

ColoradoExample.R

nychka

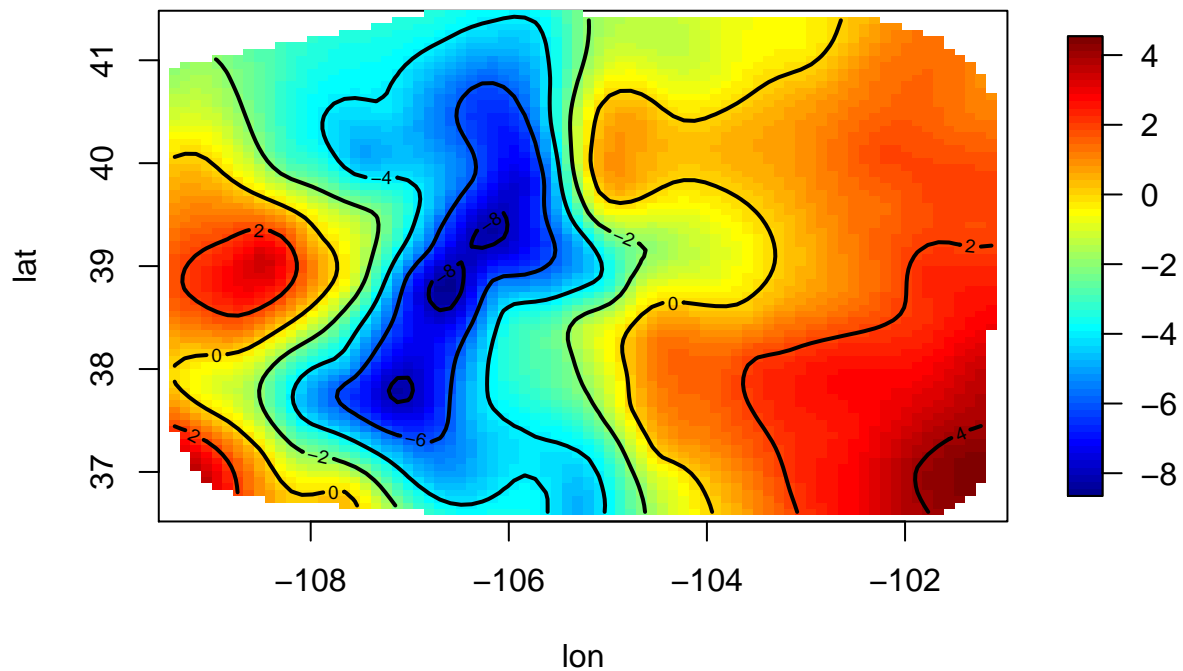
2019-09-13

```
suppressMessages(library( fields))

# colorado climate data
data(COmonthlyMet)

x<- CO.loc
y<- CO.tmin.MAM.climate
elev<- CO.elev
good<- !is.na( y)
x<- x[good,]
y<- y[good]
elev<- elev[good]

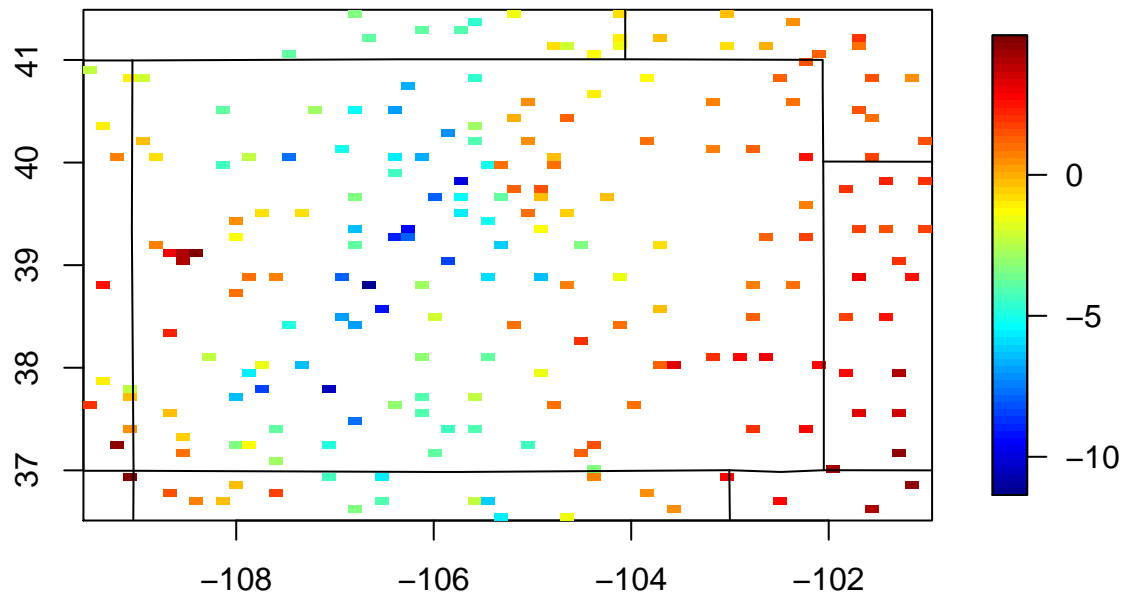
### A quick spatial analysis -- lots of defaults and w/o elevation
obj<- spatialProcess( x,y)
surface( obj)
```



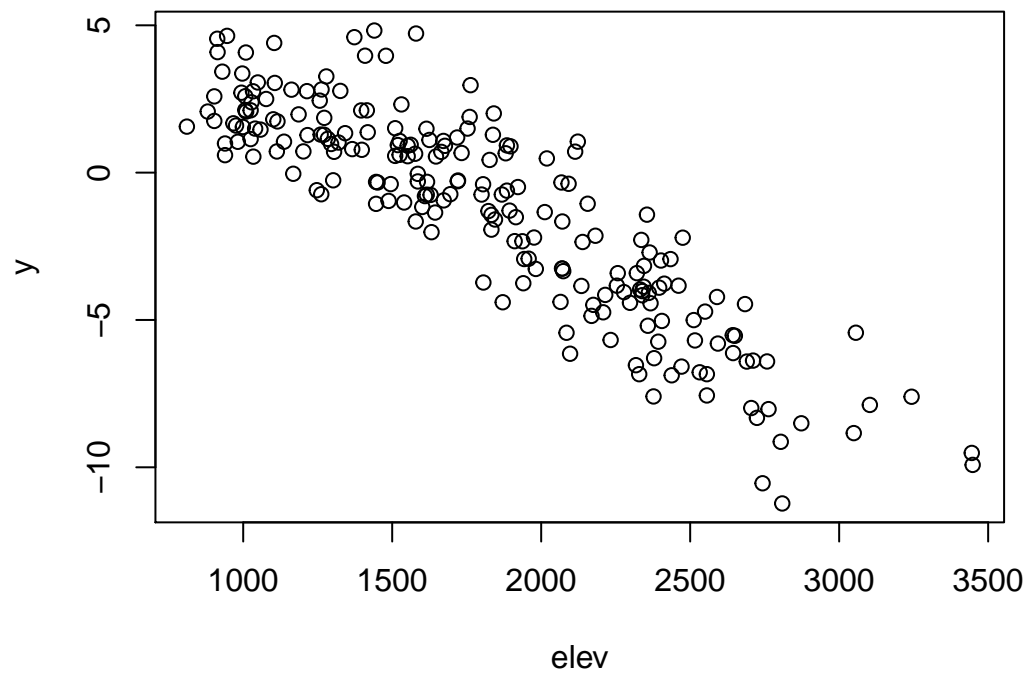
```
NGRID <- 50
# get elevations on a grid (will use these later)
COGrid<- fields.x.to.grid( x, nx=NGRID, ny=NGRID)
COGridPoints<- make.surface.grid( COGrid)
data( RMelevation)

COElev<- interp.surface( RMelevation, COGridPoints )
COElevGrid<- as.surface( COGridPoints, COElev)
```

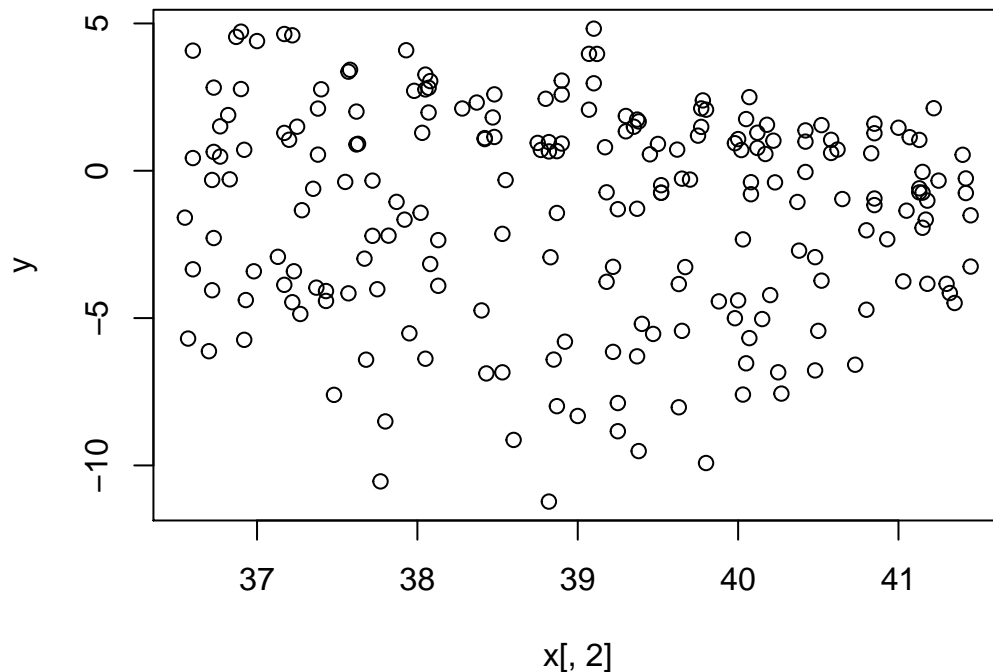
```
# take a look at the data
quilt.plot( x, y)
US( add=TRUE)
```



```
plot( elev, y)
```



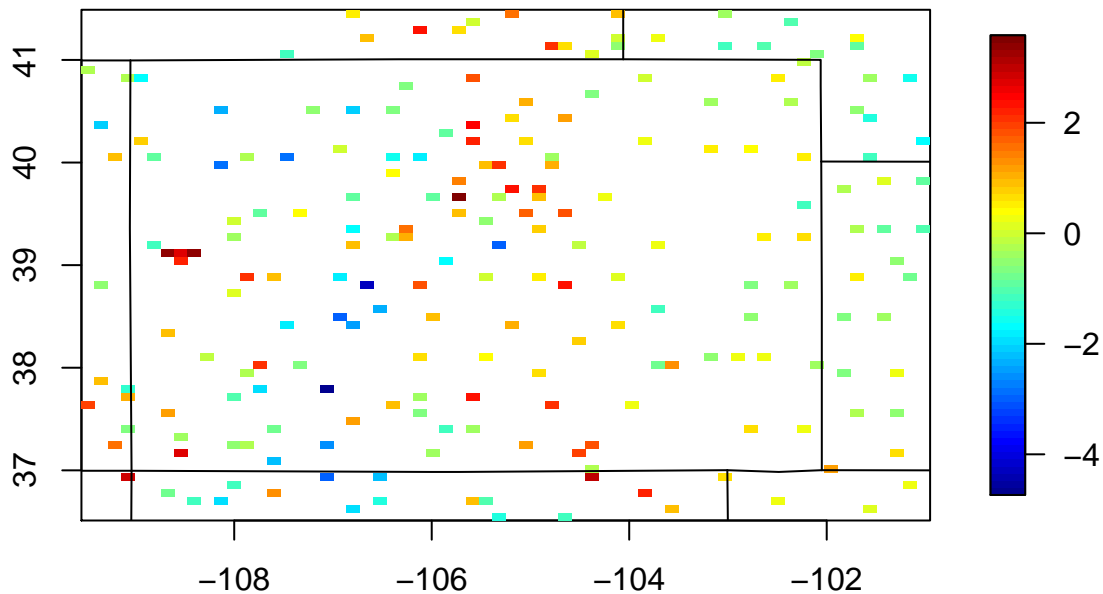
```
plot( x[,2], y)
```



```
X<- cbind( x, elev)
#
lmObj<- lm( y ~ lon+lat +elev, data=X )
summary( lmObj)

##
## Call:
## lm(formula = y ~ lon + lat + elev, data = X)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.6706 -0.8541 -0.1308  0.9088  3.5183
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  7.9372343   5.9584509   1.332   0.184
## lon         -0.1982295   0.0499149  -3.971 9.83e-05 ***
## lat         -0.4951096   0.0691068  -7.164 1.32e-11 ***
## elev        -0.0059590   0.0002029 -29.374 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.412 on 209 degrees of freedom
## Multiple R-squared:  0.8415, Adjusted R-squared:  0.8393
## F-statistic: 370 on 3 and 209 DF, p-value: < 2.2e-16

quilt.plot( x, lmObj$residuals)
US( add=TRUE)
```



```
fitOE<- Tps( x,y, Z=elev- mean(elev))
```

```
# fit a Kriging estimator Matern covariance smoothness =1.0
```

```
# range and nugget estimated by maximum likelihood
```

```
fit1<- spatialProcess( x,y)
```

```
# summary of the fit
```

```
print( fit1)
```

```
## CALL:
```

```
## spatialProcess(x = x, y = y)
```

```
##
```

```
## SUMMARY OF MODEL FIT:
```

```
##
```

```
## Number of Observations: 213
```

```
## Degree of polynomial in fixed part: 1
```

```
## Total number of parameters in fixed part: 3
```

```
## MLE nugget variance ( sigma^2) 1.55
```

```
## MLE process variance (rho) 7.591
```

```
## MLE range parameter (theta, units of distance): 0.8786
```

```
## Approx 95% CI for theta: [ 0.5774 , 1.639 ]
```

```
## Approx. degrees of freedom for curve 84.73
```

```
## Standard Error of df estimate: 2.661
```

```
## Nonzero entries in covariance 45369
```

```
##
```

```
## ESTIMATED COEFFICIENTS FOR FIXED PART:
```

```
##
```

```
## estimate SE pValue
```

```
## d1 55.0800 36.6800 0.1332
```

```
## d2 0.3434 0.3055 0.2610
```

```
## d3 -0.4924 0.4484 0.2722
```

```
##
```

```
## COVARIANCE MODEL: stationary.cov
```

```
## Covariance function: Matern
```

```
## Non-default covariance arguments and their values
```

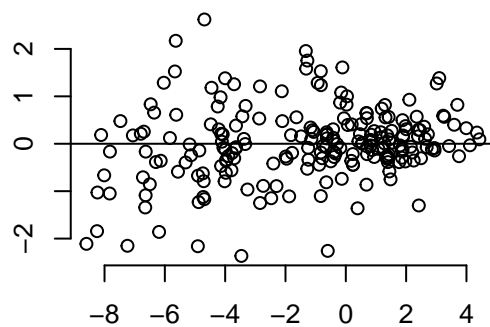
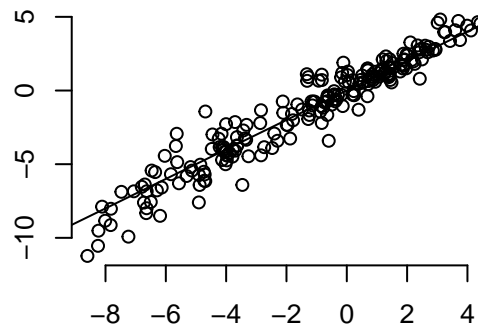
```
## Argument: Covariance has the value(s):
## [1] "Matern"
## Argument: smoothness has the value(s):
## [1] 1
## Argument: theta has the value(s):
## theta
## 0.8785521
## Argument: onlyUpper has the value(s):
## [1] FALSE
## Argument: distMat has the value(s):
## [1] NA
```

```
# diagnostic plots
set.panel(2,2)
```

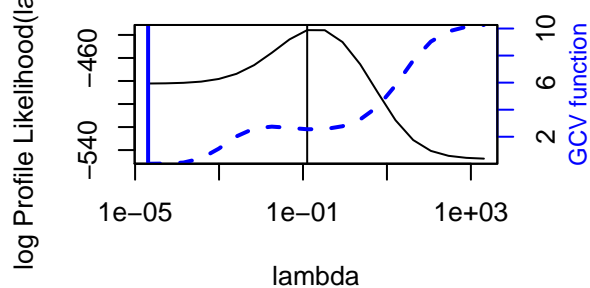
```
## plot window will lay out plots in a 2 by 2 matrix
```

```
plot( fit1)
```

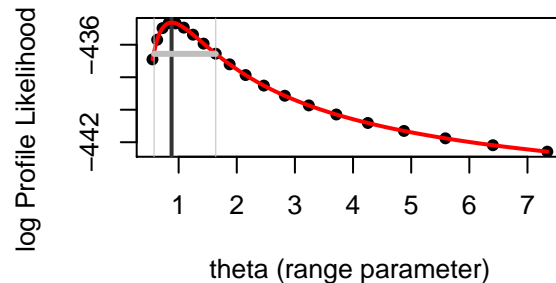
Observations by predicted values



**Profile likelihood over lambda
(with theta at MLE)**



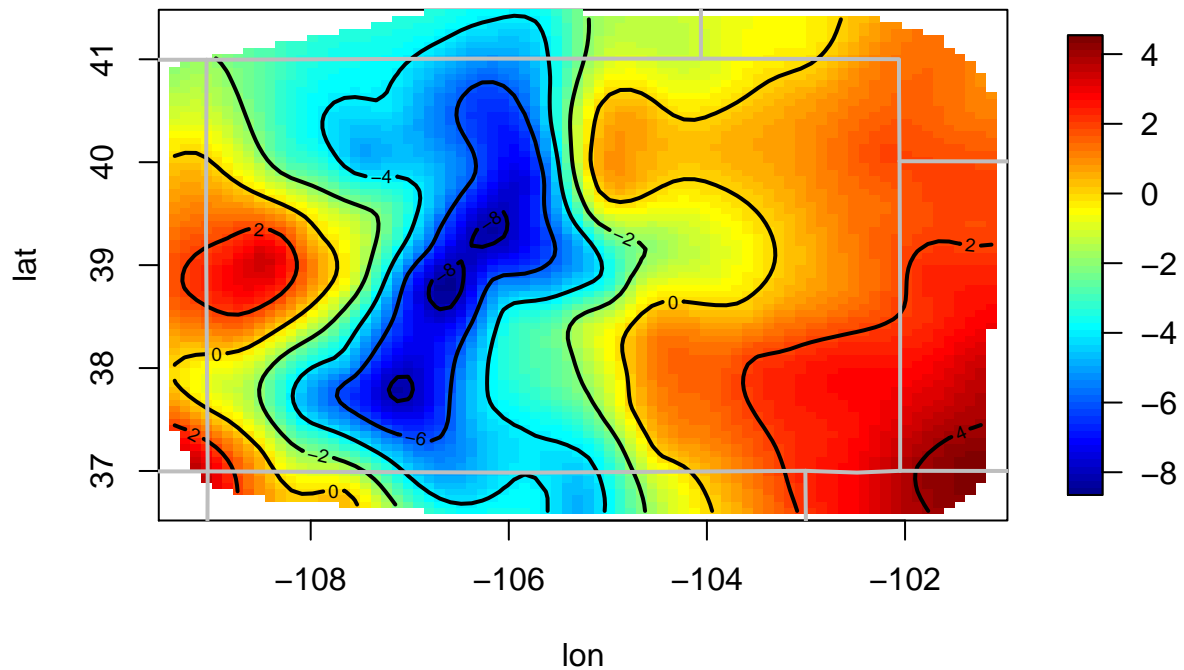
**Profile likelihood for theta
(range parameter)**



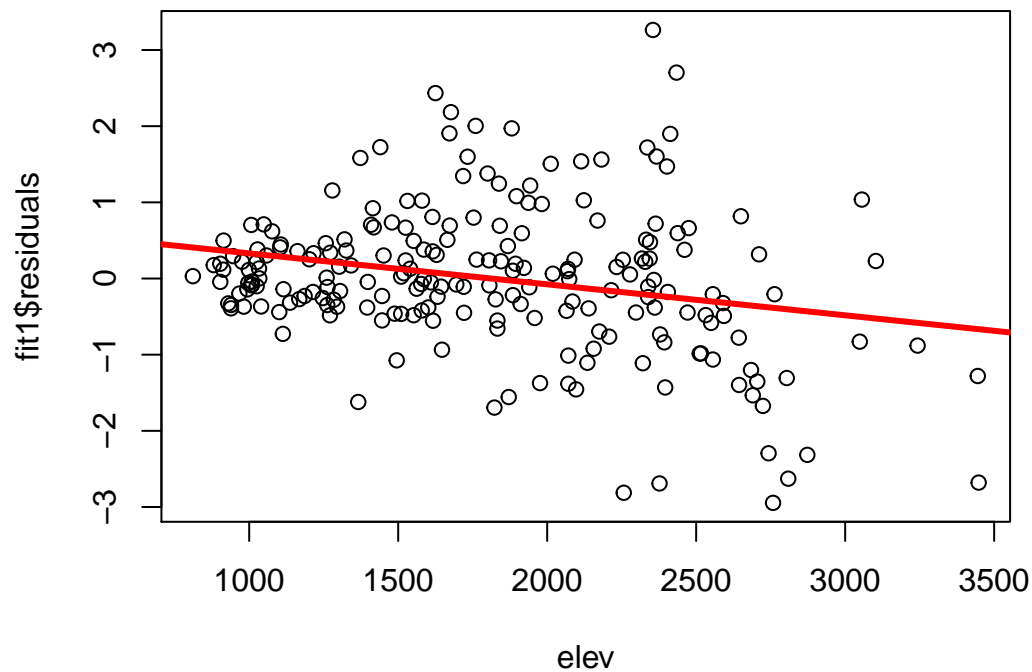
```
set.panel()
```

```
## plot window will lay out plots in a 1 by 1 matrix
```

```
surface( fit1)
US( add=TRUE, col="grey", lwd=2)
```

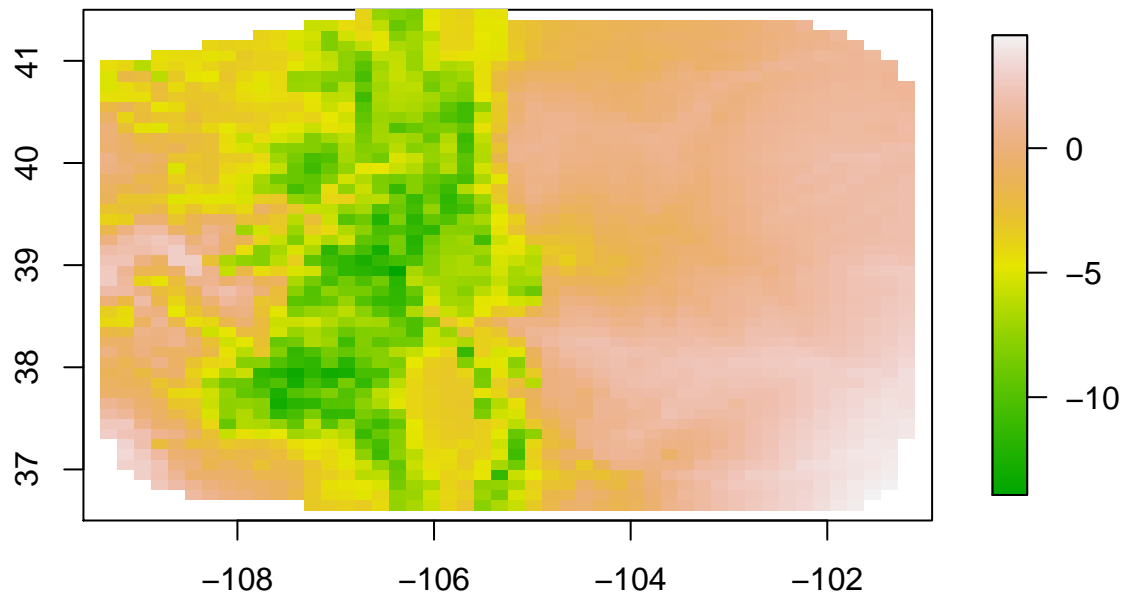


```
# take a look at residuals
plot( elev, fit1$residuals)
lmObj<- lm( fit1$residuals ~ elev)
abline( lmObj, col="red", lwd=3)
```

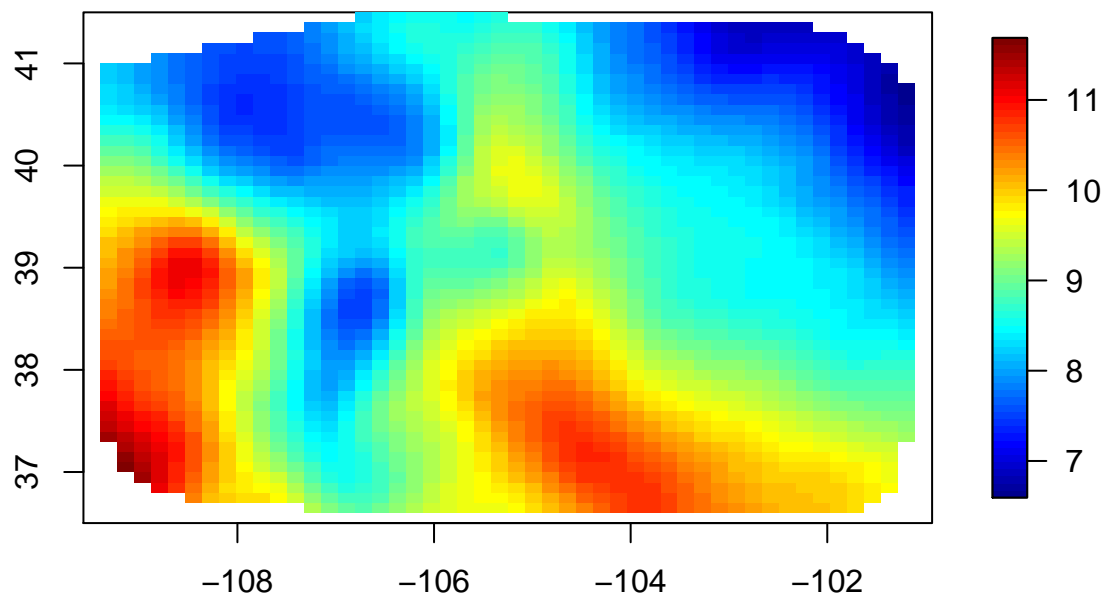


```
# with elevations
fit1E<- spatialProcess( x,y, Z = elev)

sur0<- predictSurface( fit1E, nx=NGRID, ny=NGRID, Z= COElevGrid)
image.plot( sur0, col=terrain.colors(256))
```



```
surOSmooth<- predictSurface( fit1E, nx=NGRID, ny=NGRID, drop.Z= TRUE)
image.plot( surOSmooth)
```



```
# uncertainty 40 draws from posterior distribution
set.seed(123)
# next command takes a minute or so
SEout<- sim.spatialProcess( fit1E, xp = COGridPoints,
                           Z = COElevGrid,
                           M = 40,
                           drop.Z = TRUE)

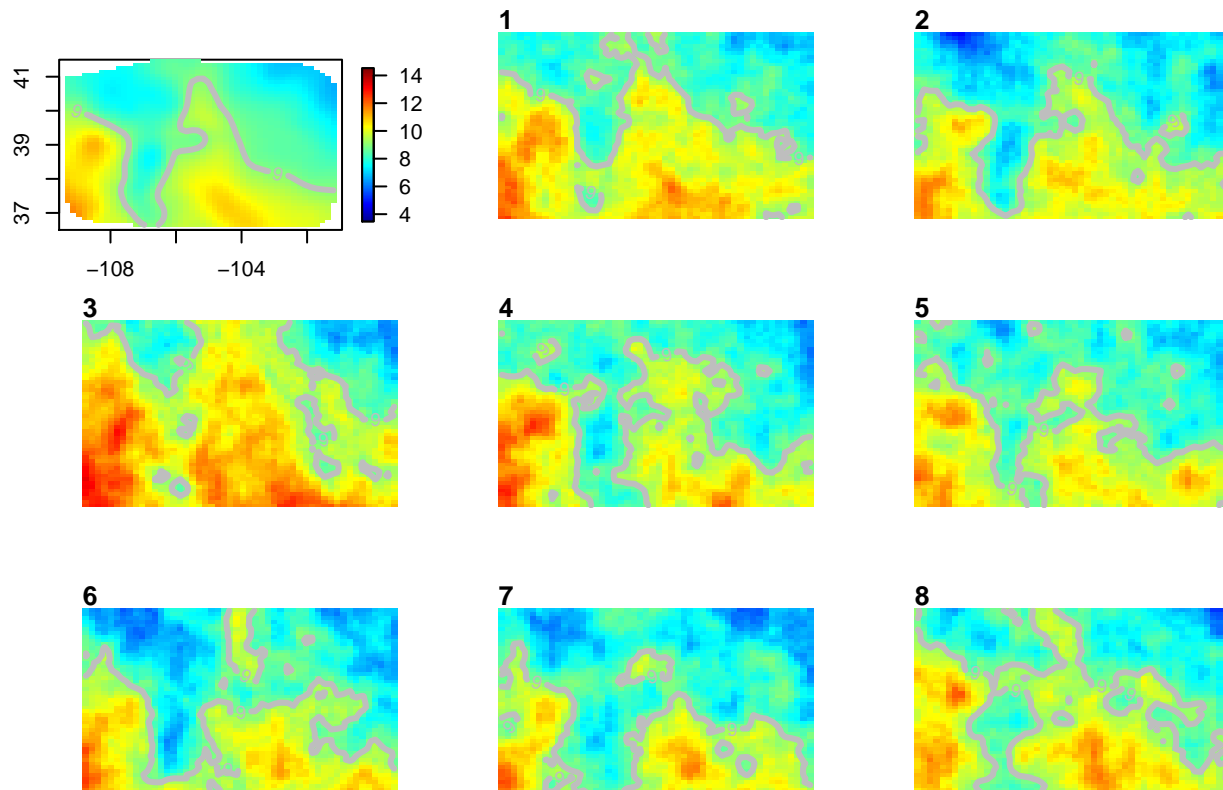
set.panel( 3,3)
```

```
## plot window will lay out plots in a 3 by 3 matrix
image.plot( surOSmooth,
            xlim=c(3.5,14.5), col=tim.colors(256))
```

```

contour( surOSmooth, levels= 9, lwd=3, col="grey", add=TRUE)
par( mar=c(3,3,1,1))
for( k in 1:8){
  image( as.surface( COGridPoints, SEout[k,]),
        zlim =c(3.5,14.5),col=tim.colors(256), axes=FALSE)
  contour( as.surface( COGridPoints, SEout[k,]),
          lwd=3, col="grey",level=9, add=TRUE)
  title( k, adj=0, cex=2)
}

```



```

set.panel()

## plot window will lay out plots in a 1 by 1 matrix
surSE<- apply( SEout, 2, sd )

image.plot( as.surface( COGridPoints, surSE))
points( x, col="magenta", pch=16)
contour( COElevGrid ,
        level= c(2000, 3000), add= TRUE, col="grey30", lwd=3)
US( add=TRUE, col="grey", lwd=2)

```