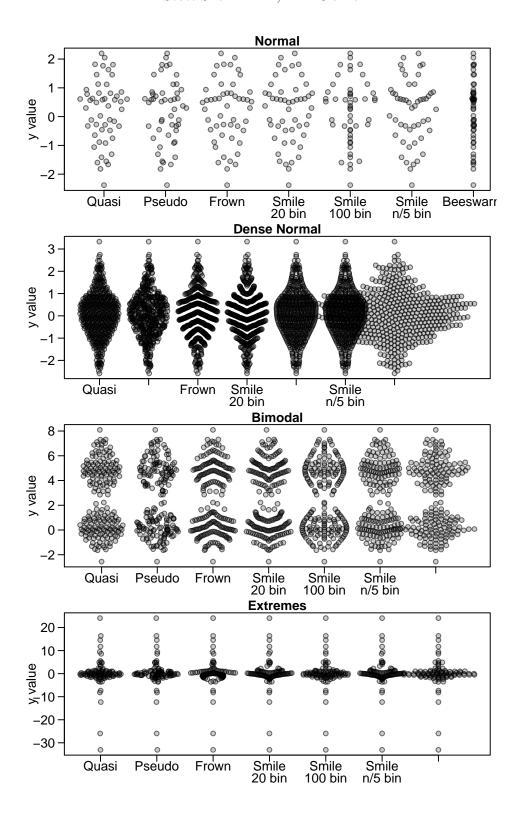
To compare between methods we'll generate some simulated data from normal, bimodal (two

normal) and Cauchy distributions:

```
> library(beeswarm)
> library(violinpoint)
> set.seed(12345)
> dat <- list(rnorm(50), rnorm(500), c(rnorm(100), rnorm(100,5)), rcauchy(100))
> names(dat) <- c("Normal", "Dense Normal", "Bimodal", "Extremes")</pre>
```

Here, we plot the data using quasirandom, pseudorandom, alternating and beeswarm methods:

```
par(mfrow=c(4,1), mar=c(2.5,3.1, 1.2, 0.5), mgp=c(2.1,.75,0),
      cex.axis=1.2,cex.lab=1.2,cex.main=1.2)
   dummy<-sapply(names(dat),function(label) {</pre>
      y<-dat[[label]]
      offsets <- list(
        'Quasi'=offsetX(y), # Default
        'Pseudo'=offsetX(y, method='pseudorandom',nbins=100),
        'Frown'=offsetX(y, method='frowney',nbins=20),
        'Smile\n20 bin'=offsetX(y, method='smiley',nbins=20),
        'Smile\n100 bin'=offsetX(y, method='smiley',nbins=100),
        'Smile\nn/5 bin'=offsetX(y, method='smiley',nbins=round(length(y)/5)),
        'Beeswarm'=swarmx(rep(0,length(y)),y)$x
      ids <- rep(1:length(offsets), each=length(y))</pre>
      plot(unlist(offsets) + ids, rep(y, length(offsets)),
          xlab='', xaxt='n', pch=21,las=1,main=label, ylab='y value',
          col='#00000099',bg='#00000033')
   par(lheight=.8)
    axis(1, 1:length(offsets), names(offsets), padj=1, mgp=c(0, -.3, 0), tcl=-.5)
+ })
```



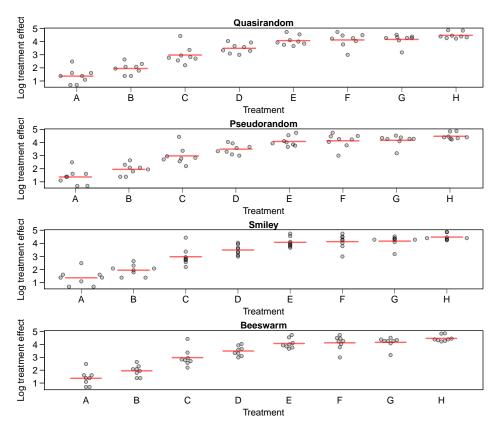
#### 3. Real data

#### 3.1. Few data points

An example with few data points (maybe a bit too few for optimal use of this package) using the OrchardSprays data from the datasets package:

```
par(mfrow=c(4,1), mar=c(3.5,3.1, 1.2, 0.5), mgp=c(2.1,.75,0),
      cex.axis=1.2,cex.lab=1.2,cex.main=1.2)
    #simple function to avoid repeating code
   plotFunc<-function(x,y,offsetXArgs){</pre>
      vpPlot(
      x, y,
       las=1, ylab='Log treatment effect',
      pch=21, col='#00000099',bg='#00000033',
       offsetXArgs=offsetXArgs
      title(xlab='Treatment')
      addMeanLines(x,y)
+
   addMeanLines<-function(x,y){
      means<-tapply(y,x,mean)</pre>
      segments(
       1:length(means)-.3,means,1:length(means)+.3,means,
       col='#FF000099',1wd=2
   }
   #quasirandom
   plotFunc(OrchardSprays$treatment,log(OrchardSprays$decrease),NULL)
   title(main='Quasirandom')
   #pseudorandom
   plotFunc(
      OrchardSprays$treatment,log(OrchardSprays$decrease),
      list(method='pseudo')
   )
   title(main='Pseudorandom')
   #smiley
   plotFunc(
      OrchardSprays$treatment,log(OrchardSprays$decrease),
      list(method='smiley')
   )
   title(main='Smiley')
   #beeswarm
   beeInput<-split(
      log(OrchardSprays$decrease),
     OrchardSprays$treatment
   )
```

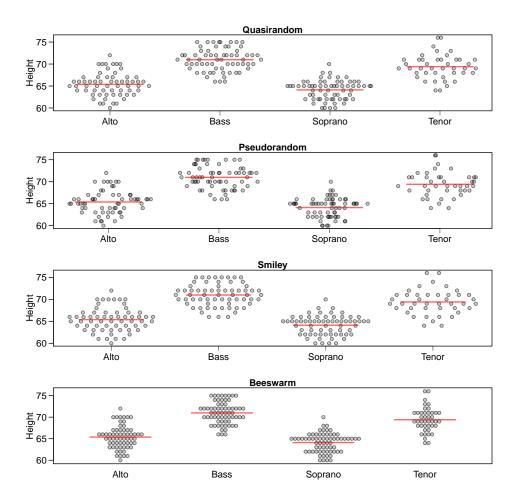
```
> beeswarm(
+ beeInput,las=1,
+ ylab='Log treatment effect',xlab='Treatment',
+ pch=21, col='#00000099',bg='#00000033',
+ main='Beeswarm'
+ )
> addMeanLines(
+ OrchardSprays$treatment,
+ log(OrchardSprays$decrease)
+ )
```



#### 3.2. Discrete data

Data with discrete bins are plotted adequately although other display choices (e.g. multiple barplots) might be better for final publication. For example the **singer** data from the **lattice** package has its data rounded to the nearest inch:

```
vpPlot(
      x, y,
       las=1, ylab='Height',
       pch=21, col='#00000099',bg='#00000033',
       offsetXArgs=offsetXArgs
      )
      addMeanLines(x,y)
+
    }
    #quasirandom
   plotFunc(parts, singer$height, NULL)
    title(main='Quasirandom')
    #pseudorandom
   plotFunc(
      parts, singer $ height,
      list(method='pseudo')
   title(main='Pseudorandom')
  #smiley
   plotFunc(
      parts, singer $ height,
      list(method='smiley')
    )
   title(main='Smiley')
    #beeswarm
    beeInput<-split(singer$height, parts)</pre>
    beeswarm(
      beeInput, las=1,
      ylab='Height',main='Beeswarm',
     pch=21, col='#00000099',bg='#00000033',
    addMeanLines(parts, singer$height)
```

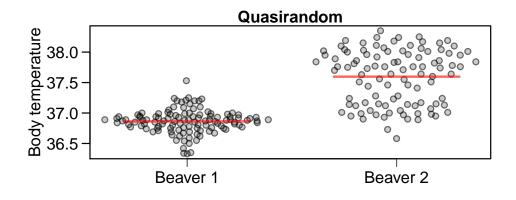


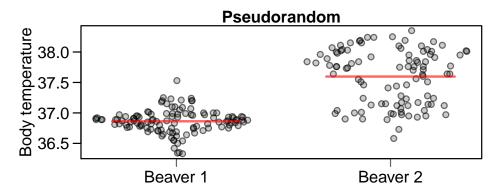
### 3.3. Moderately sized data

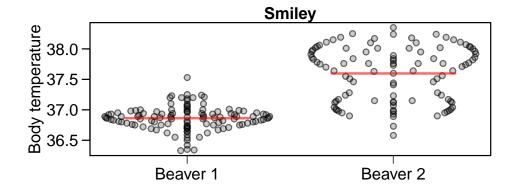
An example with using the beaver1 and beaver2 data from the datasets package:

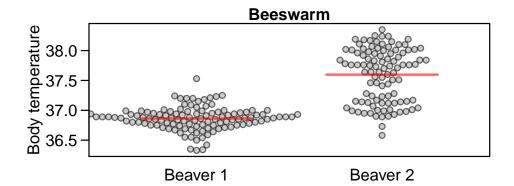
```
> y<-c(beaver1$temp,beaver2$temp)
> x<-rep(
+ c('Beaver 1','Beaver 2'),
+ c(nrow(beaver1),nrow(beaver2))
+ )
> par(mfrow=c(4,1), mar=c(3.5,4, 1.2, 0.5),mgp=c(3,.75,0),
+ cex.axis=1.2,cex.lab=1.2,cex.main=1.2)
> #simple function to avoid repeating code
> plotFunc<-function(x,y,offsetXArgs){
+ vpPlot(
+ x,y,
+ las=1, ylab='Body temperature',
+ pch=21, col='#00000099',bg='#00000033',
+ offsetXArgs=offsetXArgs
+ )</pre>
```

```
+
      addMeanLines(x,y)
+
   #quasirandom
   plotFunc(x,y,NULL)
   title(main='Quasirandom')
   #pseudorandom
   plotFunc(
     x, y,
      list(method='pseudo')
   )
   title(main='Pseudorandom')
   #smiley
  plotFunc(
     x, y,
     list(method='smiley')
  title(main='Smiley')
  #beeswarm
   beeInput<-split(y,x)</pre>
   beeswarm(
     beeInput,las=1,
     ylab='Body temperature',main='Beeswarm',
     pch=21, col='#00000099',bg='#00000033',
   addMeanLines(x,y)
```





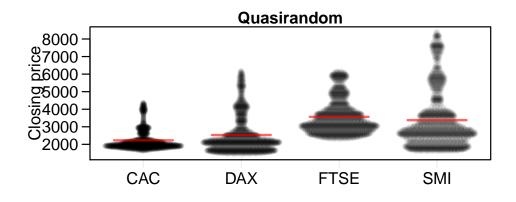


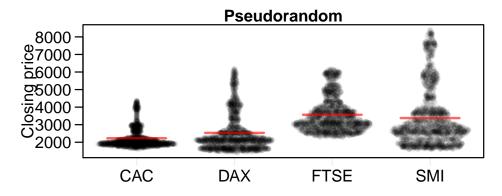


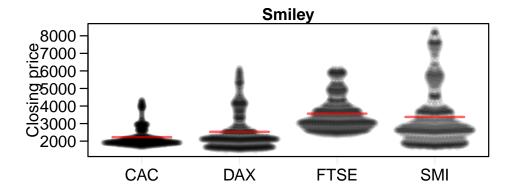
#### 3.4. Larger data

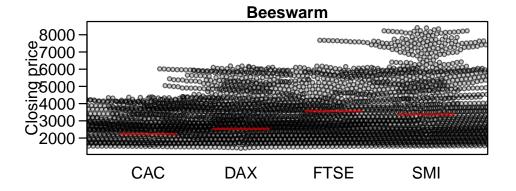
An example with using the EuStockMarkets data from the datasets package:

```
y<-as.vector(EuStockMarkets)</pre>
x<-rep(colnames(EuStockMarkets), each=nrow(EuStockMarkets))
par(mfrow=c(4,1), mar=c(3.5,4, 1.2, 0.5), mgp=c(3,.75,0),
  cex.axis=1.2,cex.lab=1.2,cex.main=1.2)
#simple function to avoid repeating code
plotFunc<-function(x,y,offsetXArgs){</pre>
  vpPlot(
   x, y,
   las=1, ylab='Closing price',cex=.7,
   pch=21, col='#00000011',bg='#00000011',
   offset XArgs = offset XArgs
  addMeanLines(x,y)
}
#quasirandom
plotFunc(x,y,NULL)
title(main='Quasirandom')
#pseudorandom
plotFunc(
  x, y,
  list(method='pseudo')
title(main='Pseudorandom')
#smiley
plotFunc(
  x, y,
  list(method='smiley')
title(main='Smiley')
#beeswarm
beeInput<-split(y,x)</pre>
beeswarm(
  beeInput,las=1,cex=.7,
  ylab='Closing price',main='Beeswarm',
  pch=21, col='#00000099',bg='#00000033',
addMeanLines(x,y)
```









## Affiliation:

Github: http://github.com/sherrillmix/violinpoint