

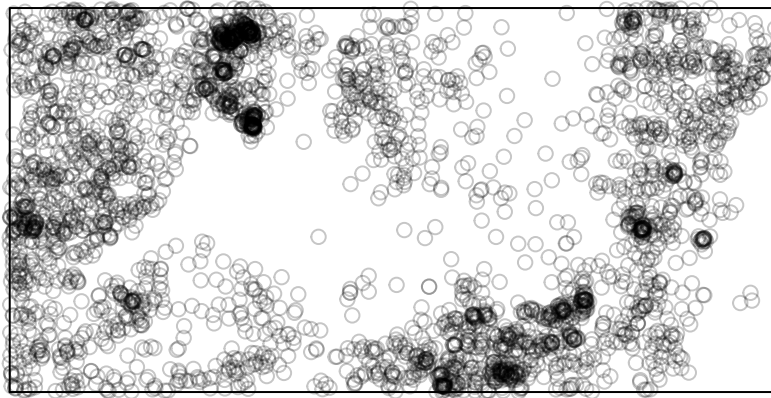
Modeling Point Processes with Spatstat

Model Fitting

The `ppm` function can be used for model fitting with a point process.

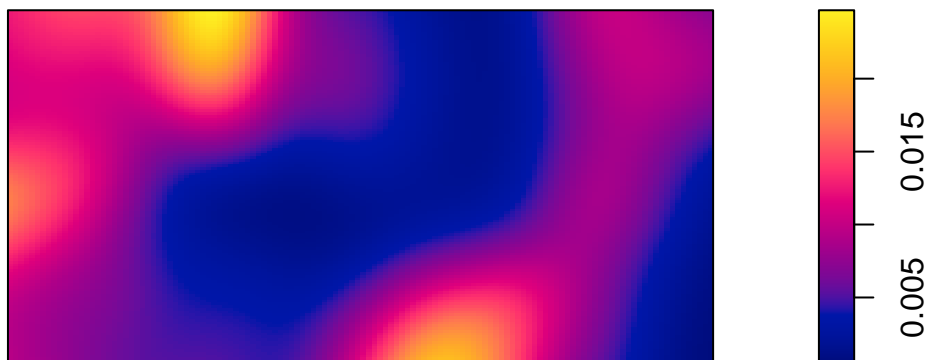
```
plot(bei)
```

bei



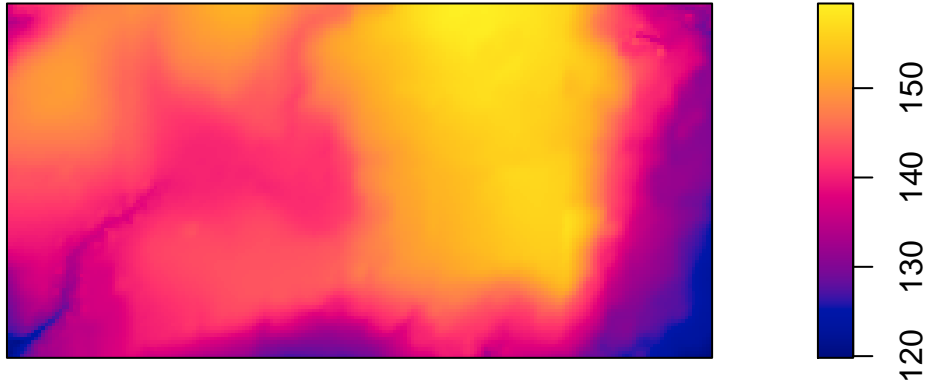
```
plot(density.ppp(bei))
```

density.ppp(bei)



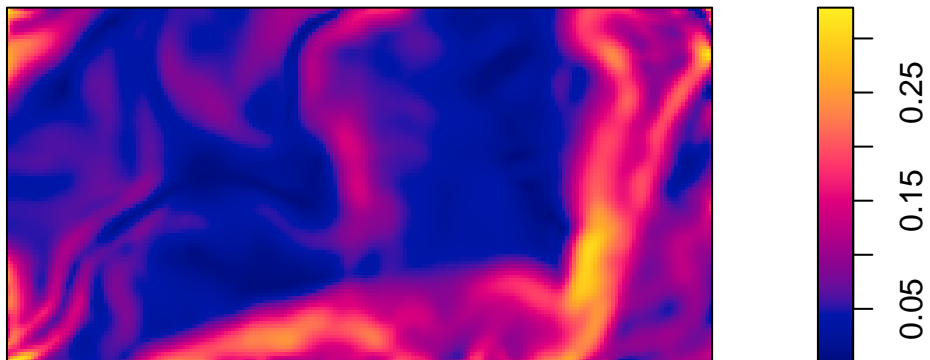
```
elev <- bei.extra$elev  
grad <- bei.extra$grad  
plot(elev)
```

elev



```
plot(grad)
```

grad

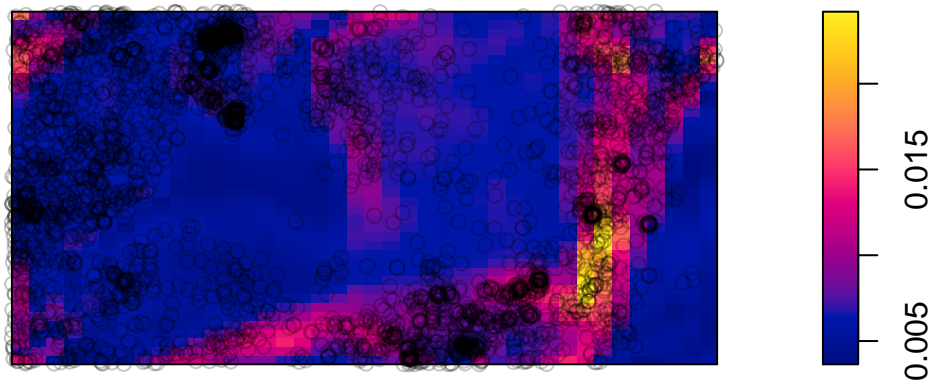


The ppm function allows model fitting

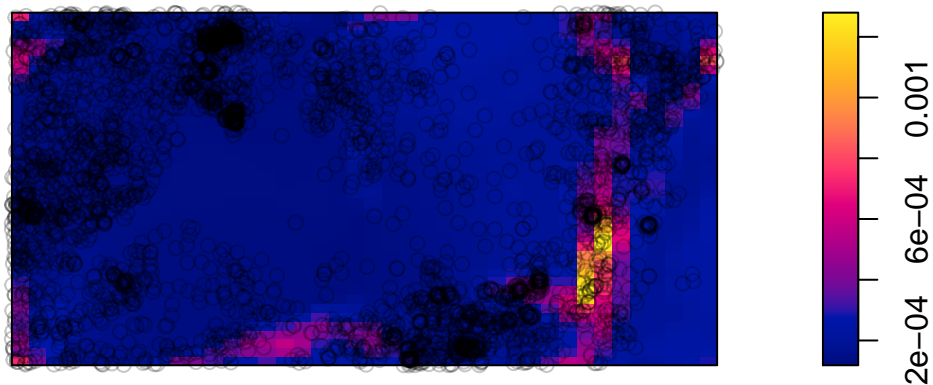
```
tree.model <- ppm(bei ~ elev + grad);  
tree.model
```

```
## Nonstationary Poisson process  
##  
## Log intensity: ~elev + grad  
##  
## Fitted trend coefficients:  
## (Intercept)      elev      grad  
## -8.56355220  0.02143995  5.84646680  
##  
##              Estimate      S.E.      CI95.lo      CI95.hi Ztest      Zval  
## (Intercept) -8.56355220 0.341113849 -9.23212306 -7.89498134 *** -25.104675  
## elev         0.02143995 0.002287866  0.01695581  0.02592408 ***  9.371155  
## grad         5.84646680 0.255781018  5.34514522  6.34778838 *** 22.857313  
  
plot(tree.model)
```

Fitted trend



Estimated se

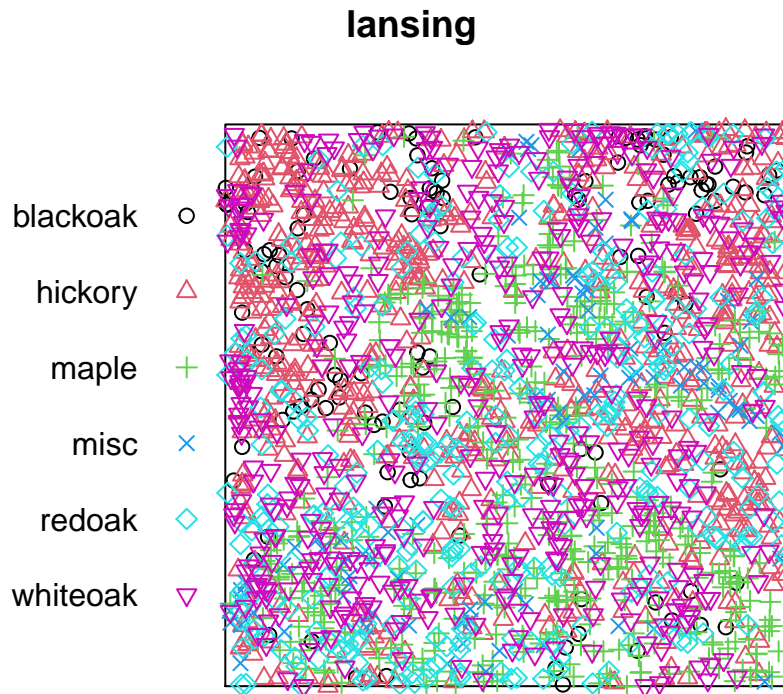


For more complicated models, kppm can be used for clustering behavior.

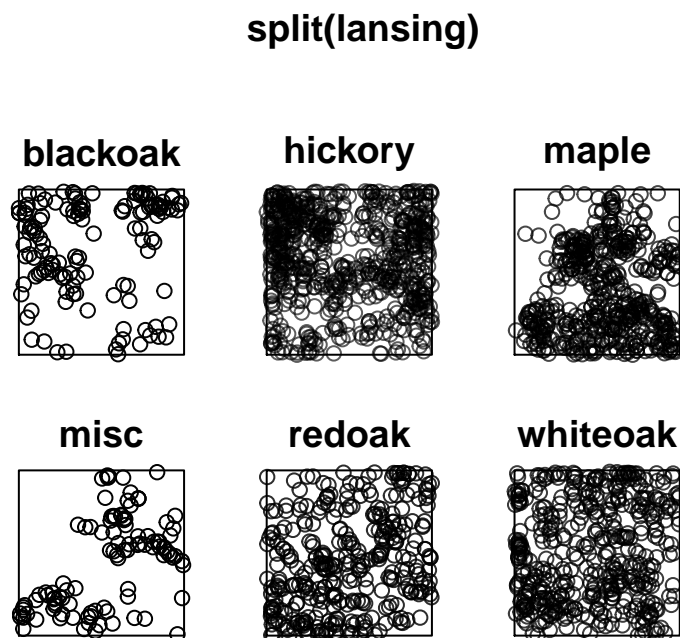
Marked Point Patterns

The `lansing` data set contains locations of six types of trees.

```
plot(lansing, cols = 1:6)
```



```
plot(split(lansing))
```



To analyze this data, consider the following model.

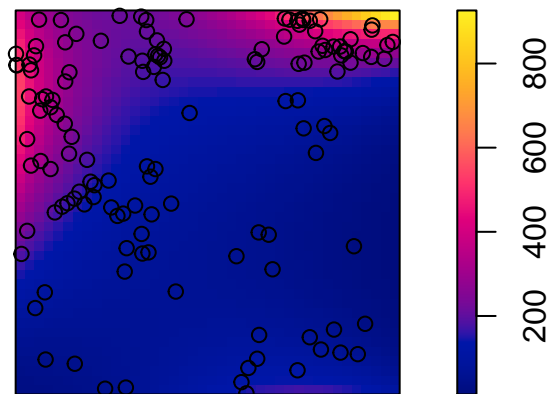
```
lansing.model <- ppm(lansing ~ marks - 1)
lansing.model

## Stationary multitype Poisson process
##
## Possible marks: 'blackoak', 'hickory', 'maple', 'misc', 'redoak' and 'whiteoak'
##
## Log intensity: ~marks - 1
##
## Intensities:
## beta_blackoak  beta_hickory  beta_maple  beta_misc  beta_redoak
##           135           703           514           105           346
## beta_whiteoak
##           448
##
##           Estimate      S.E. CI95.lo CI95.hi Ztest      Zval
## marksblackoak 4.905275 0.08606630 4.736588 5.073962 *** 56.99414
## markshickory  6.555357 0.03771571 6.481435 6.629278 *** 173.80970
## marksmapple   6.242223 0.04410811 6.155773 6.328674 *** 141.52099
## marksmisc     4.653960 0.09759001 4.462687 4.845233 *** 47.68890
## marksredoak   5.846439 0.05376033 5.741070 5.951807 *** 108.75005
## markswhiteoak 6.104793 0.04724556 6.012194 6.197393 *** 129.21412
```

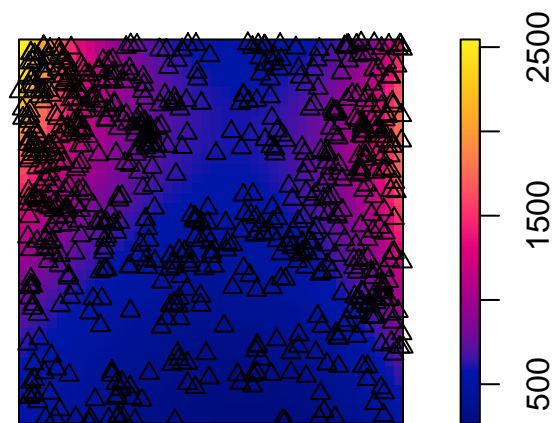
In contrast with this model, we can also include

```
lansing.model2 <- ppm(lansing ~ marks * polynom(x,y,3))
#lansing.model2
plot(lansing.model2)
```

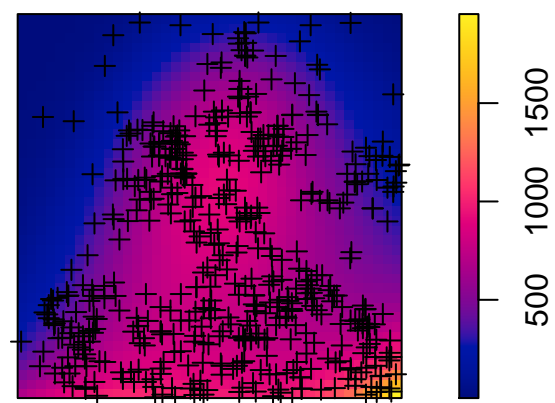
**Fitted trend
mark = blackoak**



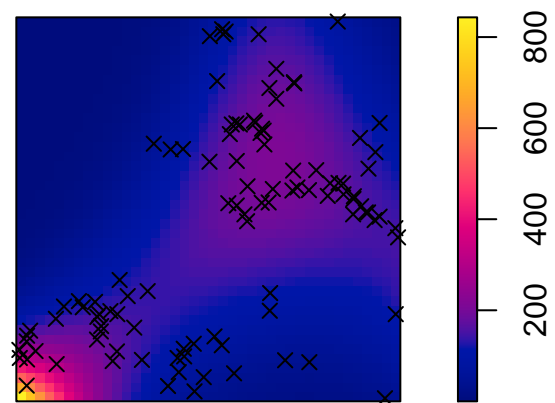
**Fitted trend
mark = hickory**



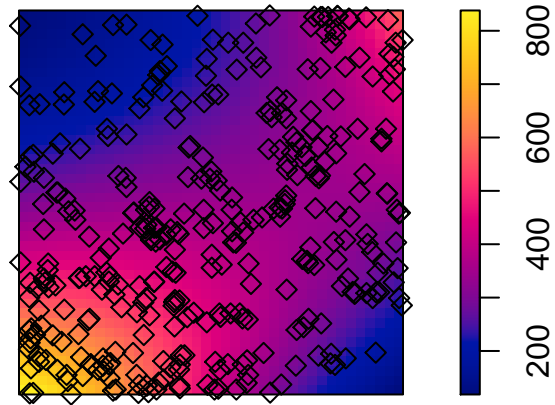
**Fitted trend
mark = maple**



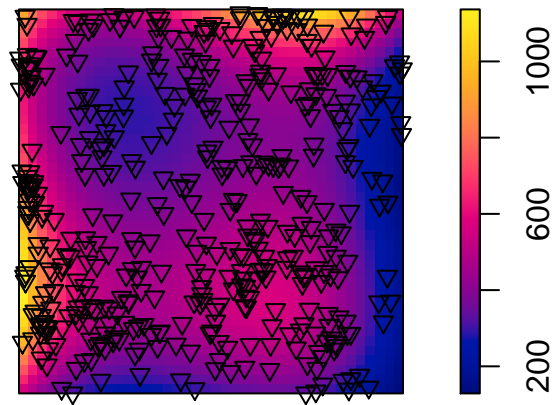
**Fitted trend
mark = misc**



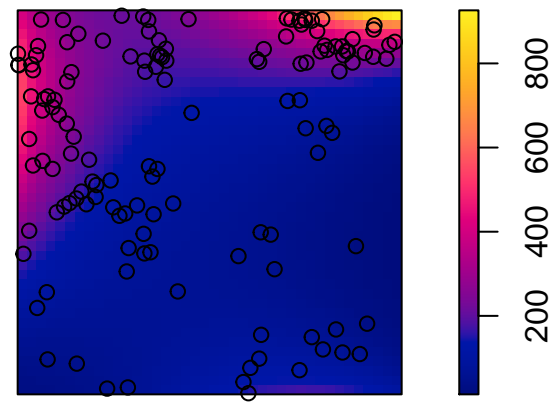
Fitted trend
mark = redoak



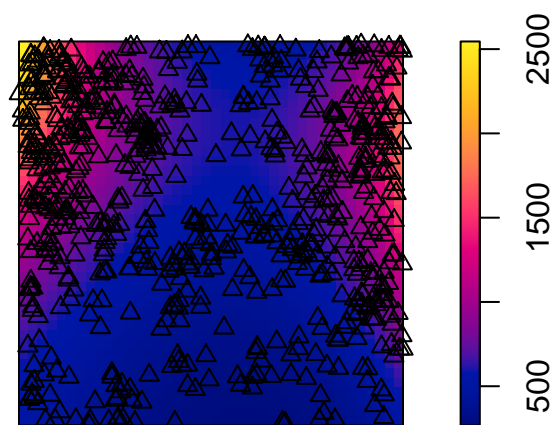
Fitted trend
mark = whiteoak



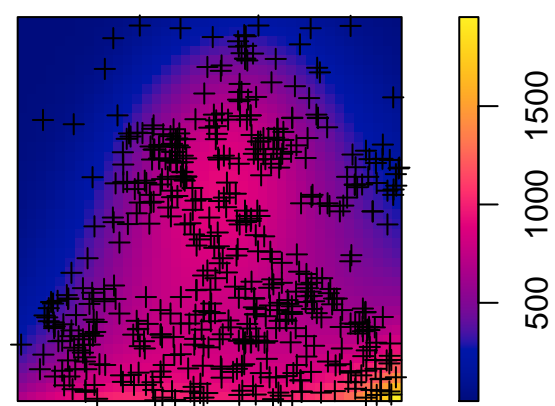
Estimated se
mark = blackoak



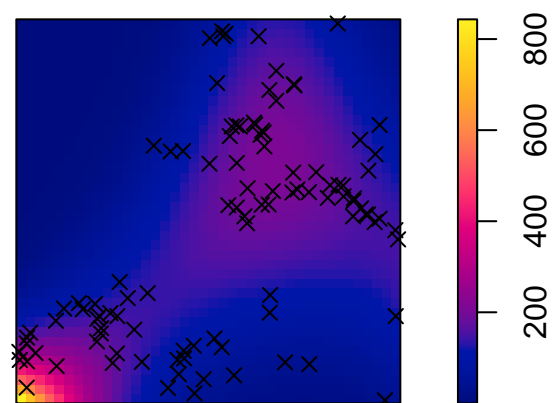
**Estimated se
mark = hickory**



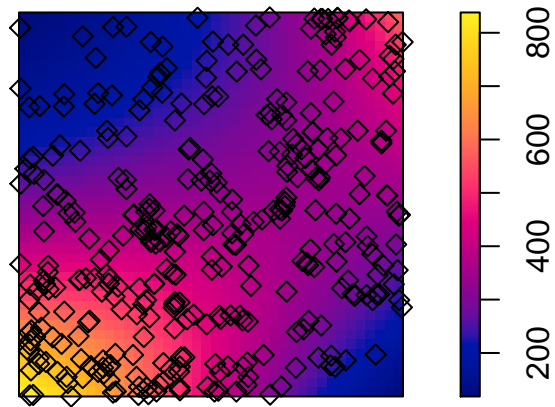
**Estimated se
mark = maple**



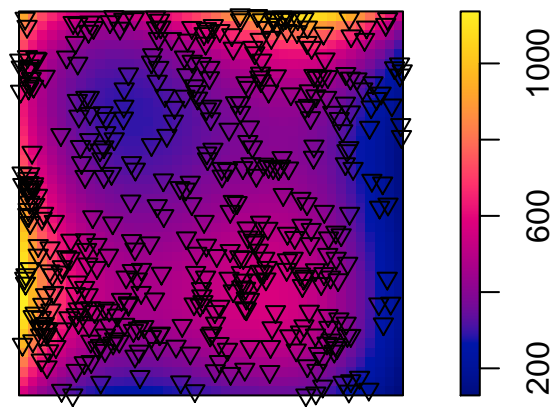
**Estimated se
mark = misc**



**Estimated se
mark = redoak**



**Estimated se
mark = whiteoak**



More advanced point pattern models

Cluster processes Clustering is not well defined. In general the idea is that the point distances are shorter than expected. However, there “is a fundamental ambiguity between heterogeneity and clustering” (Diggle 2007).