

hw7.R

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Mon Oct 20 11:22:53 2014

```
# Attach necessary spatial packages
library(rgdal)
```

```
## Loading required package: sp
## rgdal: version: 0.9-1, (SVN revision 518)
## Geospatial Data Abstraction Library extensions to R successfully loaded
## Loaded GDAL runtime: GDAL 1.10.1, released 2013/08/26
## Path to GDAL shared files: /usr/share/gdal/1.10
## Loaded PROJ.4 runtime: Rel. 4.8.0, 6 March 2012, [PJ_VERSION: 480]
## Path to PROJ.4 shared files: (autodetected)
```

```
library(gstat)
library(geoR)
```

```
## Loading required package: MASS
## -----
## Analysis of geostatistical data
## For an Introduction to geoR go to http://www.leg.ufpr.br/geoR
## geoR version 1.7-4.1 (built on 2012-06-29) is now loaded
## -----
```

```
library(RColorBrewer)
```

```
# Set wd
setwd("/home/joebrew/Documents/uf/phc6194/hw7")
```

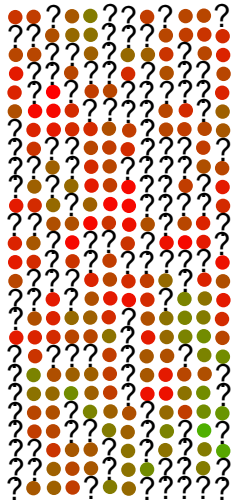
```
# Read in data
s1 <- readOGR("data", "SubSample1")
```

```
## OGR data source with driver: ESRI Shapefile
## Source: "data", layer: "SubSample1"
## with 187 features and 6 fields
## Feature type: wkbPoint with 2 dimensions
```

```
s2 <- readOGR("data", "SubSample2")
```

```
## OGR data source with driver: ESRI Shapefile
## Source: "data", layer: "SubSample2"
## with 125 features and 6 fields
## Feature type: wkbPoint with 2 dimensions
```

```
# Plot
xcol <- colorRampPalette(c("green", "red"))(max(ceiling(s1$PM25)))
plot(s1, pch = 16, col = xcol[ceiling(s1$PM25)])
points(s2, col = "black", pch = "?")
```



```
# Polygonal interpolation #####

# Create thiessen / voronoi polygons
# first, function:
# (from http://stackoverflow.com/questions/9403660/how-to-create-thiessen-polygons-from-points-using-r)
voronoipolygons <- function(x) {
  require(deldir)
  require(sp)
  if (.hasSlot(x, 'coords')) {
    crds <- x@coords
  } else crds <- x
  z <- deldir(crds[,1], crds[,2])
  w <- tile.list(z)
  polys <- vector(mode='list', length=length(w))
  for (i in seq(along=polys)) {
    pcrds <- cbind(w[[i]]$x, w[[i]]$y)
    pcrds <- rbind(pcrds, pcrds[1,])
    polys[[i]] <- Polygons(list(Polygon(pcrds)), ID=as.character(i))
  }
  SP <- SpatialPolygons(polys)
  voronoi <- SpatialPolygonsDataFrame(SP, data=data.frame(x=crds[,1],
                                                            y=crds[,2], row.names=apply(slot(SP, 'polygons'),
                                                            function(x) slot

```

```
## Loading required package: deldir
## deldir 0.1-6
```

```
##
## PLEASE NOTE: The components "delsgs" and "summary" of the
## object returned by deldir() are now DATA FRAMES rather than
## matrices (as they were prior to release 0.0-18).
## See help("deldir").
##
```

```

##      PLEASE NOTE: The process that deldir() uses for determining
##      duplicated points has changed from that used in version
##      0.0-9 of this package (and previously). See help("deldir").

# Bring in data
s1poly@data <- s1@data

# Define projection string for s1poly
proj4string(s1poly) <- proj4string(s1)

# Which of s1's polygons do all the s2 points fall into?
x <- over(s2, polygons(s1poly))

# Get all of s1's appropriate values for the s2 points

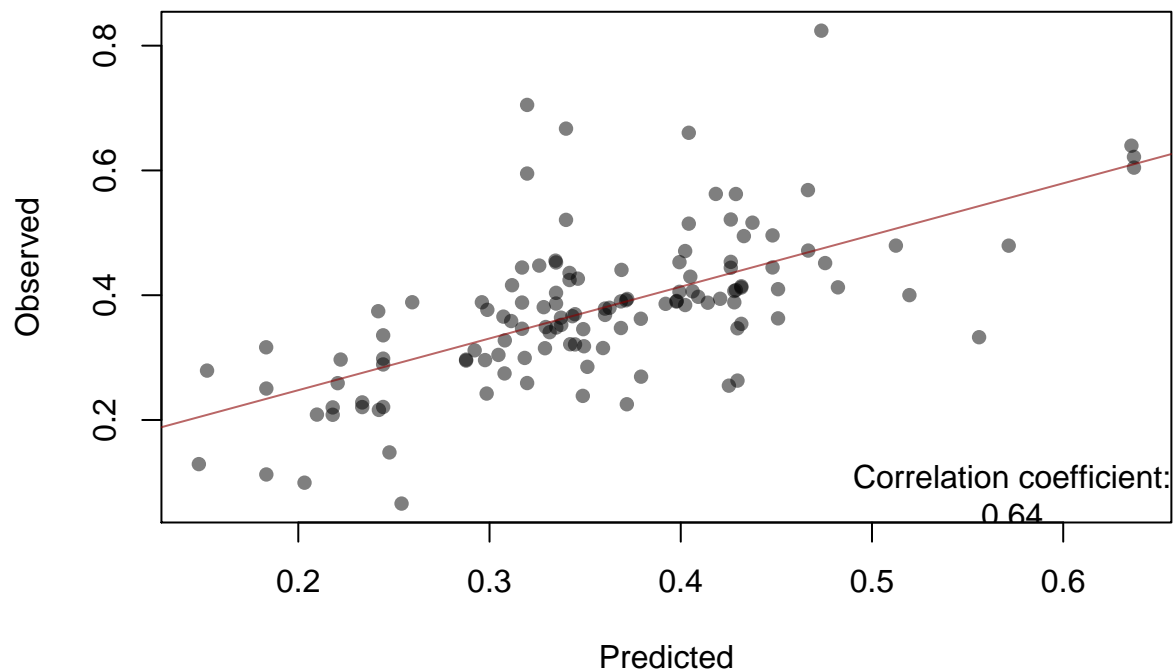
# Monthly aod
s2$MonthlyAOD.p <- NA
for (i in 1:nrow(s2)){
  ind <- x[i]
  s2$MonthlyAOD.p[i] <-
    s1poly$MonthlyAOD[ind]
}

# PM25
s2$PM25.p <- NA
for (i in 1:nrow(s2)){
  ind <- x[i]
  s2$PM25.p[i] <-
    s1poly$PM25[ind]
}

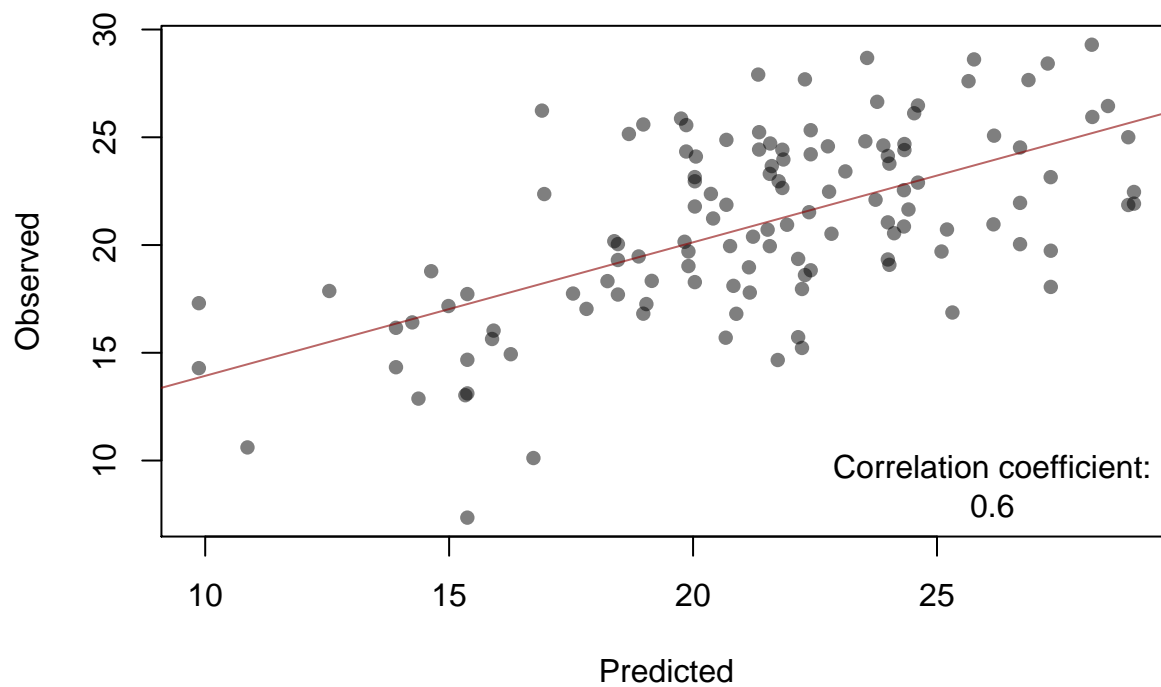
# Plot Correlation between monthly AOD true and predicted
JoePlot <- function(x,y){
  plot(x,y,
    xlab = "Predicted",
    ylab = "Observed",
    pch = 16,
    col = adjustcolor("black", alpha.f = 0.5))
  mylm <- lm(y ~x)
  abline(mylm, col = adjustcolor("darkred", alpha.f = 0.6))
  mycor <- cor(x,y)
  text(0.9*max(x),
    1.2*min(y),
    labels = paste0("Correlation coefficient:\n", round(mycor, digits =2)))
}

# MONTHLY AOD
JoePlot(s2$MonthlyAOD.p, s2$MonthlyAOD)

```

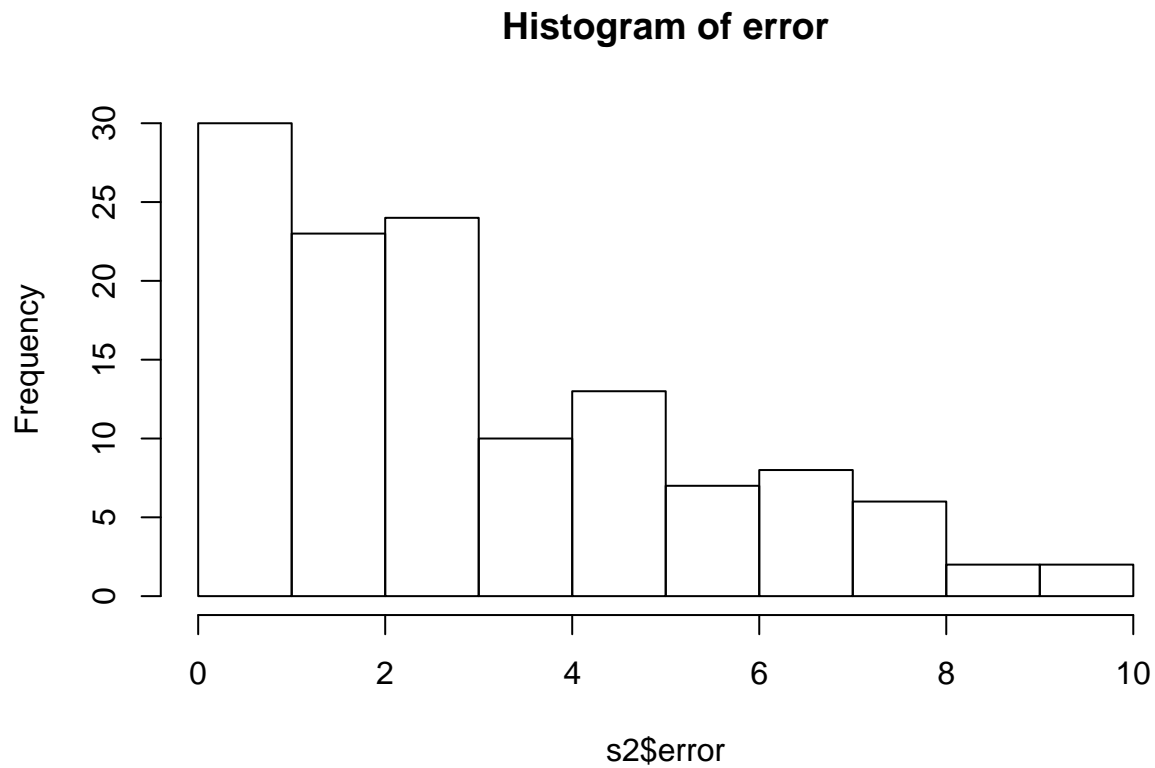


```
# PM 25
JoePlot(s2$PM25.p, s2$PM25)
```



```
# Calculate error
ErrorCalc <- function(x,y){
  z <- x-y
  z[which(z <= 0)] <- z[which(z <= 0)] * -1
  return(z)
}
s2$error <- ErrorCalc(s2$PM25.p, s2$PM25)
```

```
# Explore error  
hist(s2$error, main = "Histogram of error")
```



```
# Fill out table  
mean(s2$PM25.p)
```

```
## [1] 21.32539
```

```
median(s2$PM25.p)
```

```
## [1] 21.58075
```

```
min(s2$PM25.p)
```

```
## [1] 9.869708
```

```
max(s2$PM25.p)
```

```
## [1] 29.0372
```

```
sd(s2$PM25.p)
```

```
## [1] 4.149048
```

```
mean(s2$error)
```

```
## [1] 2.96558
```

```
median(s2$error)
```

```
## [1] 2.240936
```

```
min(s2$error)
```

```
## [1] 0.075162
```

```
max(s2$error)
```

```
## [1] 9.335048
```

```
sd(s2$error)
```

```
## [1] 2.296784
```

```
mean(s2$error^2)
```

```
## [1] 14.02768
```

```
## TRIANGULATION #####
```

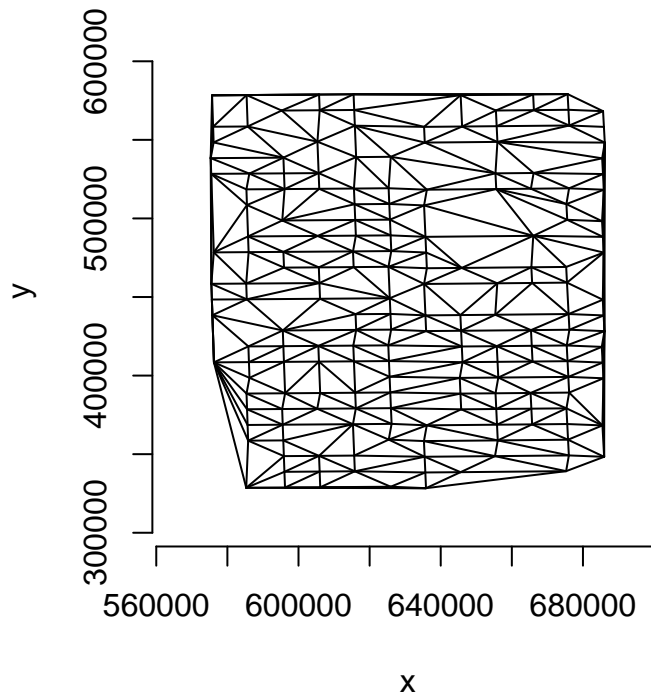
```
# Calculate triangles
```

```
library(deldir)
```

```
s1tri <- deldir(x = coordinates(s1)[,1],  
               y = coordinates(s1)[,2])
```

```
# plot
```

```
plot(s1tri, wl = "triang", pch = NA)
```



```
# calculate tile perimeters
tilePerim(tile.list(s1tri))
```

```
## $perimeters
## [1] 104199.96 95335.94 87963.06 100767.65 111084.28 87847.33 104439.15
## [8] 56881.75 47193.40 56033.32 49088.38 44767.42 39612.18 74294.65
## [15] 63160.89 60044.83 58665.71 62763.20 51817.43 47194.71 56716.21
## [22] 52557.11 61207.26 61128.05 66066.88 69322.43 67097.81 59910.19
## [29] 56499.76 58332.12 67508.33 68930.67 70841.94 44614.42 39485.09
## [36] 48170.59 63259.65 59245.33 59104.44 54812.99 47635.68 47909.01
## [43] 46326.03 50746.62 62772.76 68009.65 53397.43 43672.37 60011.18
## [50] 81313.23 48176.76 39292.10 63536.59 58199.46 59810.35 47313.27
## [57] 43804.54 58387.86 60669.56 72436.46 52701.07 40291.65 39740.39
## [64] 66949.98 82043.77 60756.67 84417.55 46402.49 47583.41 44139.45
## [71] 39689.72 53823.10 75074.27 48507.27 43493.87 47612.86 47236.42
## [78] 40723.94 62796.70 56830.29 59186.35 62621.36 54459.56 55389.17
## [85] 56861.28 59413.52 51085.70 61140.23 52634.91 56411.73 63638.56
## [92] 55505.68 70909.41 74841.62 52239.18 39820.05 46529.77 51104.30
## [99] 59475.67 52061.41 52075.50 61642.73 47172.41 39122.31 44173.44
## [106] 43768.63 48956.43 40061.04 52913.27 58276.55 39663.17 44461.91
## [113] 39807.22 46500.32 47232.65 43621.16 39664.94 39463.91 52460.79
## [120] 84710.04 39679.07 46727.86 49677.37 46727.86 50142.29 49764.49
## [127] 39680.83 43666.04 39120.31 52653.45 55428.66 53819.06 50391.53
## [134] 44126.06 43422.91 66855.19 70896.56 46907.20 49508.17 46754.45
## [141] 49193.18 49196.03 39519.88 44072.56 54679.33 72485.94 39533.73
## [148] 43891.03 40187.02 47066.33 50506.67 39728.21 39728.21 53738.50
## [155] 72798.28 46748.13 51729.71 40620.82 51429.52 46750.37 43765.42
## [162] 39931.20 74006.34 86599.56 47886.78 48255.71 49972.61 56989.44
## [169] 50956.04 53752.43 47847.99 53609.62 54144.52 67096.52 49994.06
## [176] 87740.38 53499.51 39788.98 50722.68 46071.91 85675.69 138067.45
## [183] 128748.01 81791.04 88251.35 90170.69 114170.58
```

```

##
## $totalPerim
## [1] 10672339
##
## $meanPerim
## [1] 57071.33

# Create list of triangles
trilist <- triang.list(s1tri)

# Make list of triangles a polygon object
mypolys <- list()
for (i in 1:length(trilist)){

  x <- rbind(cbind(trilist[[i]]$x, trilist[[i]]$y),
             cbind(trilist[[i]]$x[1], trilist[[i]]$y[1]))

  x <- Polygon(x)

  mypolys[i] <- x
}

# Create a spatial polygons object from my polys
mytris <- SpatialPolygons(list(Polygons(mypolys, ID = 1)))

# give proj4string
proj4string(mytris) <- proj4string(s1)

# Get the mean value of the three coordinates which make up
# the vertices of each triangle
# And populate a vector
val_vector <- vector(length = length(mytris@polygons[[1]]@Polygons),
                     mode = "numeric")
for (i in 1: length(mytris@polygons[[1]]@Polygons)){

  # Extract vertices
  mymat <- mytris@polygons[[1]]@Polygons[[i]]@coords

  # Remove fourth (since it's just the loop back to close the triangle)
  mymat <- mymat[-4,]

  # #make df
  # mymat <- data.frame(mymat)

  # Get corresponding values in s1
  myvals <- s1$PM25[which(round(coordinates(s1)[,1], digits = 1) %in% round(mymat[,1], digits = 1) &
                        round(coordinates(s1)[,2], digits = 1) %in% round(mymat[,2], digits = 1))]

  #Get mean
  myval <- mean(myvals)
  val_vector[i] <- myval
}

```

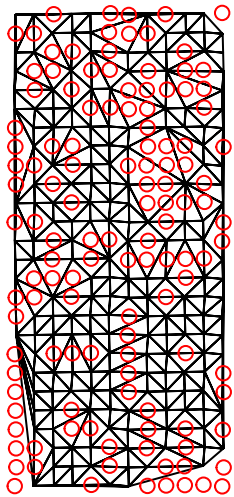


```

# Bind the triangles with the val_vector (the interpolated values) # won't work
# mytris2 <- SpatialPolygonsDataFrame(mytris@polygons[[1]]@Polygons,
#                                   data.frame("predicted" = val_vector))

# Can't quite get this to work
plot(mytris) # prediction triangles
points(s2, col = "red")

```



```

## IDW #####
library(spatstat)

##
## spatstat 1.38-1      (nickname: 'Le Hardi')
## For an introduction to spatstat, type 'beginner'
##
## Attaching package: 'spatstat'
##
## The following object is masked from 'package:gstat':
##
##     idw

```

```

# # make ppp object of coordinates s1 and s2
# s1ppp <- as.ppp(coordinates(s1),
#                 c(min(coordinates(s1)[,1]),
#                   max(coordinates(s1)[,1]),
#                   min(coordinates(s1)[,2]),
#                   max(coordinates(s1)[,2])))
# s2ppp <- as.ppp(coordinates(s2),
#                 c(min(coordinates(s2)[,1]),
#                   max(coordinates(s2)[,1]),
#                   min(coordinates(s2)[,2]),
#                   max(coordinates(s2)[,2])))
# idw(x = s1ppp,
#     power = 1,
#     at = s2ppp)

```

```
## KRIGING #####
# Define color vector (using colorbrewer)
my_colors <- colorRampPalette(c("darkgreen", "red"))(100)

# boundary_points <- boundary@polygons[[1]]@Polygons
# boundary_points <- boundary_points[[1]]@coords

# Get trap locations and data values
a <- data.frame("x" = coordinates(s1)[,1],
               "y" = coordinates(s1)[,2],
               "z" = s1$PM25)
# Make into a geodata object
b <- as.geodata(a)

# Predict multiple points in Alachua County's boundary
x <- seq(min(coordinates(s2)[,1]), max(coordinates(s2)[,1]), length = 100)
y <- seq(min(coordinates(s2)[,2]), max(coordinates(s2)[,2]), length = 100)

# Make a grid of those points
pred.grid <- expand.grid(x,y)

# kriging calculations
kc <- krige.conv(geodata = b, coords = b$coords, data = b$data,
               locations = pred.grid,
               krige = krige.control(type.krige = "ok",
                                   cov.pars = c(10, 10000)))
```

```
## krige.conv: model with constant mean
## krige.conv: Kriging performed using global neighbourhood
```

```
# Plot!
# displaying predicted values
image(kc, loc = pred.grid,
      col = my_colors,
      xlab=NA, ylab=NA,
      xaxt = "n",
      yaxt = "n",
      xpd = NA,
      bty = "n")

# Define percentiles for legend
legtemp <- round(quantile(kc$predict, probs = seq(0,1,, length = 10)))

legend(x="topright",
      fill = my_colors[c(1,11,22,33,44,55,66,77,88,100)],
      legend = c(legtemp[1], NA, NA, legtemp[4], NA, NA, legtemp[7], NA, NA, legtemp[10]),
      border = FALSE,
      bty = "n",
      ncol = 1,
```

```
y.intersp = 0.5,  
title = " KrigingInterpolation",  
cex = 0.75)
```

