GES673 Lab @ UMBC by Richard Heimann

Introduction

Data analysis is like an interrogation. That is, the interviewer hopes to use a series of questions in order to discover information - if not the truth. The questions the interrogator asks, of course, are subjectively chosen, at least initiall, but in time are selected based on question utility i.e. those questions that produce maximum yield. As such, the information that one interrogator gets out of an interrogatee might be fairly different from the information that another interviewer gets out of the same person. That is, the efficacy of one will be different than another based on experience. Exploratory /Spatial/ Data Analysis provides the data analyst the intuition to interrogate data to maximize information yield. This lab provides some efficient ways to gracefully handle datasets of unknown information yield. The hazard is not to torture your data as it will speak to you in unreliable ways. The lab is an R exercise, which hopefully adds more pragmatic and systematic description of the process. That said, the commands (and thus the analysis) below are not the only way of analyzing the data. When you understand what the commands are doing, you might decide to take a different approach to analyzing the data - please do so, and be sure to share what you find!

Dataset Background

The datasets, for this lab relate to council areas in Scotland (roughly equivalent to counties). The one which I have labeled 'main' has numbers representing the number of drug related deaths by council area, with most of its columns containing counts that relate to specific drugs. It also contains geographical coordinates of the council areas, in latitude and longitude. The one which I have labeled 'pop' contains population numbers. The rest of the datasets contain numbers relating to problems with crime, education, employment, health, and income. The datasets contain proportions in them, such that values closer to 1 indicate that the council area is more troubled, while values closer to 0 indicate that the council area is less troubled in that particular way.

```
# install.packages('dplyr', repos='http://cran.us.r-project.org')
library(dplyr)
```

```
##
## Attaching package: 'dplyr'
##
## The following objects are masked from 'package:stats':
##
##     filter, lag
##
## The following objects are masked from 'package:base':
##
##     intersect, setdiff, setequal, union
```

```
# install.packages('stats', repos='http://cran.us.r-project.org')
library(stats)
```

```
# Set working directory
setwd("/Users/heimannrichard/Google Drive/GIS Data/drugdata_scotland")
# Loading all the datasets
main <- read.csv("2012-drugs-related-cx.csv")
pop <- read.csv("scotland pop by ca.csv")
crime <-
read.csv("most_deprived_datazones_by_council_(crime)_2012.csv")
edu <-
read.csv("most_deprived_datazones_by_council_(education)_2012.csv")
emp <-
read.csv("most_deprived_datazones_by_council_(employment)_2012.csv")
health <-
read.csv("most_deprived_datazones_by_council_(health)_2012.csv")
income <-
read.csv("most_deprived_datazones_by_council_(income)_2012.csv")
```

```
# Explorattion by indexing the data
names(main)
```

```
##  [1] "Council.area"            "All.drug.related.deaths"
##  [3] "Heroin...morphine"       "Methadone"
##  [5] "Any.benzo.diazepine"     "Diazepam"
##  [7] "Temazepam"               "Cocaine"
##  [9] "Ecstasy"                 "Amphetamines"
## [11] "Alcohol"                 "Latitude"
## [13] "Longitude"
```

```
main$Council.area
```

```
##  [1] Aberdeen City          Aberdeenshire          Angus
##  [4] Argyll and Bute        Clackmannanshire       Dumfries and
Galloway
##  [7] Dundee City            East Ayrshire          East
Dunbartonshire
## [10] East Lothian           East Renfrewshire      City of Edinburgh
## [13] Eilean Siar            Falkirk                Fife
## [16] Glasgow City           Highland               Inverclyde
## [19] Midlothian             Moray                  North Ayrshire
## [22] North Lanarkshire      Orkney Islands         Perth and Kinross
## [25] Renfrewshire           Scottish Borders       Shetland Islands
## [28] South Ayrshire         South Lanarkshire      Stirling
## [31] West Dunbartonshire    West Lothian
## 32 Levels: Aberdeen City Aberdeenshire Angus ... West Lothian
```

```
main$Council.area[1:10]
```

```
##  [1] Aberdeen City          Aberdeenshire          Angus
##  [4] Argyll and Bute        Clackmannanshire       Dumfries and
Galloway
##  [7] Dundee City            East Ayrshire          East
Dunbartonshire
## [10] East Lothian
## 32 Levels: Aberdeen City Aberdeenshire Angus ... West Lothian
```

```
main[1:10, 1]
```

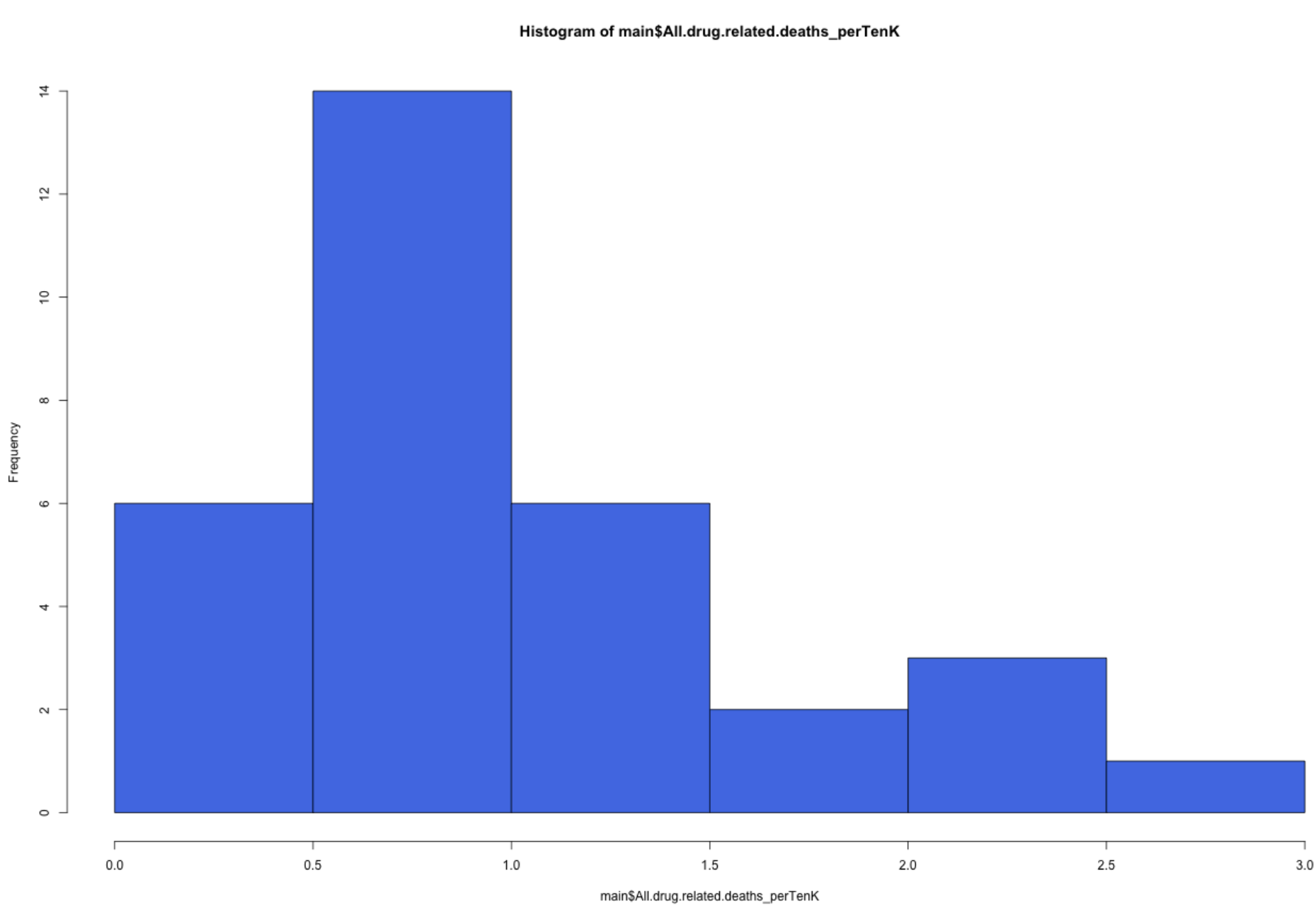
```
##  [1] Aberdeen City          Aberdeenshire          Angus
##  [4] Argyll and Bute        Clackmannanshire       Dumfries and
Galloway
##  [7] Dundee City            East Ayrshire          East
Dunbartonshire
## [10] East Lothian
## 32 Levels: Aberdeen City Aberdeenshire Angus ... West Lothian
```

```
# Merging other relevant data with the main dataset
main <- merge(main, pop[, c(2, 3)], by.x = "Council.area", by.y =
"Council.area",
    all.x = TRUE)
main <- merge(main, crime[, c(1, 4)], by.x = "Council.area", by.y =
"label",
    all.x = TRUE)
main <- merge(main, edu[, c(1, 4)], by.x = "Council.area", by.y =
"label", all.x = TRUE)
main <- merge(main, emp[, c(1, 4)], by.x = "Council.area", by.y =
"label", all.x = TRUE)
main <- merge(main, health[, c(1, 4)], by.x = "Council.area", by.y =
"label",
    all.x = TRUE)
main <- merge(main, income[, c(1, 4)], by.x = "Council.area", by.y =
"label",
    all.x = TRUE)
```

1. Provide a few sentences about drug related deaths? You can find this information by performing some desktop research. e.g. Google Search, Google Scholar, UMBC Library

```
# Weighting the number of drug related deaths by the population of
each
# council area new variable named All.drug.related.deaths_perTenK
(drug
# deaths / (population/10,000))
main$All.drug.related.deaths_perTenK <-
(main$All.drug.related.deaths/(main$Population/10000))
```

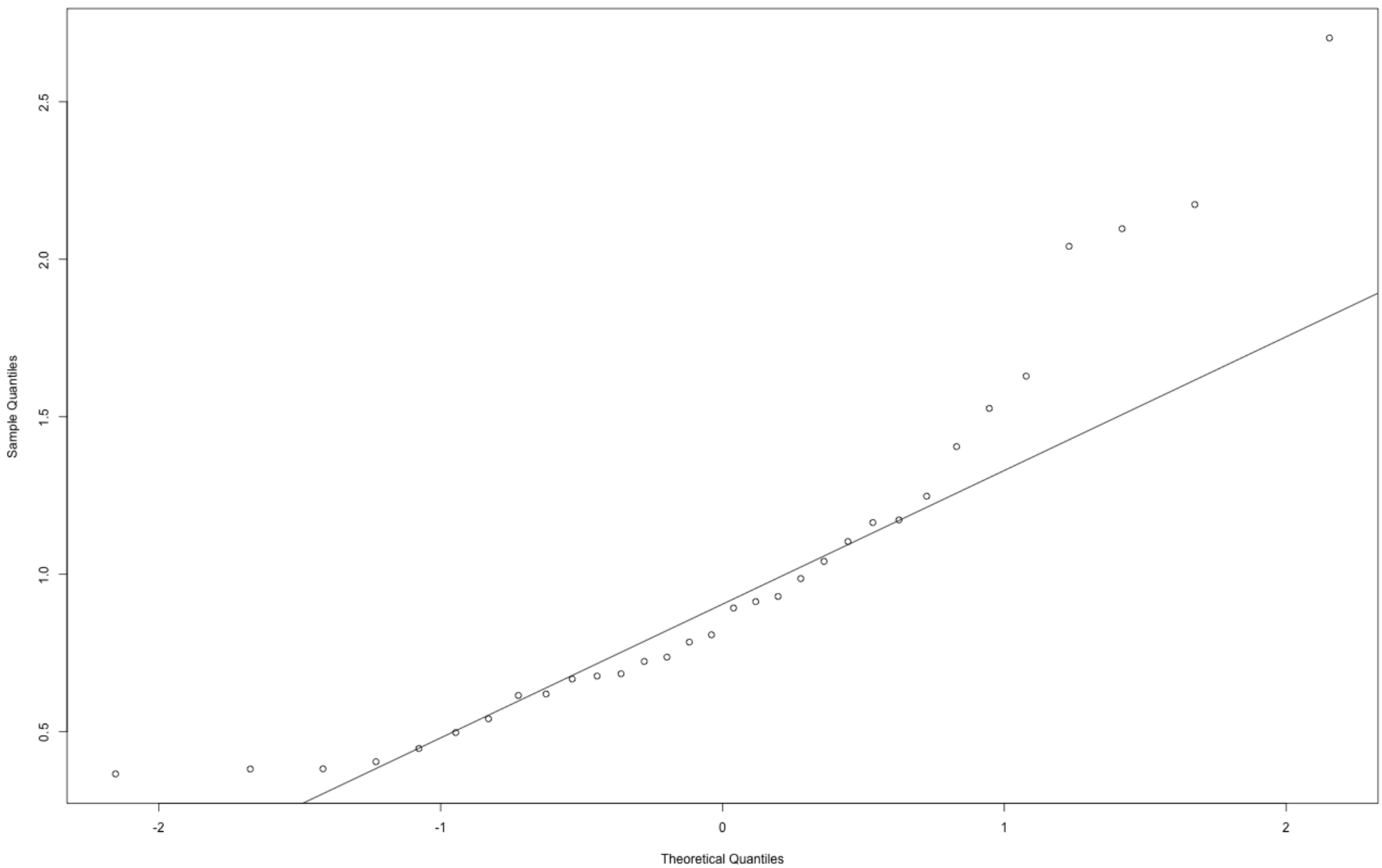
```
# A histogram of the number of drug related deaths per 10,000 people
hist(main$All.drug.related.deaths_perTenK, col = "royal blue")
```



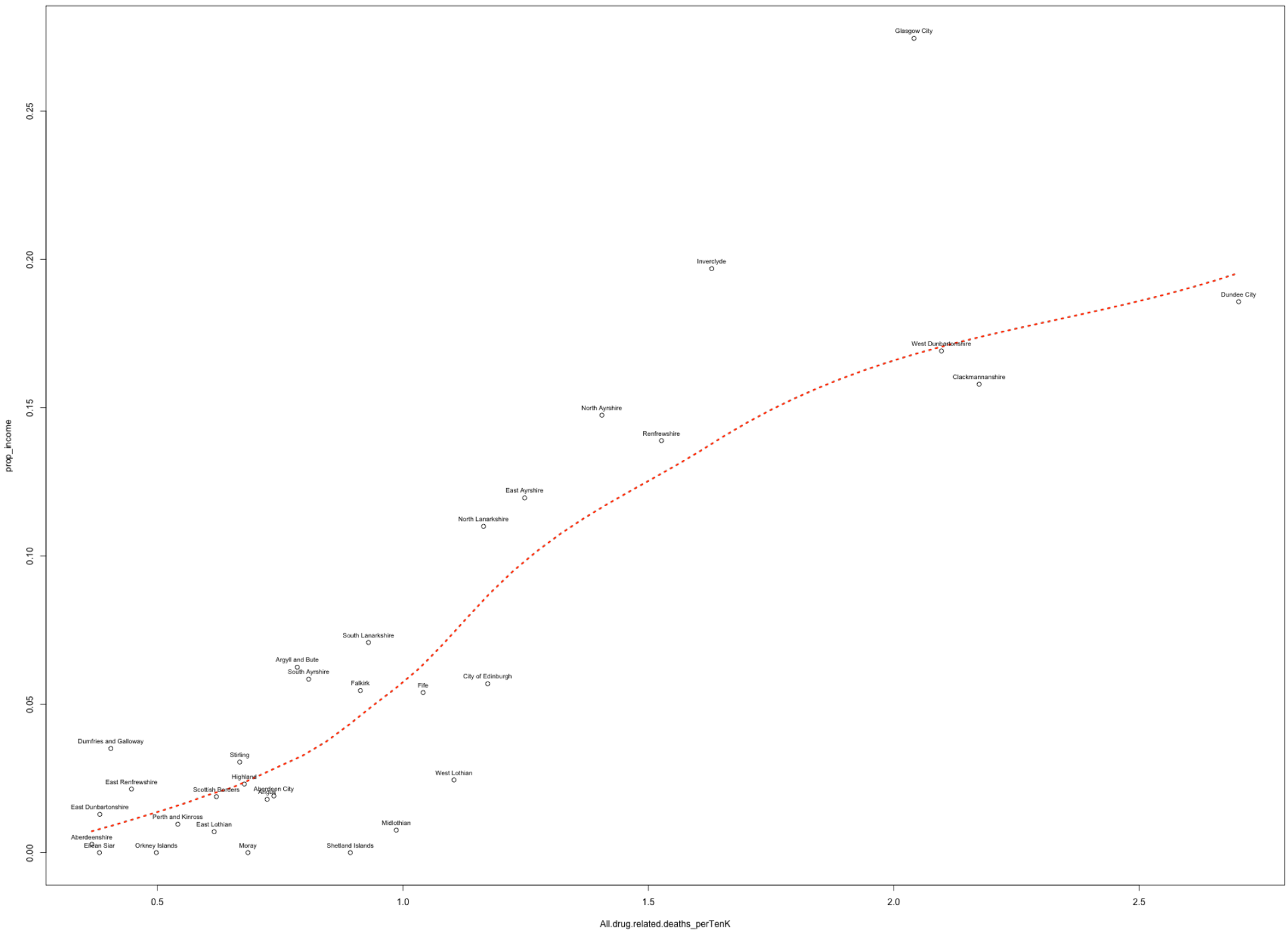
2. How is a histogram useful?

3. What emerges when examining the univariate histogram of drug related deaths / 10,000?

```
# Q-Q plots are another way to check for normality. As our histogram
# suggested we have a non-normal distrubution
qqnorm(main$All.drug.related.deaths_perTenK)
qqline(main$All.drug.related.deaths_perTenK)
```




```
# scatterplot (drug deaths ~ income)
with(main, scatter.smooth(All.drug.related.deaths_perTenK,
prop_income, lpars = list(col = "red",
    lwd = 3, lty = 3)))
text(main$All.drug.related.deaths_perTenK, main$prop_income, labels =
main$Council.area,
    cex = 0.7, pos = 3)
```



4a. What appears to be the largest bivariate outlier? Where is it located? e.g: North, South, East or West?

4b. Are there any other observations that demand further investigation? If so, which and why?

```
# Simple summary stats of one variable at a time
mean(main$All.drug.related.deaths)
```

```
## [1] 18.16
```

```
median(main$All.drug.related.deaths)
```

```
## [1] 10
```

```
mean(main$All.drug.related.deaths_perTenK)
```

```
## [1] 1.011
```

```
median(main$All.drug.related.deaths_perTenK)
```

```
## [1] 0.8504
```

```
# Summary stats of all the variables in the dataset
summary(main)
```

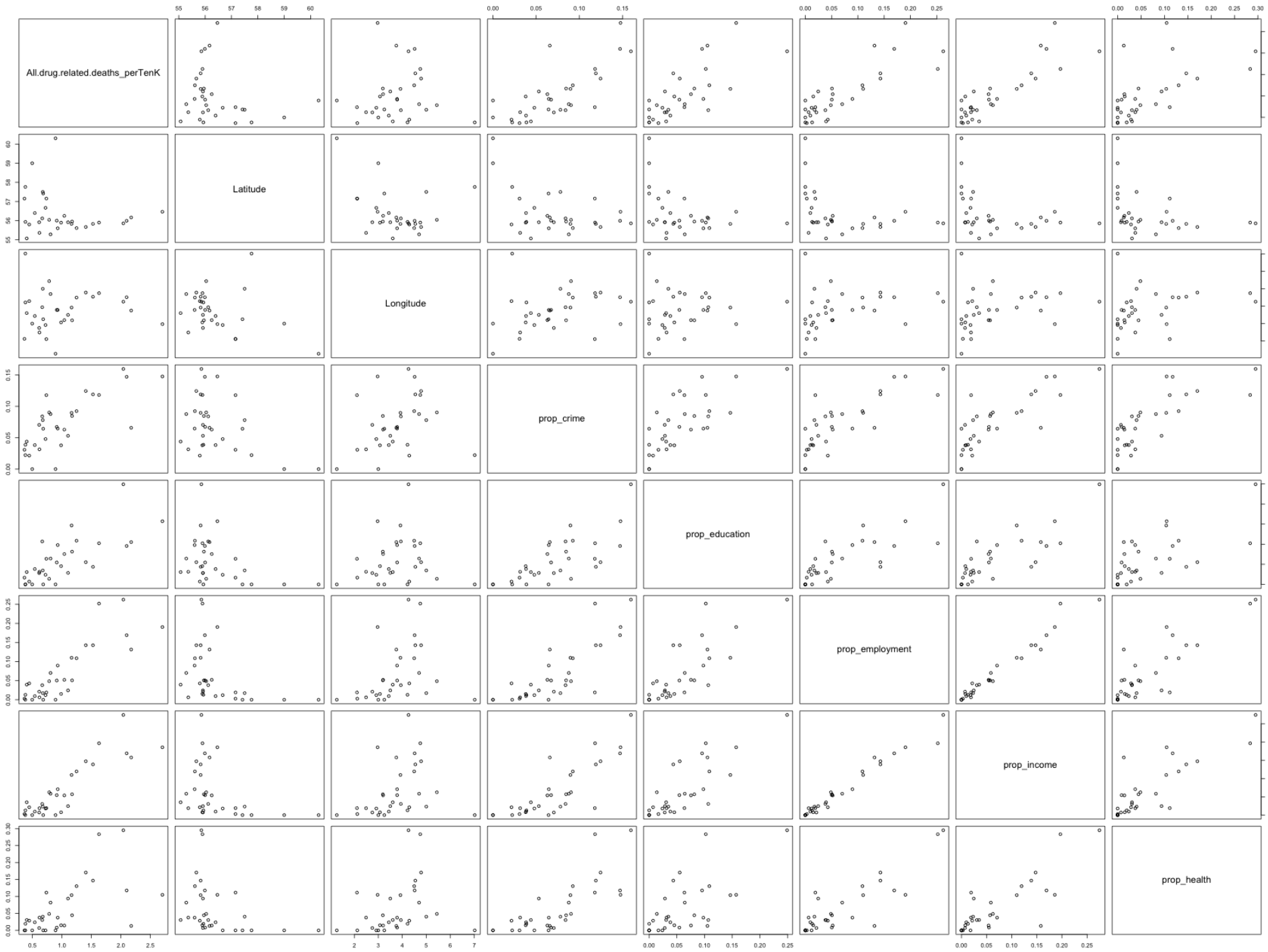
```
##              Council.area All.drug.related.deaths Heroin...morphine
##  Aberdeen City    : 1    Min.   :  1.0           Min.   : 0.00
##  Aberdeenshire    : 1    1st Qu.:  6.0           1st Qu.: 3.50
##  Angus            : 1    Median : 10.0           Median : 5.00
##  Argyll and Bute  : 1    Mean   : 18.2           Mean   : 7.59
##  City of Edinburgh: 1    3rd Qu.: 19.0           3rd Qu.: 8.50
##  Clackmannanshire : 1    Max.   :121.0           Max.   :57.00
##  (Other)          :26
##    Methadone     Any.benzo.diazepine    Diazepam       Temazepam
##  Min.   : 0.00   Min.   : 0.0        Min.   : 0.00   Min.   :0.000
##  1st Qu.: 2.00   1st Qu.: 5.0        1st Qu.: 4.75   1st Qu.:0.000
##  Median : 4.00   Median : 7.5        Median : 7.00   Median :0.000
##  Mean   : 7.91   Mean   :13.1        Mean   :12.06   Mean   :0.469
##  3rd Qu.:11.00   3rd Qu.:13.8        3rd Qu.:13.00   3rd Qu.:1.000
##  Max.   :52.00   Max.   :83.0        Max.   :74.00   Max.   :4.000
##
##     Cocaine        Ecstasy       Amphetamines      Alcohol
##  Min.   : 0.0   Min.   :0.000   Min.   :0.000   Min.   : 0.00
##  1st Qu.: 0.0   1st Qu.:0.000   1st Qu.:0.000   1st Qu.: 2.00
##  Median : 1.0   Median :0.000   Median :0.000   Median : 4.50
##  Mean   : 1.5   Mean   :0.281   Mean   :0.781   Mean   : 7.53
##  3rd Qu.: 2.0   3rd Qu.:1.000   3rd Qu.:1.000   3rd Qu.: 7.25
##  Max.   :14.0   Max.   :1.000   Max.   :6.000   Max.   :55.00
##
##     Latitude      Longitude      Population       prop_crime
##  Min.   :55.1   Min.   :1.27   Min.   : 20100   Min.   :0.0000
##  1st Qu.:55.8   1st Qu.:3.05   1st Qu.: 89425   1st Qu.:0.0386
##  Median :56.0   Median :3.77   Median :116550   Median :0.0665
```

```
##  Mean   :56.4   Mean   :3.75   Mean   :163194   Mean   :0.0719
##  3rd Qu.:56.5   3rd Qu.:4.50   3rd Qu.:183350   3rd Qu.:0.0908
##  Max.   :60.3   Max.   :7.02   Max.   :592800   Max.   :0.1596
##
##  prop_education   prop_employment   prop_health      prop_income
##  Min.   :0.0000   Min.   :0.0000   Min.   :0.0000   Min.   :0.0000
##  1st Qu.:0.0221   1st Qu.:0.0127   1st Qu.:0.0120   1st Qu.:0.0121
##  Median :0.0445   Median :0.0412   Median :0.0341   Median :0.0328
##  Mean   :0.0596   Mean   :0.0667   Mean   :0.0638   Mean   :0.0652
##  3rd Qu.:0.0962   3rd Qu.:0.1090   3rd Qu.:0.1041   3rd Qu.:0.1124
##  Max.   :0.2491   Max.   :0.2624   Max.   :0.2950   Max.   :0.2745
##
##  All.drug.related.deaths_perTenK
##  Min.   :0.366
##  1st Qu.:0.619
##  Median :0.850
##  Mean   :1.011
##  3rd Qu.:1.191
##  Max.   :2.703
##
```

5. What can you tell about the mean of drug related deaths?

6. What can you share about the summary of summary(main)?

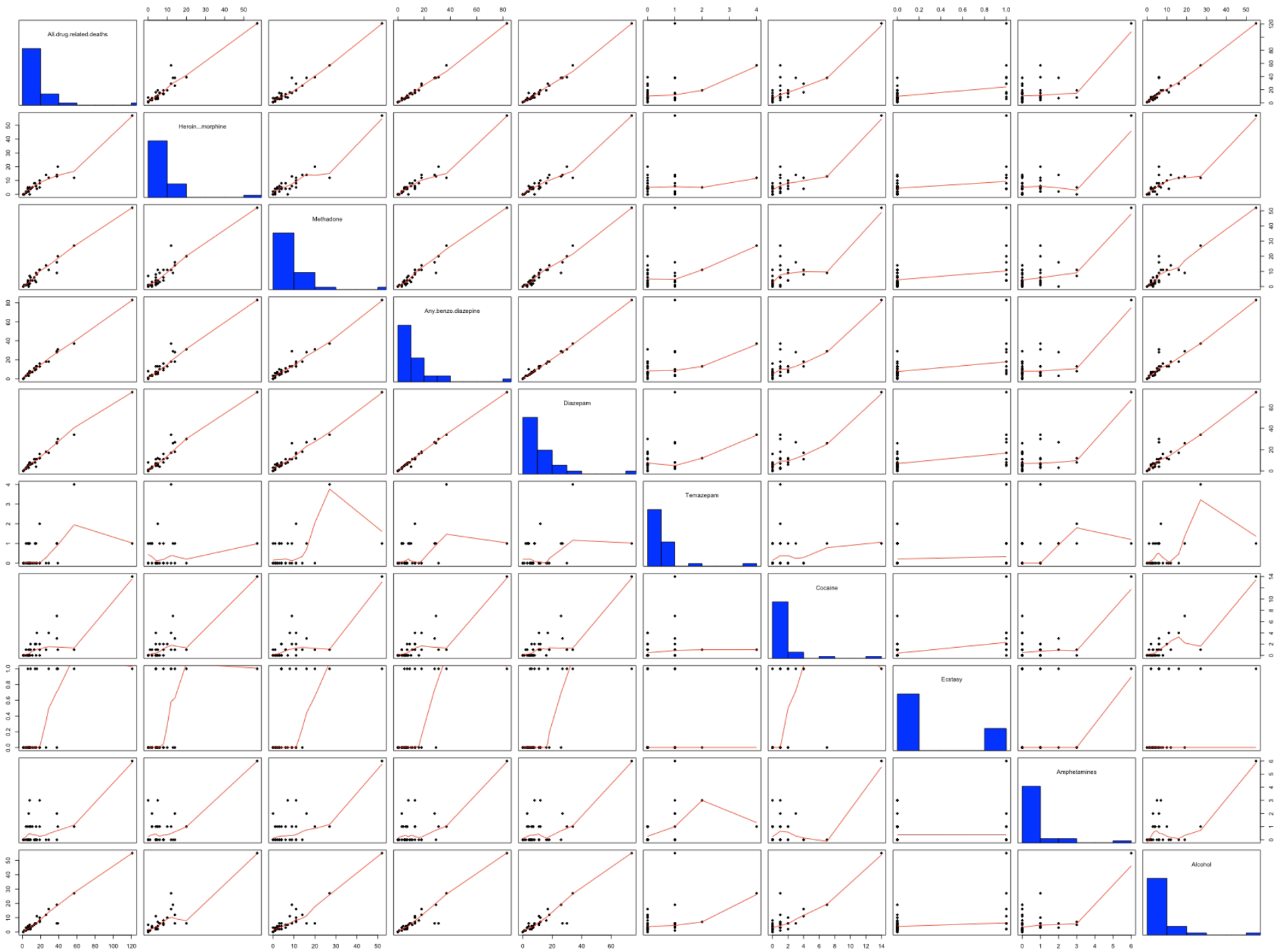
```
# A Scatterplot matrix
pairs(~All.drug.related.deaths_perTenK + Latitude + Longitude +
prop_crime +
    prop_education + prop_employment + prop_income + prop_health, data
= main)
```




```
# scatterplot matrix wiht some extra (note panel.hist function run in
# echo=FALSE mode))
colnames(main)
```

```
##  [1] "Council.area"                    "All.drug.related.deaths"
##  [3] "Heroin...morphine"               "Methadone"
##  [5] "Any.benzo.diazepine"             "Diazepam"
##  [7] "Temazepam"                       "Cocaine"
##  [9] "Ecstasy"                         "Amphetamines"
## [11] "Alcohol"                         "Latitude"
## [13] "Longitude"                       "Population"
## [15] "prop_crime"                      "prop_education"
## [17] "prop_employment"                 "prop_health"
## [19] "prop_income"
"All.drug.related.deaths_perTenK"
```

```
pairs(main[2:11], panel = panel.smooth, cex = 1, pch = 20, bg =
"blue", diag.panel = panel.hist,
    cex.labels = 1, font.labels = 1)
```



7. In what ways can a scatterplot matrix be useful? In what ways can it be used incorrectly or inefficeintly?

```
# We split our dataset into two regions for subsequent analysis. We
perform
# a median split of the longitudes of the council # areas resulting in
an
# 'east' and 'west' group.  ?cut: divides the range of x into
intervals and
# codes the values in x according to which interval they fall. The
leftmost
# # interval corresponds to level one, the next leftmost to level two
and so
# on.
main$LongSplit <- cut(main$Longitude, breaks =
quantile(main$Longitude, c(0,
    0.5, 1)), include.lowest = TRUE, right = FALSE, ordered_result =
TRUE, labels = c("East",
    "West"))
```

```
# Let's examine the number of records that result in each group:
table(main$LongSplit)
```

```
##
## East West
##   16   16
```

```
# We split our dataset into two regions for subsequent analysis. We
perform
# a median split of the longitudes of the council # areas resulting in
an
# 'north' and 'south' group.  ?cut: divides the range of x into
intervals
# and codes the values in x according to which interval they fall. The
# leftmost # interval corresponds to level one, the next leftmost to
level
# two and so on.
main$LatSplit <- cut(main$Latitude, breaks = quantile(main$Latitude,
c(0, 0.5,
    1)), include.lowest = TRUE, right = FALSE, ordered_result = TRUE,
labels = c("South",
    "North"))
```

```
# Let's examine the number of records that result in each group:
table(main$LatSplit)
```

```
##
## South North
##    16    16
```

```
data_source <- collect(main)
grouping_factors <- group_by(main, LongSplit, LatSplit)
deaths_by_area <- summarise(grouping_factors, median.deathsptk =
median(All.drug.related.deaths_perTenK),
    median.crime = median(prop_crime), median.emp =
median(prop_employment),
    median.edu = median(prop_education), num.council.areas =
length(All.drug.related.deaths_perTenK))
```

```
# Examine the summary table just created
grouping_factors
```

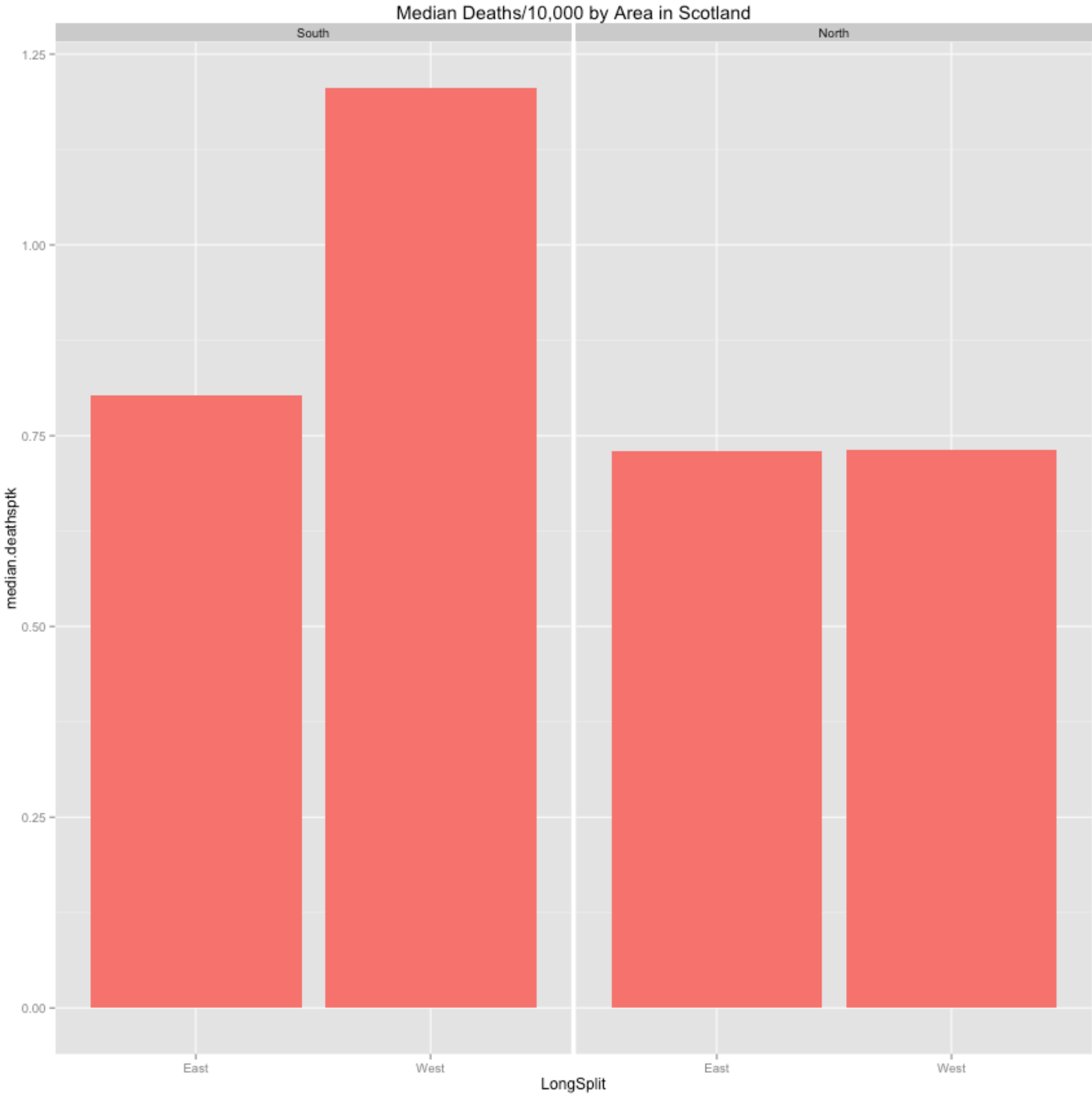
```
## Source: local data frame [32 x 22]
## Groups: LongSplit, LatSplit
##
##             Council.area All.drug.related.deaths Heroin...morphine
## 1          Aberdeen City                      16                 4
## 2          Aberdeenshire                       9                 4
## 3                 Angus                        8                 5
## 4        Argyll and Bute                       7                 4
## 5      City of Edinburgh                      57                12
## 6       Clackmannanshire                      11                 5
## 7  Dumfries and Galloway                       6                 4
## 8            Dundee City                      39                20
## 9          East Ayrshire                      15                 6
## 10   East Dunbartonshire                       4                 1
## 11          East Lothian                       6                 2
## 12     East Renfrewshire                       4                 2
## 13           Eilean Siar                       1                 0
## 14               Falkirk                      14                 8
## 15                  Fife                      38                14
## 16          Glasgow City                     121                57
## 17              Highland                      15                 6
## 18            Inverclyde                      13                 8
## 19            Midlothian                       8                 0
```

```
## 20                  Moray                                  6                    4
## 21          North Ayrshire                                 19                   10
## 22       North Lanarkshire                                 38                   13
## 23          Orkney Islands                                  1                    0
## 24       Perth and Kinross                                  8                    4
## 25            Renfrewshire                                 26                   14
## 26        Scottish Borders                                  7                    2
## 27        Shetland Islands                                  2                    0
## 28          South Ayrshire                                  9                    4
## 29       South Lanarkshire                                 29                   12
## 30                Stirling                                  6                    5
## 31   West Dunbartonshire                                   19                    8
## 32            West Lothian                                 19                    5
## Variables not shown: Methadone (int), Any.benzo.diazepine (int),
Diazepam
##   (int), Temazepam (int), Cocaine (int), Ecstasy (int),
Amphetamines
##   (int), Alcohol (int), Latitude (dbl), Longitude (dbl), Population
(int),
##   prop_crime (dbl), prop_education (dbl), prop_employment (dbl),
##   prop_health (dbl), prop_income (dbl),
All.drug.related.deaths_perTenK
##   (dbl), LongSplit (fctr), LatSplit (fctr)
```

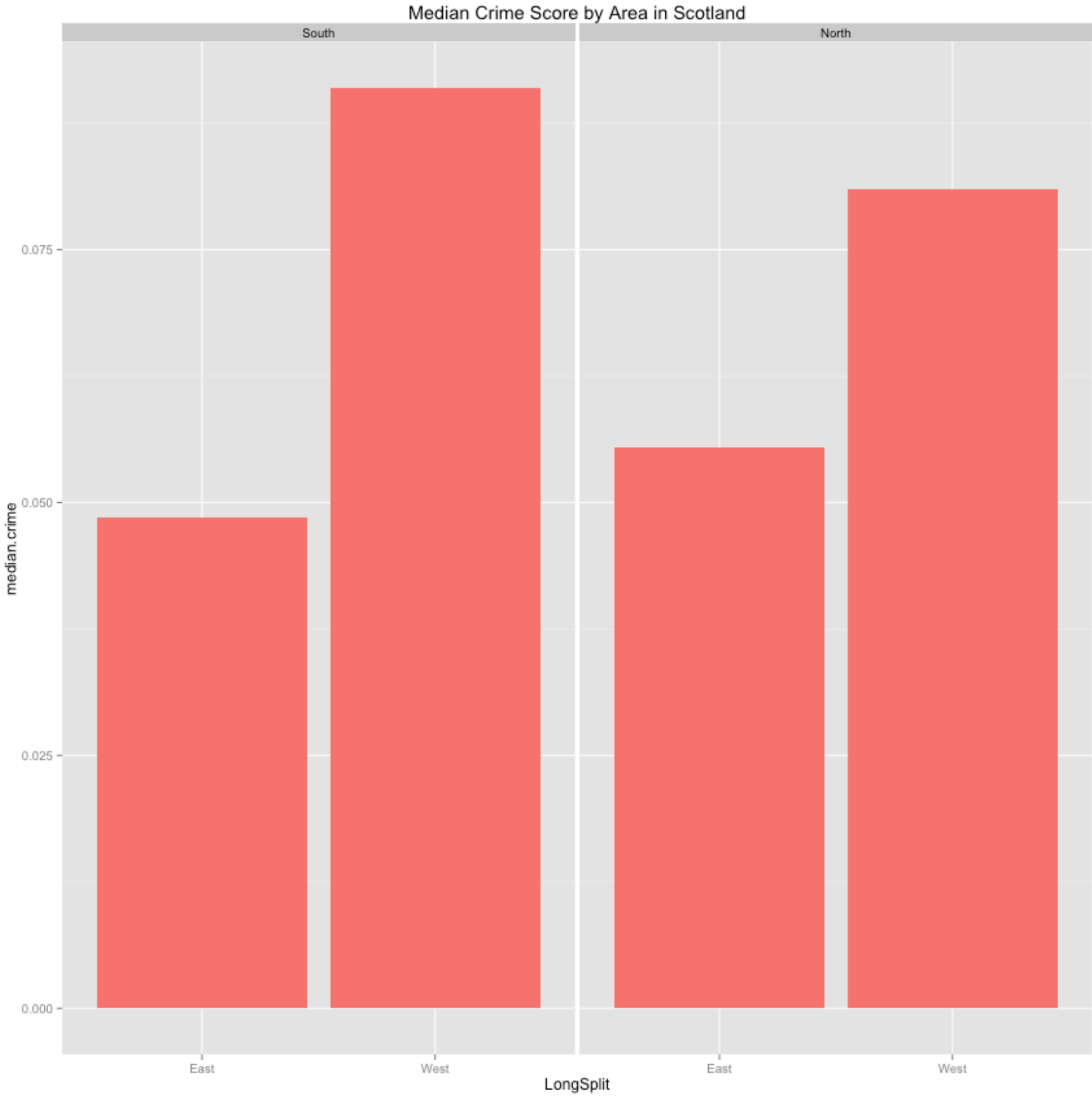
```
deaths_by_area
```

```
## Source: local data frame [4 x 7]
## Groups: LongSplit
##
##   LongSplit LatSplit median.deathsptk median.crime median.emp
median.edu
## 1      East    South           0.8032      0.04846    0.02281
0.03107
## 2      East    North           0.7302      0.05543    0.01077
0.03111
## 3      West    South           1.2061      0.09100    0.10934
0.08112
## 4      West    North           0.7308      0.08100    0.04339
0.04465
## Variables not shown: num.council.areas (int)
```

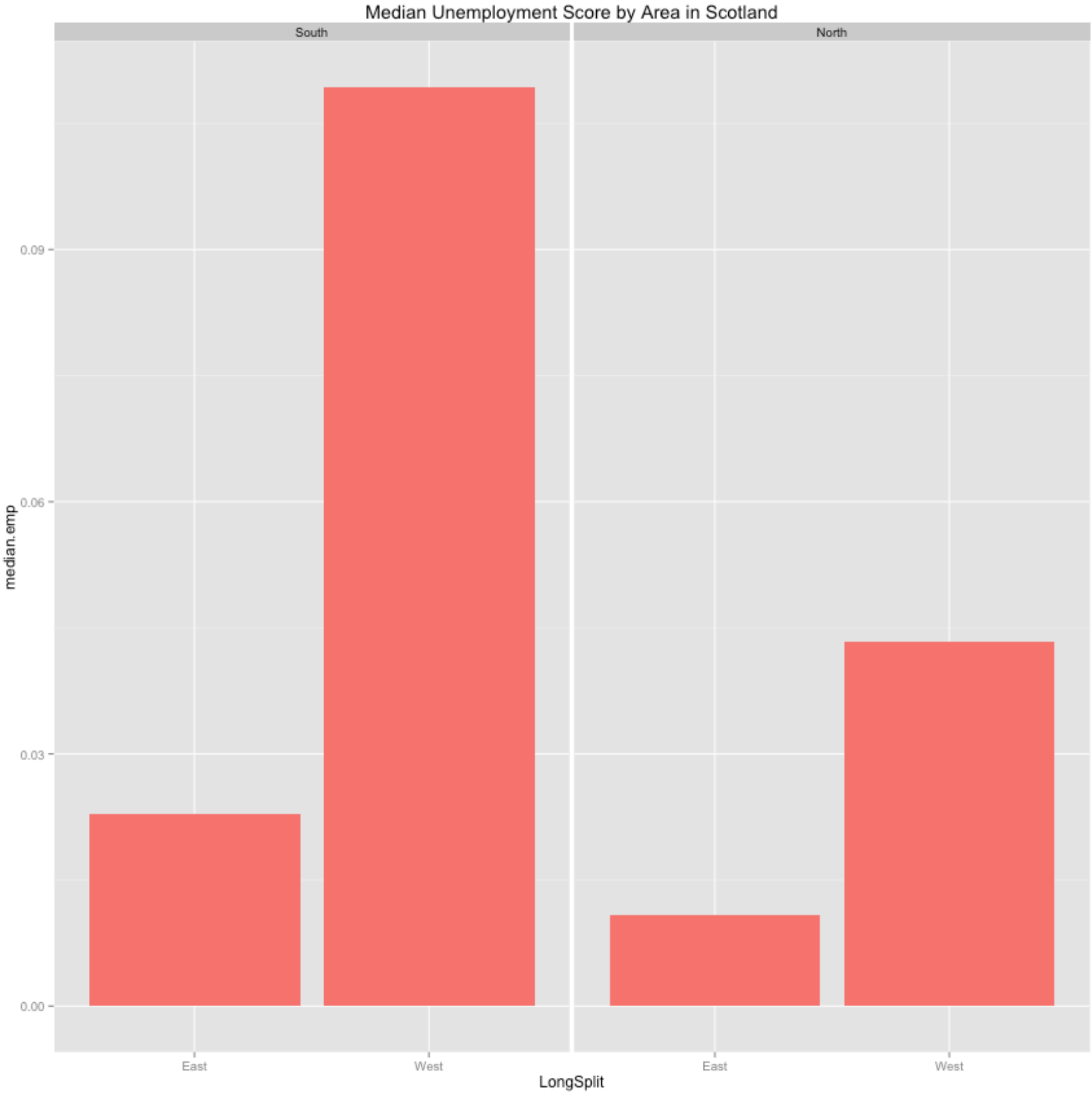
```
# Now we'll make some fun plots of the summary table

library(ggplot2)

qplot(LongSplit, median.deathsptk, data = deaths_by_area, facets =
~LatSplit,
    geom = "bar", stat = "identity", fill = "dark red", main = "Median
Deaths/10,000 by Area in Scotland") +
    theme(legend.position = "none")
```



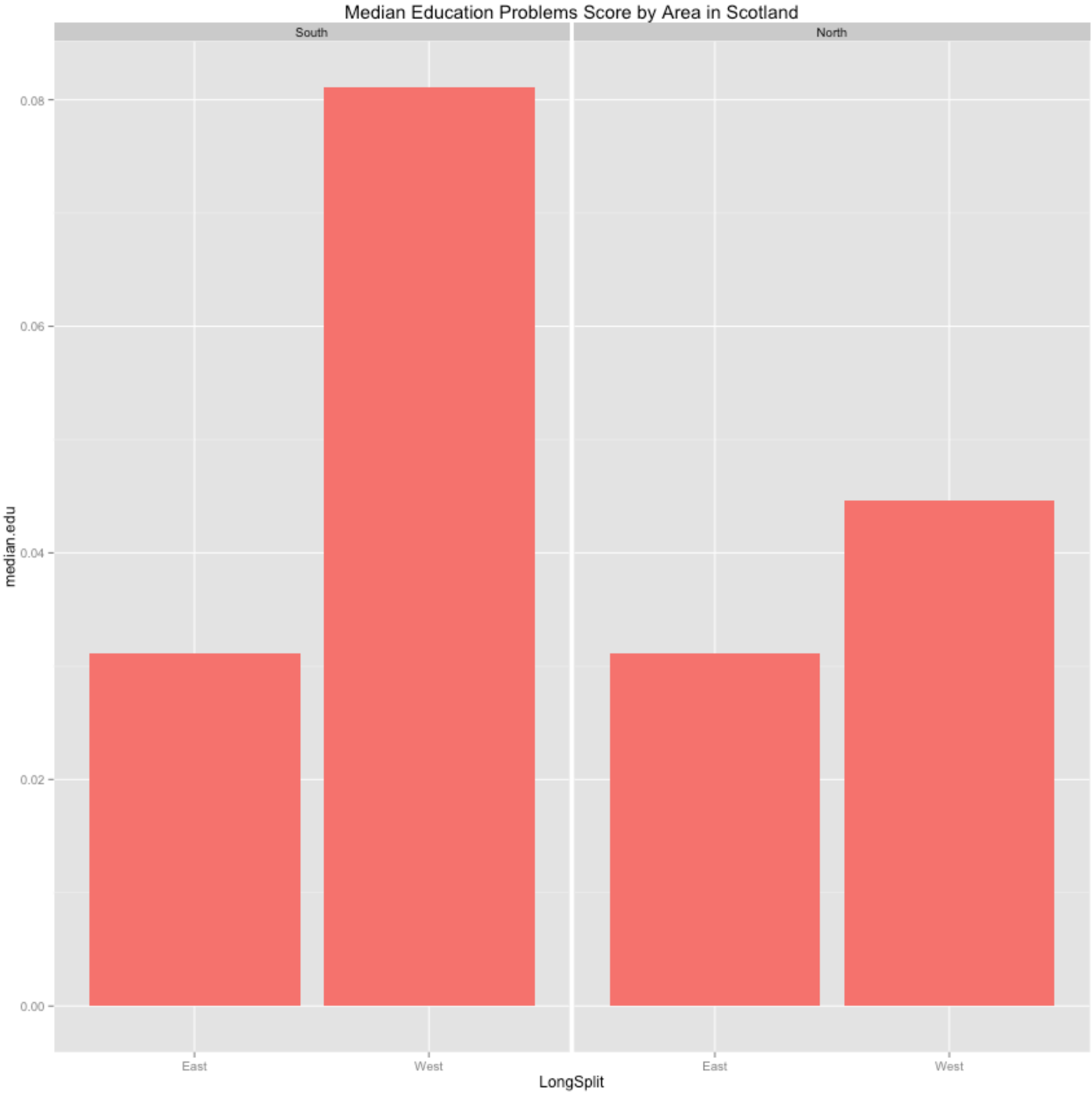

```
qplot(LongSplit, median.crime, data = deaths_by_area, facets =
~LatSplit, geom = "bar",
    stat = "identity", fill = "dark red", main = "Median Crime Score
by Area in Scotland") +
    theme(legend.position = "none")
```




```
qplot(LongSplit, median.emp, data = deaths_by_area, facets =
~LatSplit, geom = "bar",
    stat = "identity", fill = "dark red", main = "Median Unemployment
Score by Area in Scotland") +
    theme(legend.position = "none")
```




```
qplot(LongSplit, median.edu, data = deaths_by_area, facets =
~LatSplit, geom = "bar",
    stat = "identity", fill = "dark red", main = "Median Education
Problems Score by Area in Scotland") +
    theme(legend.position = "none")
```



8. What can we tell from these plots? Take a paragraph or so to share your thoughts.

****Some Online R Resources****

https://github.com/rheimann/UMBC Github is a social code repository. The link above is to where the code for this and other labs are stored.

http://www.r-bloggers.com If you are interested in R this is where you will find yourself spending alot of

your time. The site shares multiple blogs a day of varied topics.

http://stackoverflow.com/questions/tagged/r StackOverflow is a great site to go to for help.

```
sessionInfo()
```

```
## R version 3.0.2 (2013-09-25)
## Platform: x86_64-apple-darwin10.8.0 (64-bit)
##
## locale:
## [1] en_US.UTF-8/en_US.UTF-8/en_US.UTF-8/C/en_US.UTF-8/en_US.UTF-8
##
## attached base packages:
## [1] stats     graphics  grDevices utils     datasets  methods
base
##
## other attached packages:
## [1] ggplot2_0.9.3.1 dplyr_0.1.3     knitr_1.5
##
## loaded via a namespace (and not attached):
##  [1] assertthat_0.1     colorspace_1.2-4   dichromat_2.0-0
##  [4] digest_0.6.4       evaluate_0.5.1     formatR_0.10
##  [7] grid_3.0.2         gtable_0.1.2       labeling_0.2
## [10] MASS_7.3-29        munsell_0.4.2      plyr_1.8
## [13] proto_0.3-10       RColorBrewer_1.0-5 Rcpp_0.11.0
## [16] reshape2_1.2.2     scales_0.2.3       stringr_0.6.2
## [19] tools_3.0.2
```