Hand-in exercise: Murchison data

library(spatstat)

In this exercise we work with the spatstat dataset murchison rescaled to km (applying rescale to a spatial object list (solist)):

```
mur <- solapply(murchison, rescale, s = 1000, unitname = "km")</pre>
```

Exercise 1

Read the help file for murchison (in spatstat.data) and reproduce the plot given in the *Examples* section of the help file.

Exercise 2

Add the distance to the neareast fault line to the spatial object list mur:

```
mur$dfault <- distfun(mur$faults)</pre>
```

Now, consider the Poisson model:

```
model_d <- ppm(gold ~ dfault, data = mur)
model_d</pre>
```

```
## Nonstationary Poisson process
##
## Log intensity: ~dfault
##
## Fitted trend coefficients:
## (Intercept)
                    dfault
##
   -4.3412775 -0.2607664
##
##
                                S.E.
                                        CI95.lo
                                                   CI95.hi Ztest
                 Estimate
                                                                      Zval
## (Intercept) -4.3412775 0.08556260 -4.5089771 -4.1735779
                                                             *** -50.73802
              -0.2607664 0.02018789 -0.3003339 -0.2211988
## dfault
                                                             *** -12.91697
```

Write the estimated intensity function $\hat{\lambda}(u)$ as a function of the distance to the nearest fault, D(u), with the parameter values inserted.

Exercise 3

Consider the model

```
model_g <- ppm(gold ~ greenstone, data = mur)
model_g</pre>
```

```
## Nonstationary Poisson process
##
## Log intensity: ~greenstone
##
## Fitted trend coefficients:
## (Intercept) greenstoneTRUE
```

```
## -8.103178 3.980409

##

## Estimate S.E. CI95.lo CI95.hi Ztest Zval

## (Intercept) -8.103178 0.1666667 -8.429839 -7.776517 *** -48.61907

## greenstoneTRUE 3.980409 0.1798443 3.627920 4.332897 *** 22.13252
```

- What does this model state about the intensity function? (*Hint*: the plot you produce below may be helpful.)
- Use predict.ppm() to calculate the estimated intensity function and plot it.

Exercise 4

Consider the model

```
model_dg <- ppm(gold ~ dfault + greenstone, data = mur)
model_dg</pre>
```

```
## Nonstationary Poisson process
##
## Log intensity: ~dfault + greenstone
##
## Fitted trend coefficients:
##
      (Intercept)
                          dfault greenstoneTRUE
       -6.6171116
                      -0.1037835
                                       2.7539637
##
##
##
                    Estimate
                                    S.E.
                                            CI95.1o
                                                        CI95.hi Ztest
                                                                             Zval
## (Intercept)
                  -6.6171116 0.21707953 -7.0425796 -6.19164351
                                                                   *** -30.482430
## dfault
                  -0.1037835 0.01794981 -0.1389645 -0.06860255
                                                                   ***
                                                                       -5.781874
## greenstoneTRUE 2.7539637 0.20655423 2.3491248 3.15880250
                                                                        13.332885
```

and write down $\hat{\lambda}(u)$ as a function of the distance to the nearest fault, D(u), and the greenstone indicator function

$$G(u) = 1\{u \text{ in greenstone area}\}\$$

with the parameter values inserted.

Exercise 5

Fit a cluster model of your choice to the mur data with the same intensity model using kppm(gold ~ dfault + greenstone, ..., data = mur).

Compare the standard errors obtained for this cluster model with the standard errors for the Poisson model model_dg.