

Q1: Are lake and terrestrial primary productivity coherent?

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This document organizes for openness and reproducibility analyses of the temporal coherence of interannual variation in lake primary productivity with terrestrial primary productivity in the landscape surrounding the lake.

Data import

Data produced in ‘ms1_prep.Rmd’ are loaded.

```
load("/Users/jonathanwalter/Box Sync/NSF EAGER Synchrony/Data/RData files/ms1_analysis_inprogress1.RData")

any(sapply(analysislakes$lakedata, function(x){any(is.infinite(x))}))

## [1] FALSE

any(sapply(analysislakes$lakedata, function(x){any(is.na(x))}))

## [1] FALSE

which(sapply(analysislakes$lakedata, function(x){any(is.na(x))}))

## named integer(0)

analysislakes$lakeinfo[which(sapply(analysislakes$lakedata, function(x){any(is.na(x))})),]

## [1] lagoslakeid      gnis_name      nhd_lat
## [4] nhd_long      lake_area_ha    lake_perim_meters
## [7] nhd_ftype      nhd_fcode      hu4_zoneid
## [10] hu12_zoneid    state_zoneid    elevation_m
## [13] start          end
## <0 rows> (or 0-length row.names)

# image(accndvi)
# points(lakepts.prj[which(sapply(analysislakes$lakedata, function(x){any(is.na(x))})),])

dbuff[which(sapply(analysislakes$lakedata, function(x){any(is.na(x))}))]

## numeric(0)

analysislakes$lakeinfo<-analysislakes$lakeinfo[!sapply(analysislakes$lakedata, function(x){any(is.na(x))}),]
analysislakes$lakedata<-analysislakes$lakedata[!sapply(analysislakes$lakedata, function(x){any(is.na(x))}),]

analysislakes$lakeinfo$tslength<-analysislakes$lakeinfo$end-analysislakes$lakeinfo$start+1
# analysislakes$lakedata<-analysislakes$lakedata[!analysislakes$lakeinfo$tslength < 20]
# analysislakes$lakeinfo<-analysislakes$lakeinfo[!analysislakes$lakeinfo$tslength < 20,]

source("~/GitHub/AquaTerrSynch/AnalysisCode/bandtest_coh.R")

tsranges<-rbind(c(2,4),c(4,Inf),c(2,Inf))
```

```

coh.chlaXaccndvi<-NULL
coh.chlaXmaxndvi<-NULL

for(lind in 1:length(analysislakes$lakedata)){
  lakedat.ii<-cleandat(analysislakes$lakedata[[lind]], as.numeric(colnames(analysislakes$lakedata[[lind]])))
  chlaXaccndvi<-coh(lakedat.ii[1,], lakedat.ii[2,], as.numeric(colnames(analysislakes$lakedata[[lind]])))
    norm="powall", sigmethod="fast", nrand=10000)
  chlaXmaxndvi<-coh(lakedat.ii[1,], lakedat.ii[3,], as.numeric(colnames(analysislakes$lakedata[[lind]])))
    norm="powall", sigmethod="fast", nrand=10000)
  for(rind in 1:nrow(tsranges)){
    chlaXaccndvi<-bandtest.coh(chlaXaccndvi, tsranges[rind,])
    chlaXmaxndvi<-bandtest.coh(chlaXmaxndvi, tsranges[rind,])
  }
  coh.chlaXaccndvi<-rbind(coh.chlaXaccndvi, c(t(as.matrix(chlaXaccndvi$bandp[,3:5]))))
  coh.chlaXmaxndvi<-rbind(coh.chlaXmaxndvi, c(t(as.matrix(chlaXmaxndvi$bandp[,3:5]))))
}

coh.chlaXaccndvi<-as.data.frame(coh.chlaXaccndvi)
coh.chlaXmaxndvi<-as.data.frame(coh.chlaXmaxndvi)

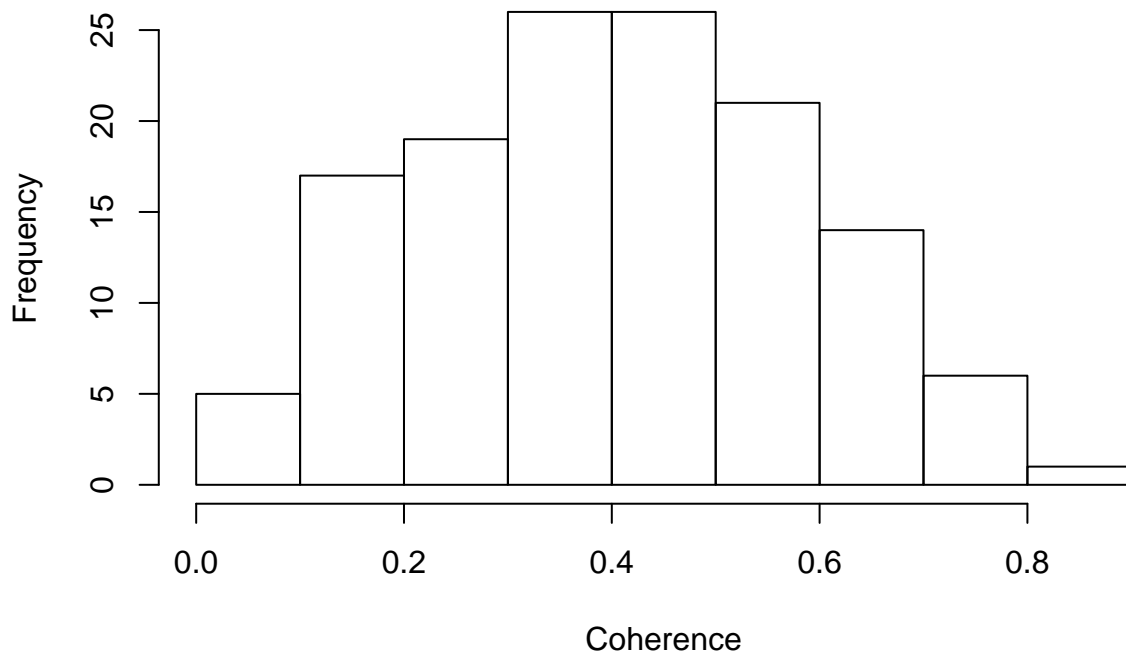
colnames(coh.chlaXaccndvi)<-paste0("accndvi",c("p.ts1","phi.ts1","coh.ts1","p.ts2","phi.ts2","coh.ts2",
colnames(coh.chlaXmaxndvi)<-paste0("maxndvi",c("p.ts1","phi.ts1","coh.ts1","p.ts2","phi.ts2","coh.ts2",

coh.chlaXaccndvi$lagoslakeid<-analysislakes$lakeinfo$lagoslakeid
coh.chlaXmaxndvi$lagoslakeid<-analysislakes$lakeinfo$lagoslakeid

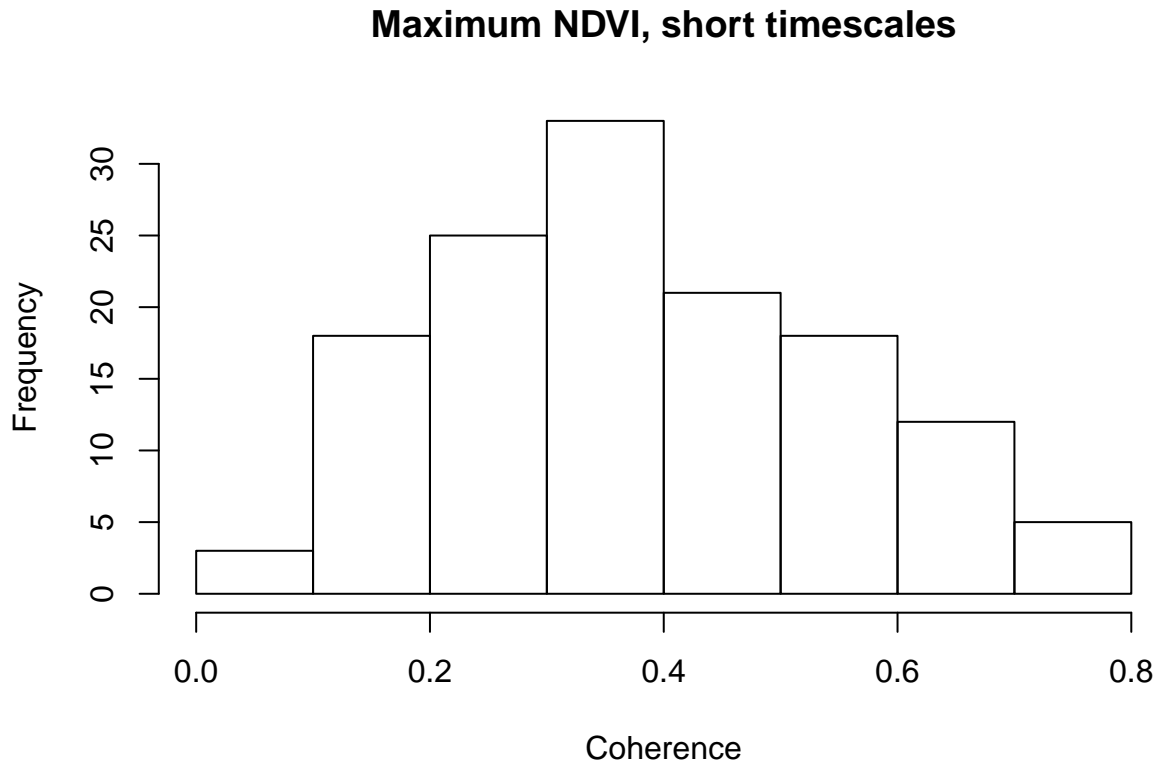
#short timescales
hist(coh.chlaXaccndvi$accndvicoh.ts1, main="Accumulated NDVI, short timescales", xlab="Coherence", ylab="Frequency")

```

Accumulated NDVI, short timescales



```
hist(coh.chlaXmaxndvi$maxndvicoh.ts1, main="Maximum NDVI, short timescales", xlab="Coherence", ylab="Fr
```



```
quantile(coh.chlaXaccndvi$accndvicoh.ts1)
```

```
##          0%          25%          50%          75%         100%
## 0.03540956 0.26015941 0.40373548 0.52492077 0.81625251
```

```
quantile(coh.chlaXmaxndvi$maxndvicoh.ts1)
```

```
##          0%          25%          50%          75%         100%
## 0.04514692 0.24996954 0.35281892 0.50311715 0.77145899
```

```
alpha=0.05
```

```
sum(coh.chlaXaccndvi$accndvip.ts1<alpha)/nrow(coh.chlaXaccndvi)
```

```
## [1] 0.06666667
```

```
sum(coh.chlaXmaxndvi$maxndvip.ts1<alpha)/nrow(coh.chlaXmaxndvi)
```

```
## [1] 0.05925926
```

```
print(coh.chlaXaccndvi$accndviphi.ts1[coh.chlaXaccndvi$accndvip.ts1<alpha]/pi) #only pattern is that la
```

```
## [1] 0.33224850 -0.97156054 -0.04413595 0.56356061 -0.86709075 -0.05260276
## [7] 0.12416199 -0.04172693 0.92429361
```

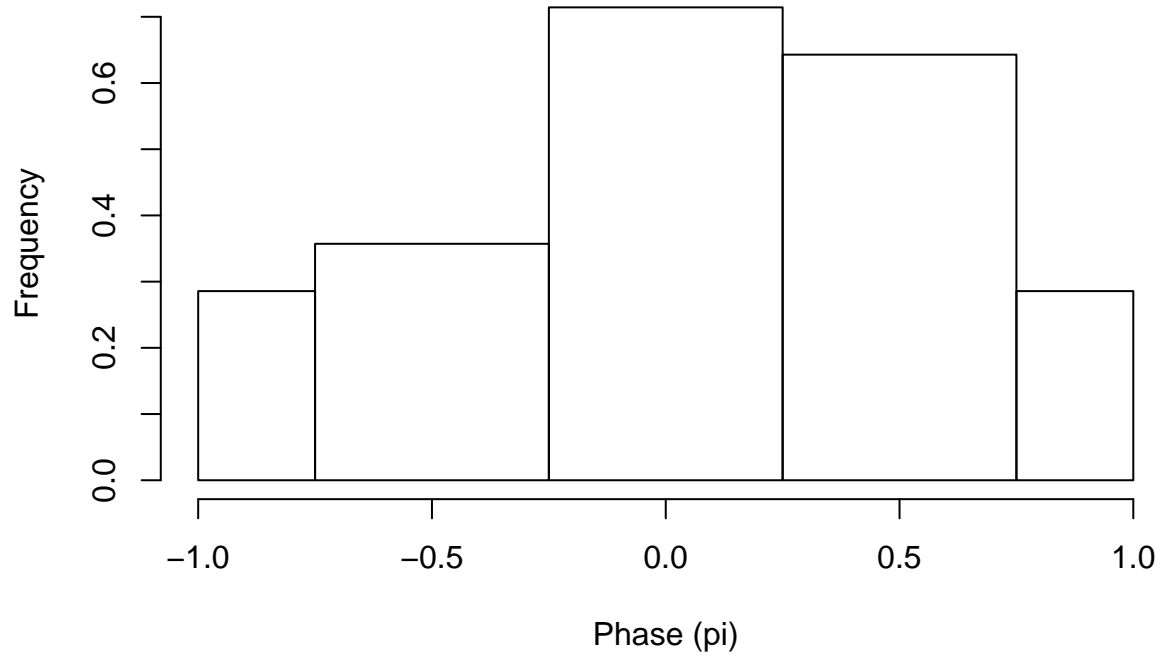
```
print(coh.chlaXmaxndvi$maxndviphi.ts1[coh.chlaXmaxndvi$maxndvip.ts1<alpha]/pi)
```

```
## [1] -0.1573764 -0.8240104 -0.7892870 -0.7185325 -0.9310910 -0.8435071
## [7] -0.2280369 0.5324496
```

```
phicls<-c(-1,-.75,-0.25,0.25,0.75,1)
```

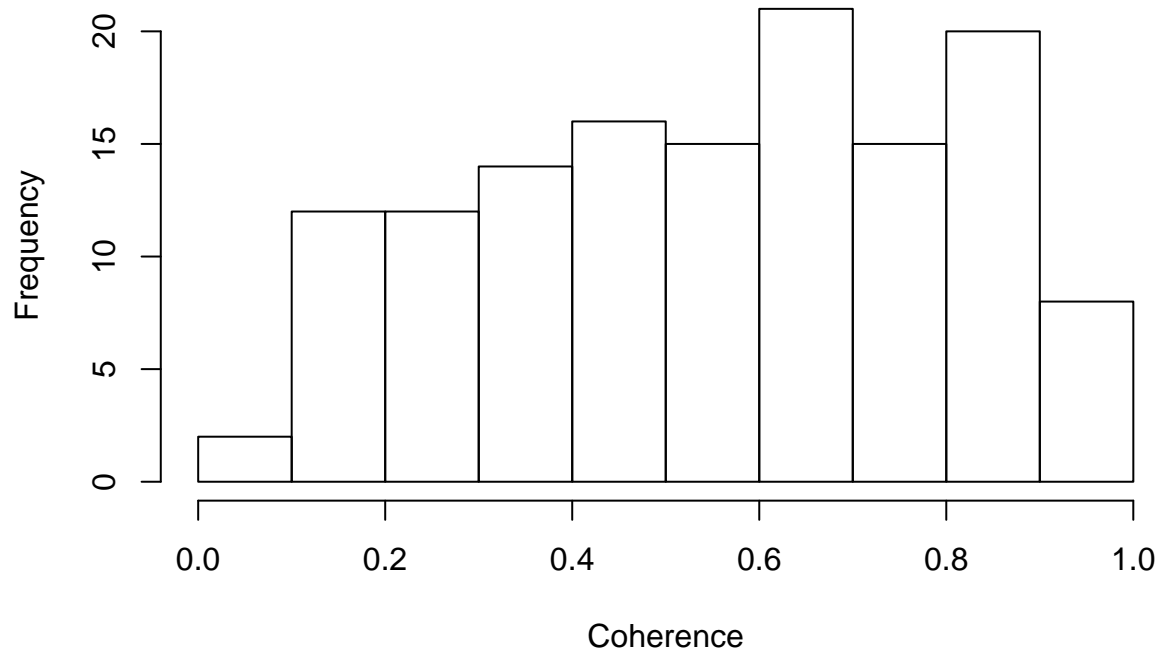
```
hist(coh.chlaXaccndvi$accndviphi.ts1[coh.chlaXaccndvi$accndvip.ts1<0.2]/pi, main="Accumulated NDVI, sho
```

Accumulated NDVI, short timescales

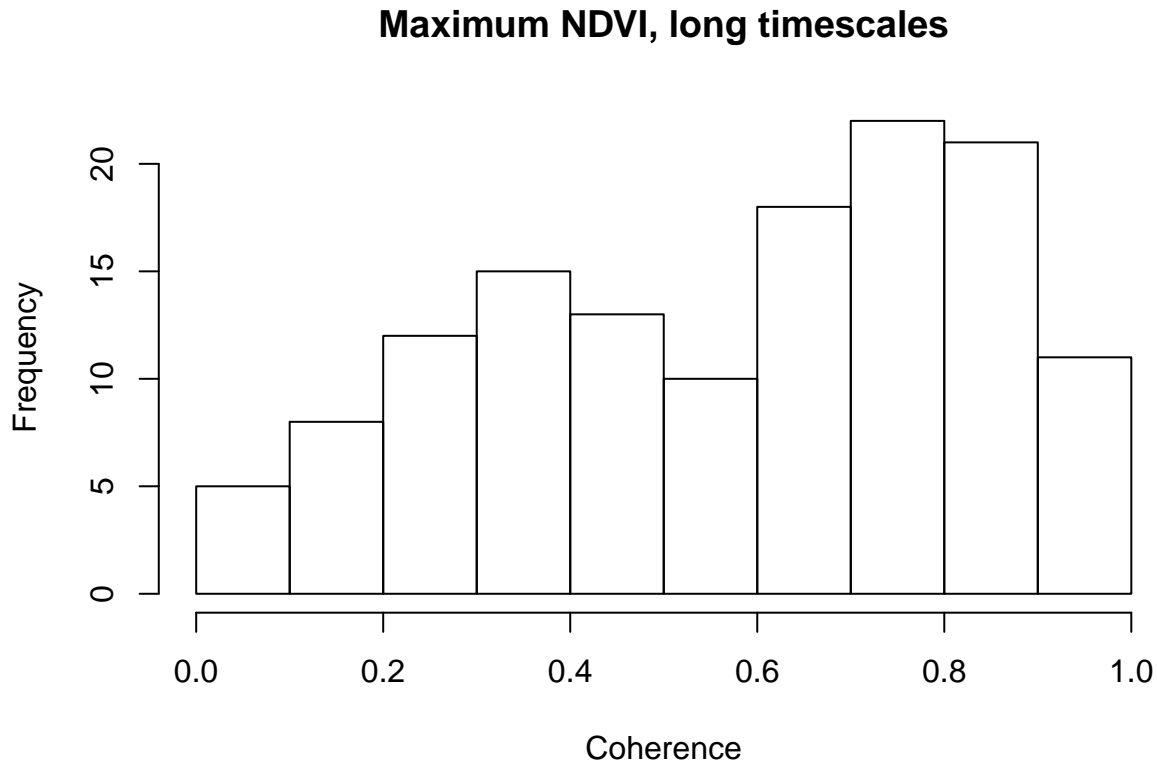


```
#hist(coh.chlaXmaxndvi$maxndviphi.ts1[coh.chlaXmaxndvi$maxndvip.ts1<0.2]/pi, main="Maximum NDVI, short  
#long timescales  
hist(coh.chlaXaccndvi$accndvicoh.ts2, main="Accumulated NDVI, long timescales", xlab="Coherence", ylab=
```

Accumulated NDVI, long timescales



```
hist(coh.chlaXmaxndvi$maxndvicoh.ts2, main="Maximum NDVI, long timescales", xlab="Coherence", ylab="Frequency")
```



```
quantile(coh.chlaXaccndvi$accndvicoh.ts2)
```

```
##          0%          25%          50%          75%         100%
## 0.06700155 0.35635453 0.56072757 0.75753276 0.96052338
```

```
quantile(coh.chlaXmaxndvi$maxndvicoh.ts2)
```

```
##          0%          25%          50%          75%         100%
## 0.04123391 0.35832298 0.61507443 0.78760333 0.96402244
```

```
alpha=0.05
```

```
sum(coh.chlaXaccndvi$accndvip.ts2<alpha)/nrow(coh.chlaXaccndvi)
```

```
## [1] 0.05185185
```

```
sum(coh.chlaXmaxndvi$maxndvip.ts2<alpha)/nrow(coh.chlaXmaxndvi)
```

```
## [1] 0.05925926
```

```
print(coh.chlaXaccndvi$accndviphi.ts2[coh.chlaXaccndvi$accndvip.ts2<alpha]/pi)
```

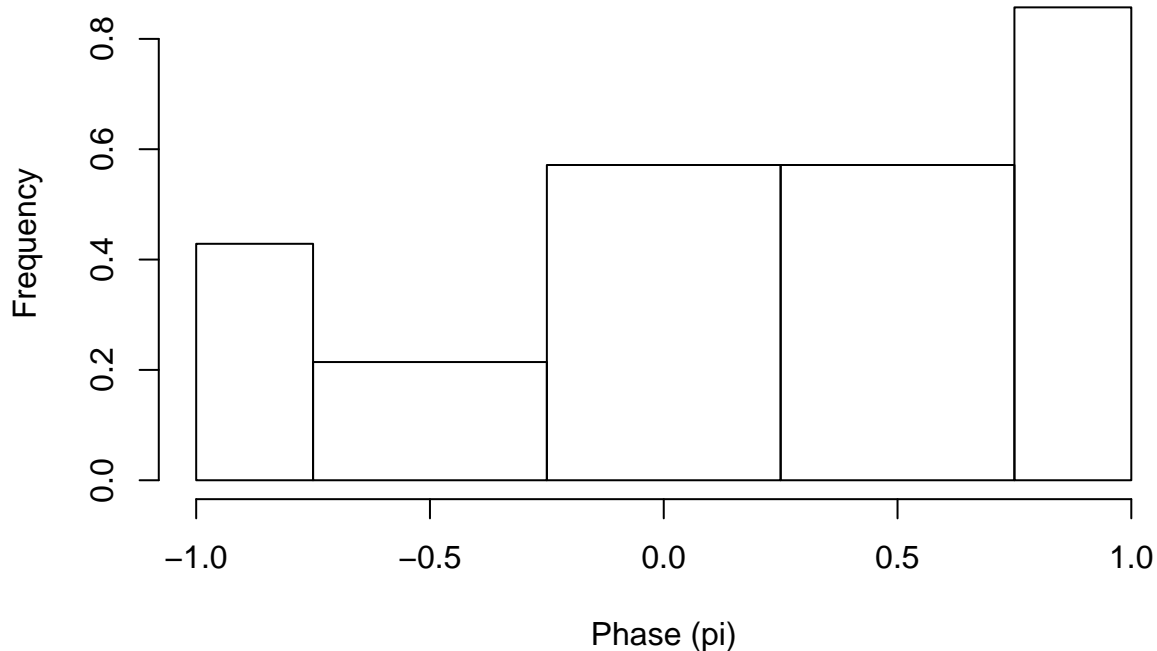
```
## [1] -0.43893809 0.25315167 -0.40935196 -0.04386325 -0.65597599 0.36382168
## [7] 0.89471121
```

```
print(coh.chlaXmaxndvi$maxndviphi.ts2[coh.chlaXmaxndvi$maxndvip.ts2<alpha]/pi)
```

```
## [1] 0.69982097 -0.97179292 -0.04190360 0.02097044 -0.67004320 -0.58501674
## [7] -0.31373024 -0.33804686
```

```
hist(coh.chlaXaccndvi$accndviphi.ts1[coh.chlaXaccndvi$accndvip.ts2<0.2]/pi, main="Accumulated NDVI, long timescales", xlab="Coherence", ylab="Frequency")
```

Accumulated NDVI, long timescales



```
#hist(coh.chlaXmaxndvi$maxndviphi.ts1[coh.chlaXmaxndvi$maxndvicoh.ts2>0.6]/pi, main="Maximum NDVI, short timescales")
```

```
states<-readOGR("~/Box Sync/NSF EAGER Synchrony/Data/statesp020.shp")
```

```
## OGR data source with driver: ESRI Shapefile
```

```
## Source: "/Users/jonathanwalter/Box Sync/NSF EAGER Synchrony/Data/statesp020.shp", layer: "statesp020"
```

```
## with 2895 features
```

```
## It has 9 fields
```

```
## Integer64 fields read as strings: STATESP020 DAY_ADM YEAR_ADM
```

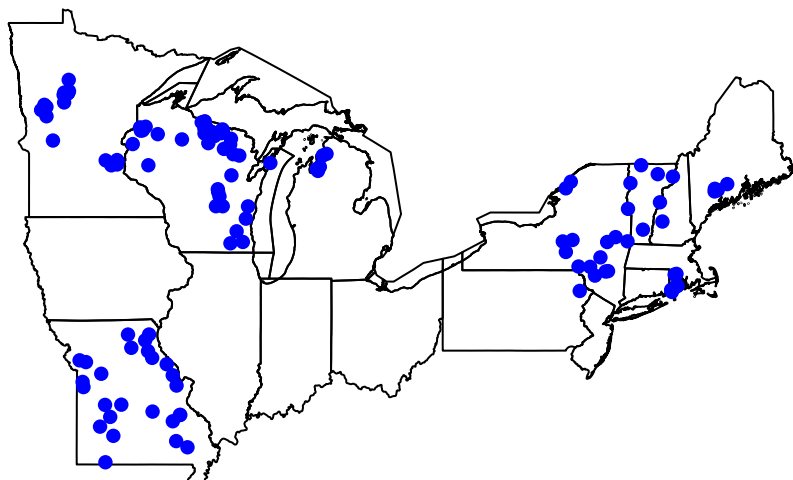
```
getstates<-c("Minnesota", "Iowa", "Wisconsin", "Illinois", "Missouri", "Michigan", "Indiana", "Ohio", "Louisiana", "Alabama", "Georgia", "Florida", "South Carolina", "North Carolina", "Virginia", "Maryland", "Delaware", "Pennsylvania", "New York", "New Jersey", "Connecticut", "Rhode Island", "Massachusetts", "Vermont", "New Hampshire", "Maine", "Hawaii")
```

```
lagosstates<-states[states@data$STATE %in% getstates,]
```

```
plot(lagosstates, main="Lakes selected for analysis")
```

```
points(analysislakes$lakeinfo$nhd_long, analysislakes$lakeinfo$nhd_lat, pch=16, cex=1, col="blue")
```

Lakes selected for analysis



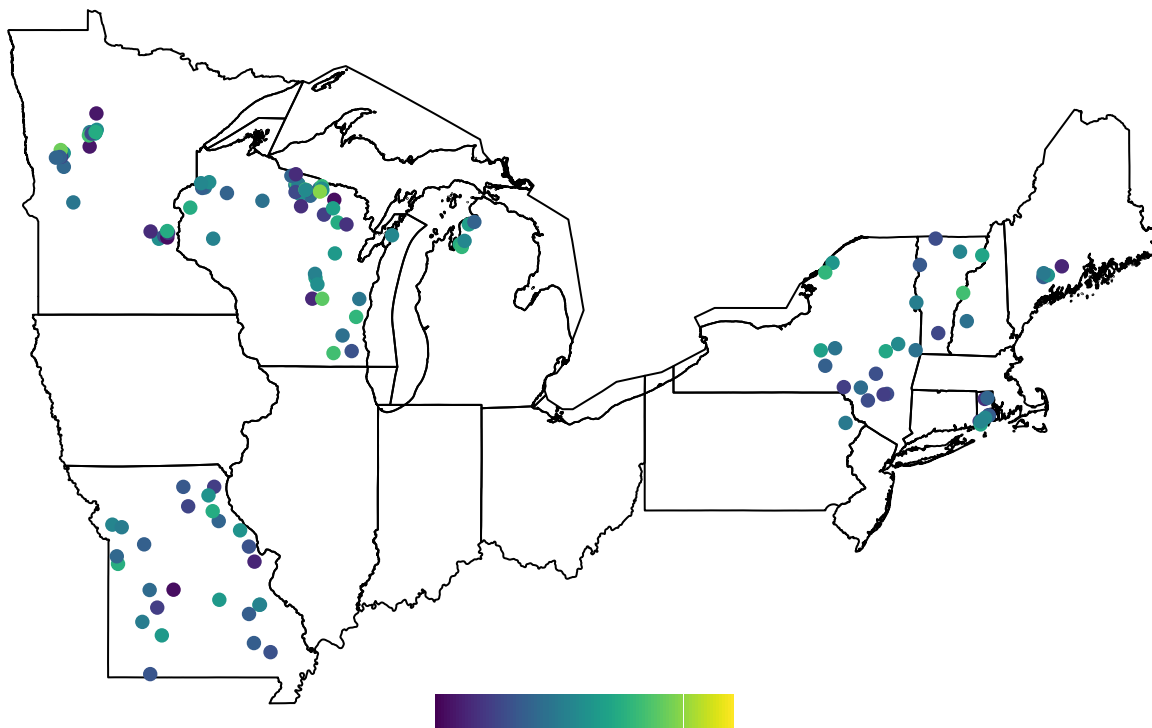
```
cohplotdata<-left_join(analysislakes$lakeinfo, coh.chlaXaccndvi, by="lagoslakeid")

pal<-viridis(100)

par(mar=c(1,0,2,0))

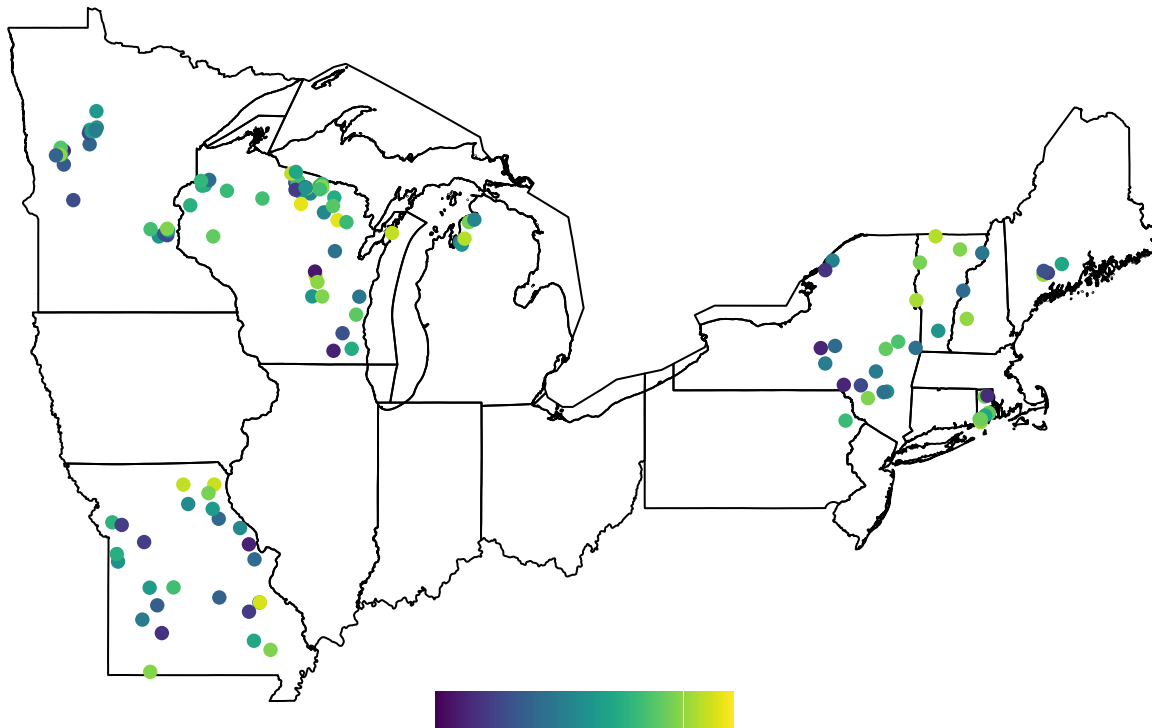
plot(lagosstates, main="Lakes by short timescale coherence")
points(cohplotdata$nhd_long, cohplotdata$nhd_lat, pch=16, cex=1, col=pal[round(cohplotdata$accndvicoh.t.
colorbar.plot(x=mean(par("usr")[1:2]),y=par("usr")[3],strip=1:100,col=pal,horizontal = T)
```

Lakes by short timescale coherence



```
plot(lagosstates, main="Lakes by long timescale coherence")
points(cohplotdata$nhd_long, cohplotdata$nhd_lat, pch=16, cex=1, col=pal[round(cohplotdata$accndvicoh.t,
colorbar.plot(x=mean(par("usr")[1:2]),y=par("usr")[3],strip=1:100,col=pal,horizontal = T)
```

Lakes by long timescale coherence



```
#Need to add: depth, average growing season Chlorophyll-a, TSI(chla) categories, pct ag

#agriculture -- is 500m buffer best? Other options include 100m buffer (probably too small) and hu12 wa
# pct.ag<-lagosne_select(table="buffer500m.lulc", vars=c("lagoslakeid","buffer500m_nlcd2001_pct_82","bu
pct.ag<-lagosne_select(table="hu12.lulc", vars=c("hu12_zoneid","hu12_nlcd2001_pct_82","hu12_nlcd2006_pc
pct.ag<-pct.ag[pct.ag$hu12_zoneid %in% analysislakes$lakeinfo$hu12_zoneid,]
pct.ag.avg<-data.frame(hu12_zoneid=pct.ag$hu12_zoneid, pct.ag=rowMeans(pct.ag[,2:4]))

#depth
depth<-lagosne_select(table="lakes_limno", vars=c("lagoslakeid","maxdepth"))
depth<-depth[depth$lagoslakeid %in% analysislakes$lakeinfo$lagoslakeid,] #use max depth because it's mo

#growing season Chlorophyll-a
chla<-lagosne_select(table="epi_nutr", vars=c("lagoslakeid","samplemonth","chla"))
chla<-chla[chla$lagoslakeid %in% analysislakes$lakeinfo$lagoslakeid,]
gs.chla<-chla[chla$samplemonth %in% 5:9,]
avg.chla<-aggregate(chla ~ lagoslakeid, data=gs.chla, FUN=mean, na.rm=T)

#Chlorophyll-a TSI class
#TSI(CHL) = 9.81 ln(CHL) + 30.6
tsi.chl<-data.frame(lagoslakeid=avg.chla$lagoslakeid, tsi=9.81 * log(avg.chla$chla) + 30.6)
tsi.chl$tsi.cat<-rep("lake",nrow(tsi.chl))

tsi.chl$tsi.cat[tsi.chl$tsi < 40]<-"oligotrophic"
```



```

tsi.chl$tsi.cat[tsi.chl$tsi >=40 & tsi.chl$tsi < 50]<-"mesotrophic"
tsi.chl$tsi.cat[tsi.chl$tsi >=50 & tsi.chl$tsi < 70]<-"eutrophic"
tsi.chl$tsi.cat[tsi.chl$tsi >= 70] <-"hypereutrophic"

#CV of terrestrial NDVI
cv.accndvi<-NULL
for(lake in 1:length(analysislakes$lakedata)){
  tmp<-analysislakes$lakedata[[lake]][rownames(analysislakes$lakedata[[lake]])=="accndvi",]
  cv.accndvi<-c(cv.accndvi, sd(tmp)/mean(tmp))
  # rm(tmp)
}
cv.accndvi<-data.frame(lagoslakeid=as.numeric(names(analysislakes$lakedata)), cv.accndvi=cv.accndvi)

#huc2 and huc4 watershed codes
huc_codes<-read.csv("/Users/jonathanwalter/GitHub/AquaTerrSynch/AnalysisCode/match_huc_codes.csv", colC

#state info
states<-lagosne_select(table="state", vars=c("state_zoneid","state_name"))

predictors<-analysislakes$lakeinfo
predictors<-left_join(predictors, depth, by="lagoslakeid")
predictors<-left_join(predictors, pct.ag.avg, by="hu12_zoneid")

## Warning: Column `hu12_zoneid` joining factors with different levels,
## coercing to character vector
predictors<-left_join(predictors, avg.chla, by="lagoslakeid")
predictors<-left_join(predictors, tsi.chl, by="lagoslakeid")
predictors<-left_join(predictors, states, by="state_zoneid")

## Warning: Column `state_zoneid` joining factors with different levels,
## coercing to character vector
predictors<-left_join(predictors, cv.accndvi, by="lagoslakeid")
#predictors<-left_join(predictors, huc_codes, by="hu4_zoneid")

for(nn in 1:ncol(predictors)){
  if(is.factor(predictors[,nn])){
    predictors[,nn]<-factor(predictors[,nn])
  }
}

str(predictors)

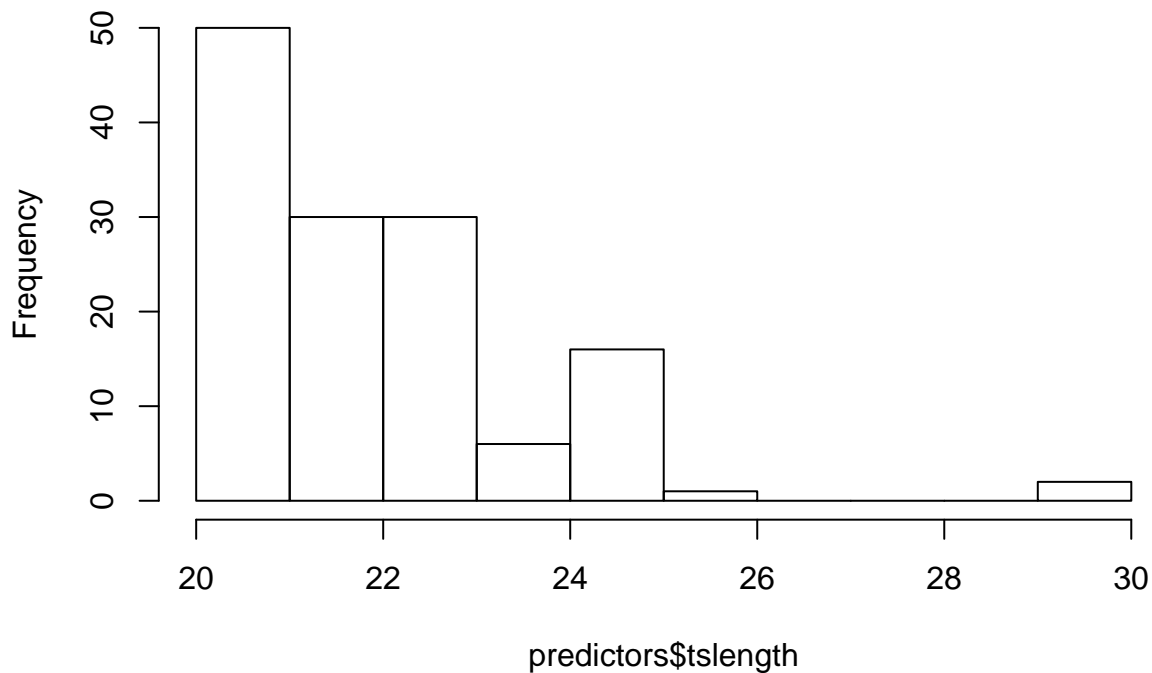
## 'data.frame': 135 obs. of 22 variables:
## $ lagoslakeid : num 211 249 618 906 969 ...
## $ gnis_name : chr NA NA "Butternut Lake" "Sparkling Lake" ...
## $ nhd_lat : num 44.5 43.7 45.9 46 45.8 ...
## $ nhd_long : num -73.3 -73.4 -89 -89.7 -89.3 ...
## $ lake_area_ha : num 113496.4 30 504.7 63.7 210.2 ...
## $ lake_perim_meters: num 1042251 3494 13134 3777 9402 ...
## $ nhd_ftype : int 390 390 390 390 390 390 390 390 390 ...
## $ nhd_fcode : int 39004 39004 39004 39004 39004 39004 39004 39004 39004 ...

```

```
## $ hu4_zoneid      : Factor w/ 28 levels "HU4_10","HU4_12",...: 17 17 11 8 12 10 10 10 10 10 ...
## $ hu12_zoneid     : chr  "HU12_17646" "HU12_16835" "HU12_13309" "HU12_13098" ...
## $ state_zoneid    : chr  "State_17" "State_5" "State_9" "State_9" ...
## $ elevation_m     : num  28.8 28.2 514.5 494.7 503.3 ...
## $ start           : num  1989 1990 1993 1989 1994 ...
## $ end             : num  2010 2010 2013 2011 2013 ...
## $ tslength        : num  22 21 21 23 20 21 21 21 21 22 ...
## $ maxdepth        : num  97 NA 12.8 20 11.6 ...
## $ pct.ag          : num  2.5298 0.4199 0.0976 0.3029 6.6886 ...
## $ chla            : num  5.39 7.94 2.44 1.86 2.04 ...
## $ tsi             : num  47.1 50.9 39.4 36.7 37.6 ...
## $ tsi.cat         : chr  "mesotrophic" "eutrophic" "oligotrophic" "oligotrophic" ...
## $ state_name      : Factor w/ 10 levels "Maine","Michigan",...: 9 6 10 10 10 2 2 2 2 2 ...
## $ cv.accndvi      : num  0.0572 0.0542 0.0443 0.0561 0.0417 ...
```

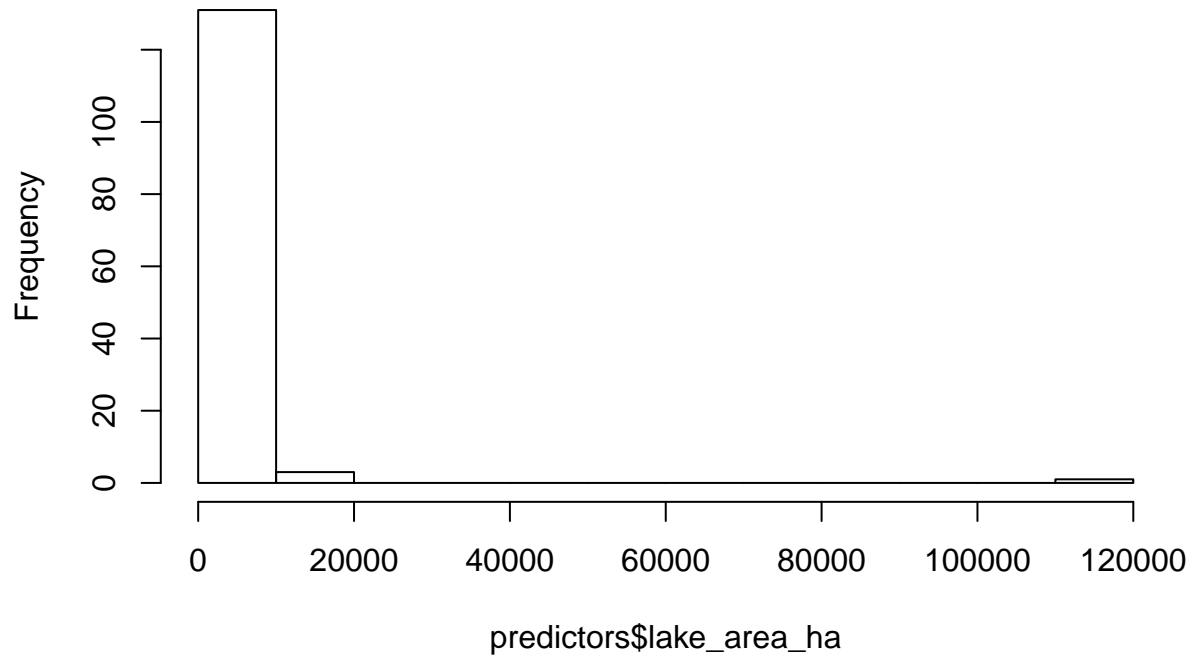
```
hist(predictors$tslength)
```

Histogram of predictors\$tslength



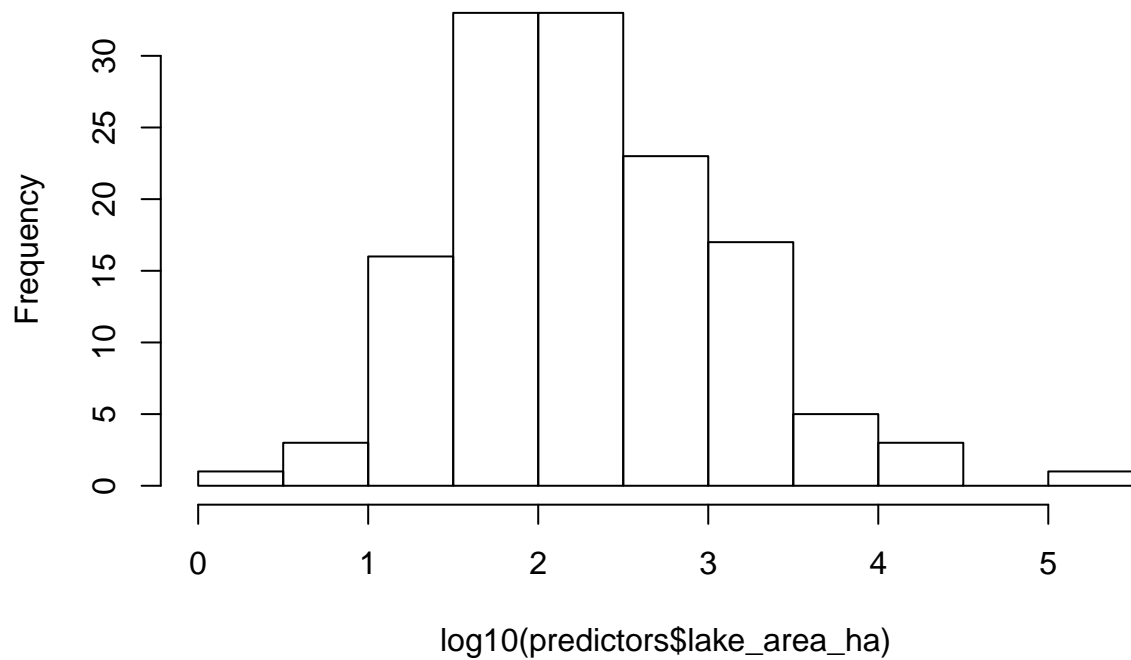
```
hist(predictors$lake_area_ha)
```

Histogram of predictors\$lake_area_ha



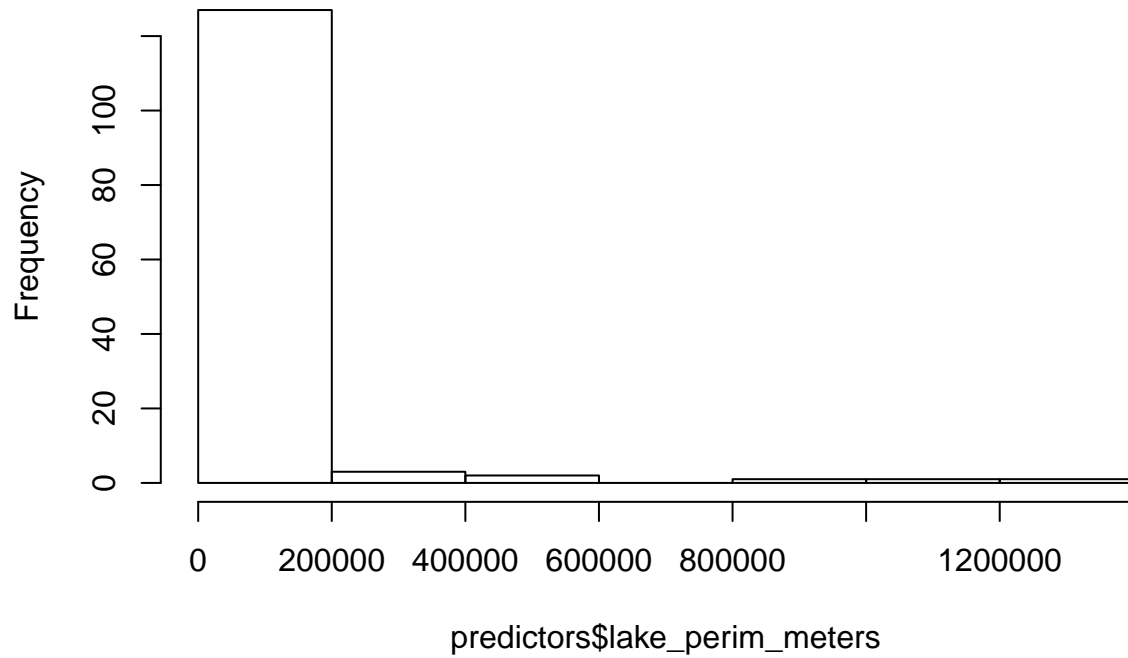
```
hist(log10(predictors$lake_area_ha))
```

Histogram of log10(predictors\$lake_area_ha)



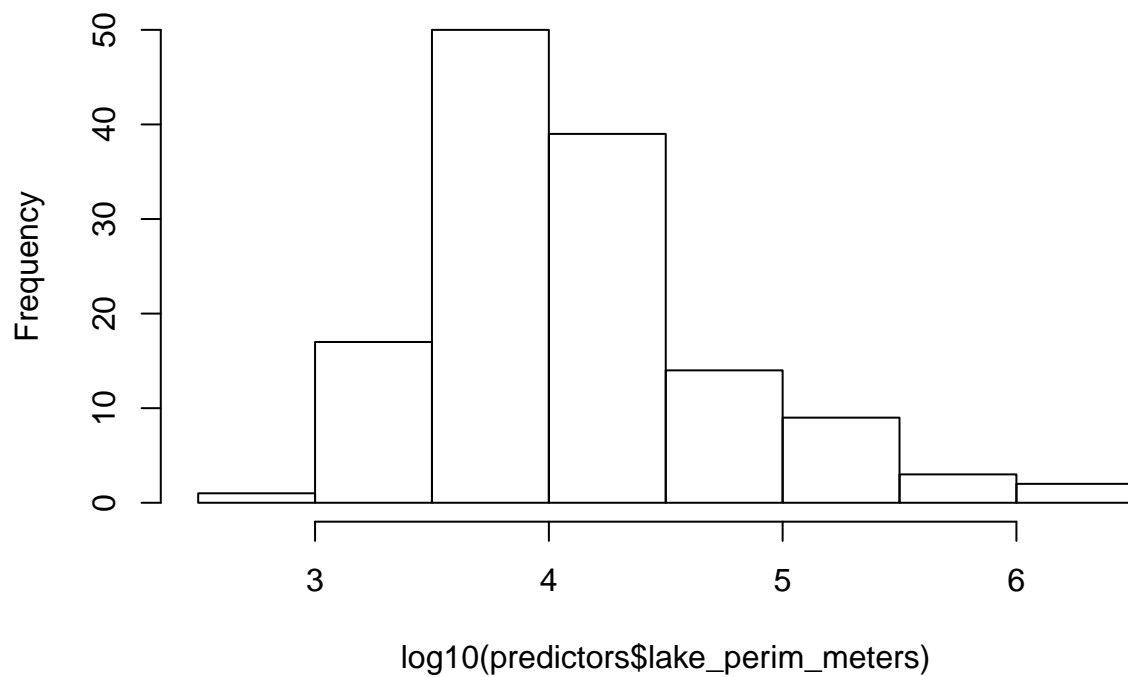
```
hist(predictors$lake_perim_meters)
```

Histogram of predictors\$lake_perim_meters



```
hist(log10(predictors$lake_perim_meters))
```

Histogram of log10(predictors\$lake_perim_meters)



```
table(predictors$nhd_fcode)
```

```
##
```

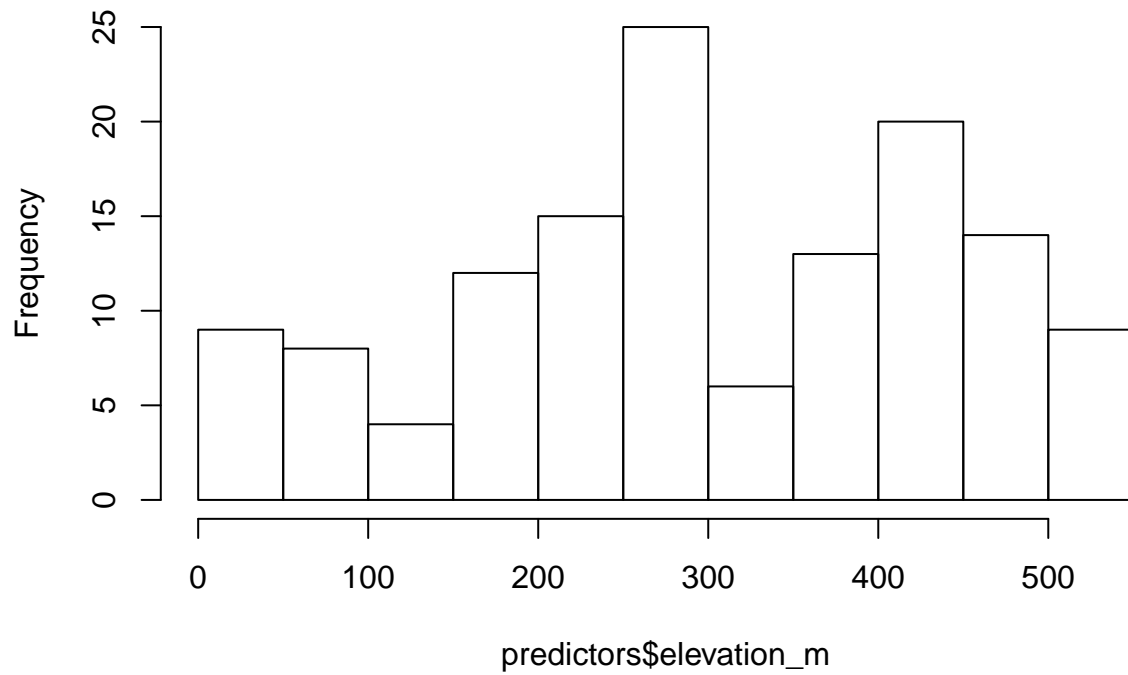
```
## 39000 39004 39009 39010 39012 43601
##      1   110    14     3     6     1
```

```
table(predictors$hul2_zoneid)
```

```
##
## HU12_10463 HU12_10471 HU12_10488 HU12_10493 HU12_10499 HU12_10676
##           1           2           1           1           1           1
## HU12_10700 HU12_10785 HU12_10862 HU12_10863 HU12_10865 HU12_11197
##           1           1           1           4           1           1
## HU12_11495 HU12_11509 HU12_11514 HU12_11515 HU12_11522 HU12_11768
##           1           1           2           1           1           1
## HU12_11816 HU12_11826 HU12_11829 HU12_11889 HU12_11938 HU12_11978
##           1           1           1           1           1           1
## HU12_12113 HU12_12125 HU12_12225 HU12_13098 HU12_13100 HU12_13125
##           1           1           1           5           1           1
## HU12_13164 HU12_13192 HU12_13234 HU12_13241 HU12_13244 HU12_13261
##           1           1           1           1           1           1
## HU12_13300 HU12_13304 HU12_13309 HU12_13354 HU12_13360 HU12_13370
##           1           1           2           1           1           3
## HU12_13374 HU12_13376 HU12_13388 HU12_13413 HU12_13616 HU12_13624
##           1           1           1           1           1           1
## HU12_13628 HU12_13633 HU12_13634 HU12_14494 HU12_14495 HU12_14496
##           1           1           1           1           1           1
## HU12_14497 HU12_14533 HU12_148 HU12_1494 HU12_15183 HU12_15280
##           4           1           1           1           1           1
## HU12_15296 HU12_15315 HU12_15329 HU12_1537 HU12_15856 HU12_16122
##           1           1           1           1           1           1
## HU12_16125 HU12_1615 HU12_1621 HU12_16347 HU12_16746 HU12_16747
##           1           1           2           1           1           2
## HU12_16749 HU12_16835 HU12_16882 HU12_17143 HU12_17178 HU12_17235
##           1           1           1           1           1           2
## HU12_17401 HU12_17407 HU12_17433 HU12_17477 HU12_17488 HU12_17504
##           1           1           1           1           2           2
## HU12_17512 HU12_17513 HU12_17541 HU12_17646 HU12_17651 HU12_17655
##           1           2           1           1           1           1
## HU12_1802 HU12_18174 HU12_1819 HU12_1828 HU12_18730 HU12_1896
##           1           1           1           1           1           1
## HU12_19726 HU12_1980 HU12_19842 HU12_20279 HU12_2173 HU12_2200
##           1           1           1           1           1           1
## HU12_2239 HU12_2410 HU12_2412 HU12_2429 HU12_4337 HU12_4347
##           1           1           1           1           1           1
## HU12_442 HU12_488 HU12_509 HU12_542 HU12_581 HU12_829
##           1           1           1           1           1           1
```

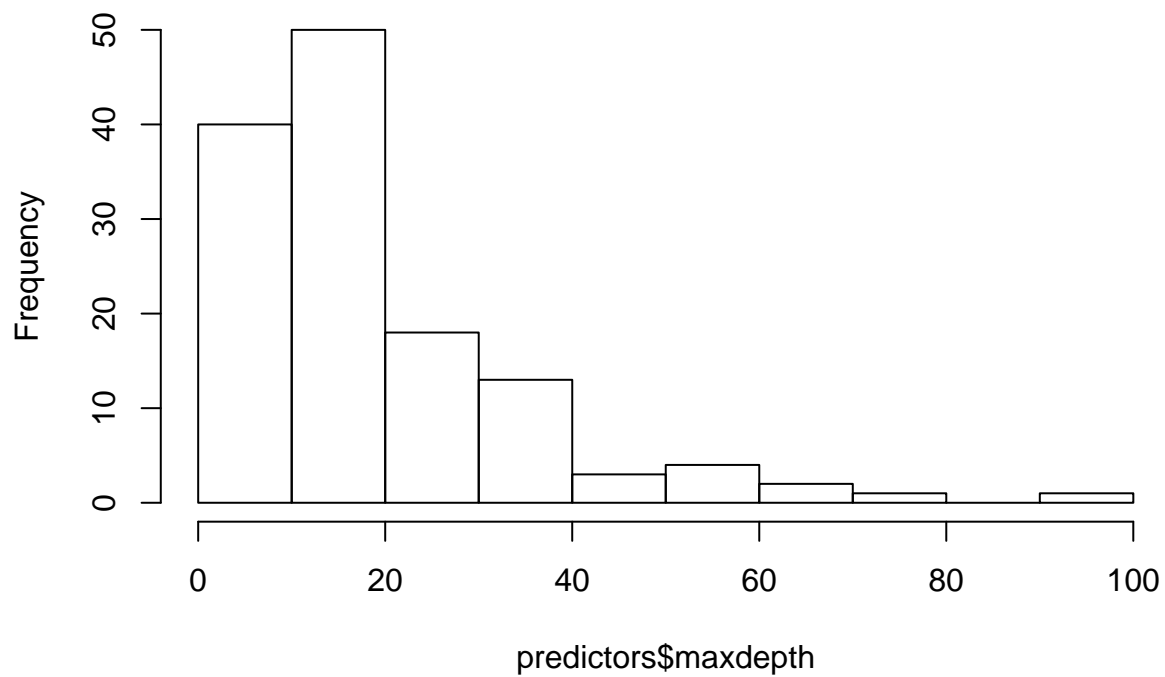
```
hist(predictors$elevation_m)
```

Histogram of predictors\$elevation_m



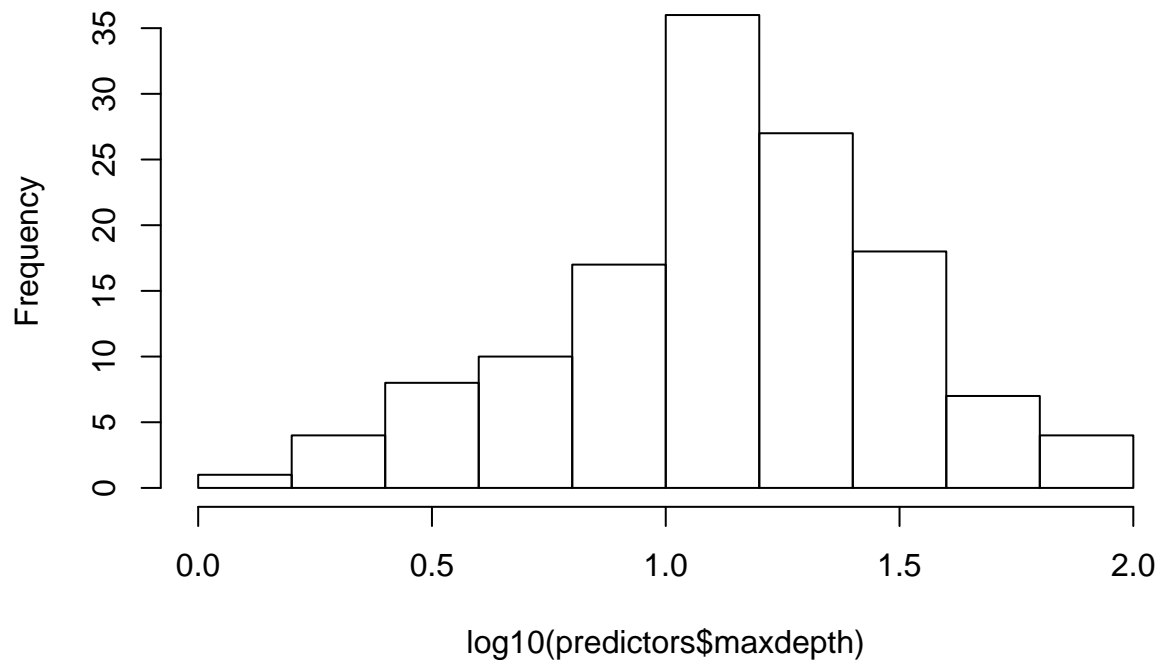
```
hist(predictors$maxdepth)
```

Histogram of predictors\$maxdepth



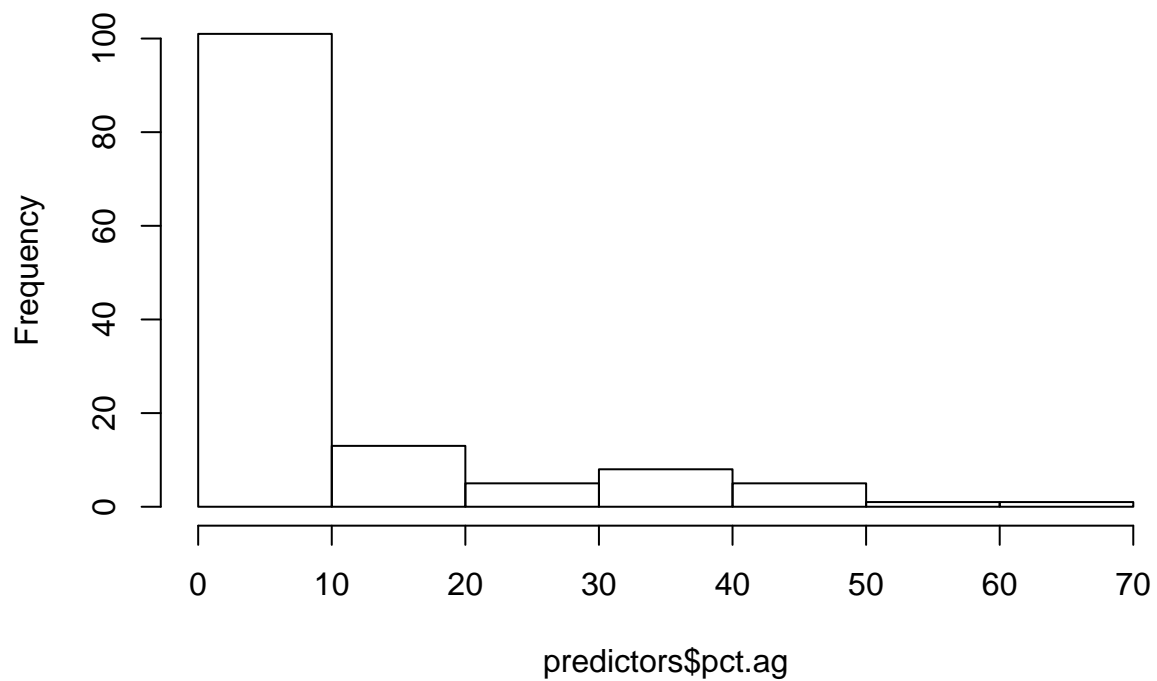
```
hist(log10(predictors$maxdepth))
```

Histogram of $\log_{10}(\text{predictors}\$maxdepth)$



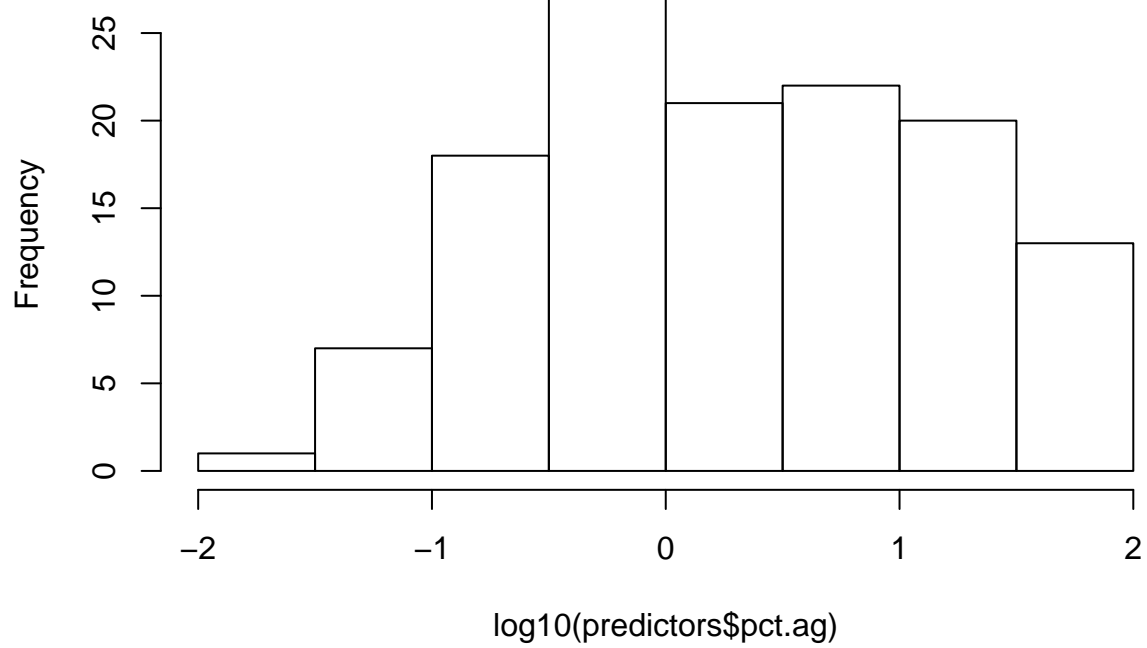
```
hist(predictors$pct.ag)
```

Histogram of $\text{predictors}\$pct.ag$



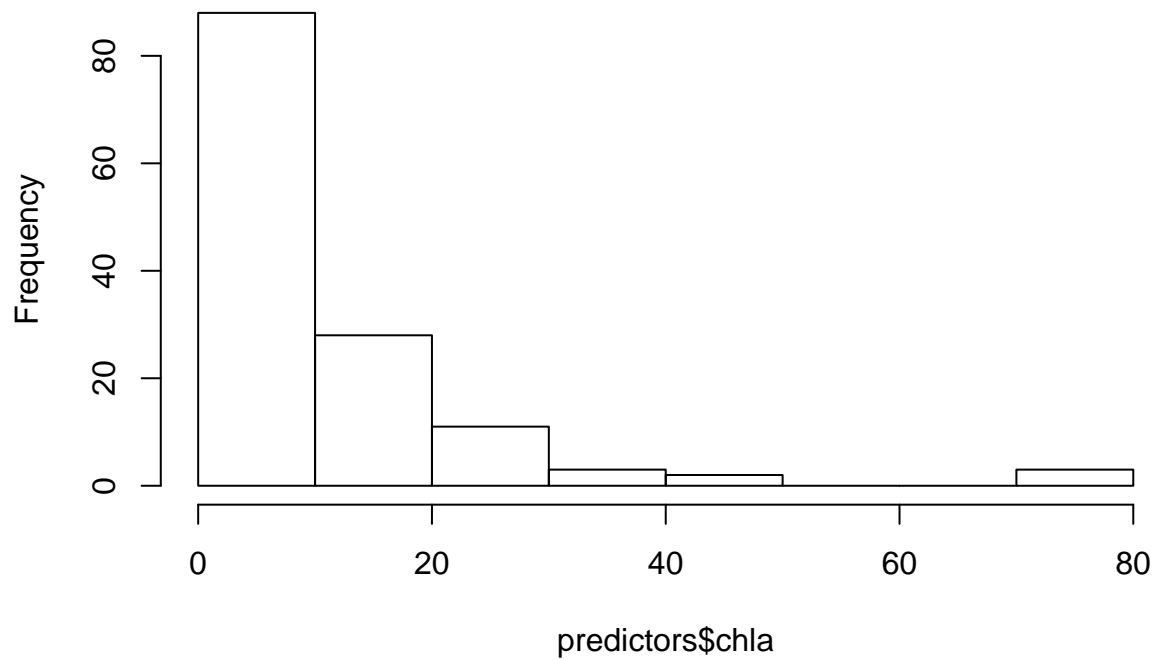
```
hist(log10(predictors$pct.ag))
```

Histogram of $\log_{10}(\text{predictors}\$pct.ag)$



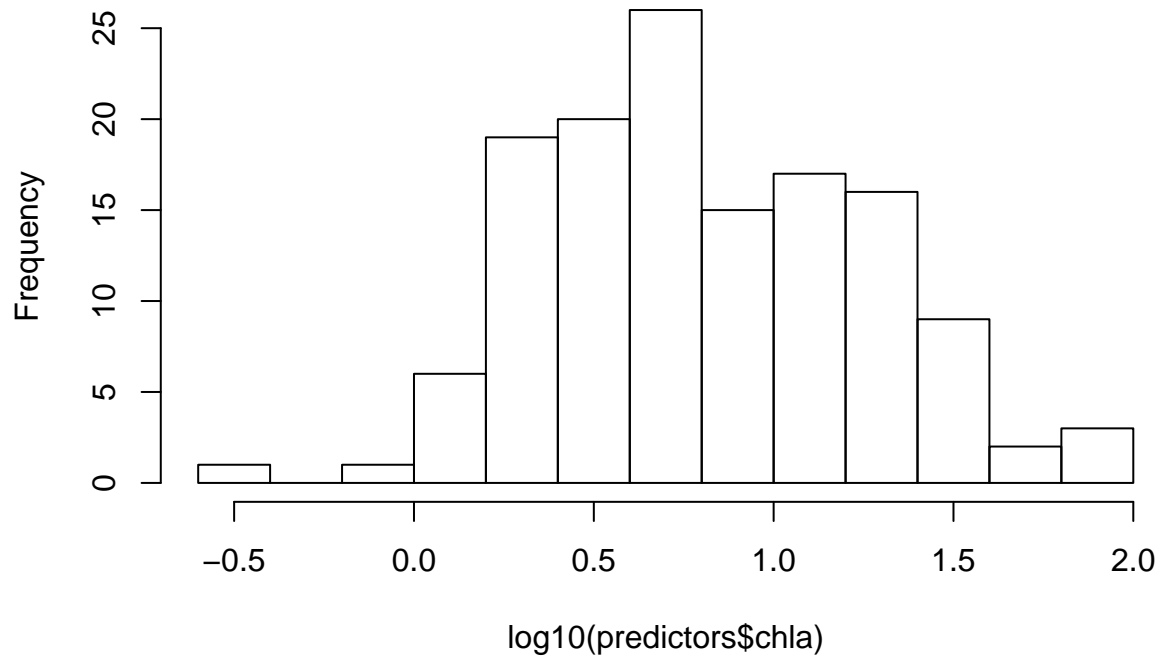
```
hist(predictors$chla)
```

Histogram of predictors\$chla



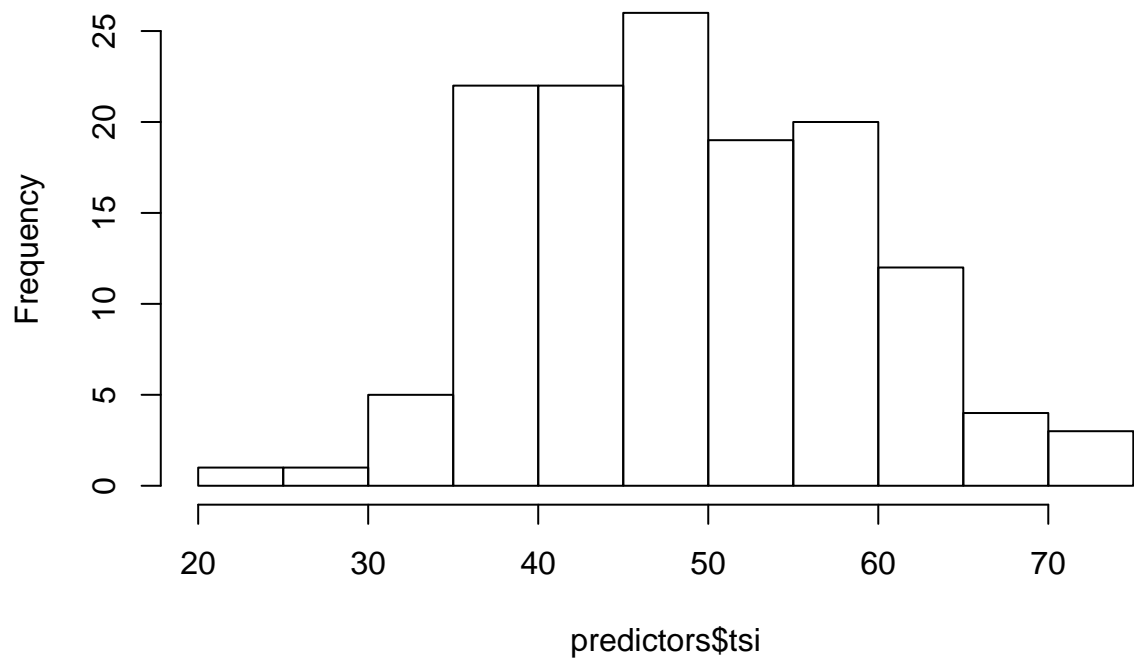
```
hist(log10(predictors$chla))
```


Histogram of $\log_{10}(\text{predictors}\$chla)$



```
hist(predictors$tsi)
```

Histogram of $\text{predictors}\$tsi$

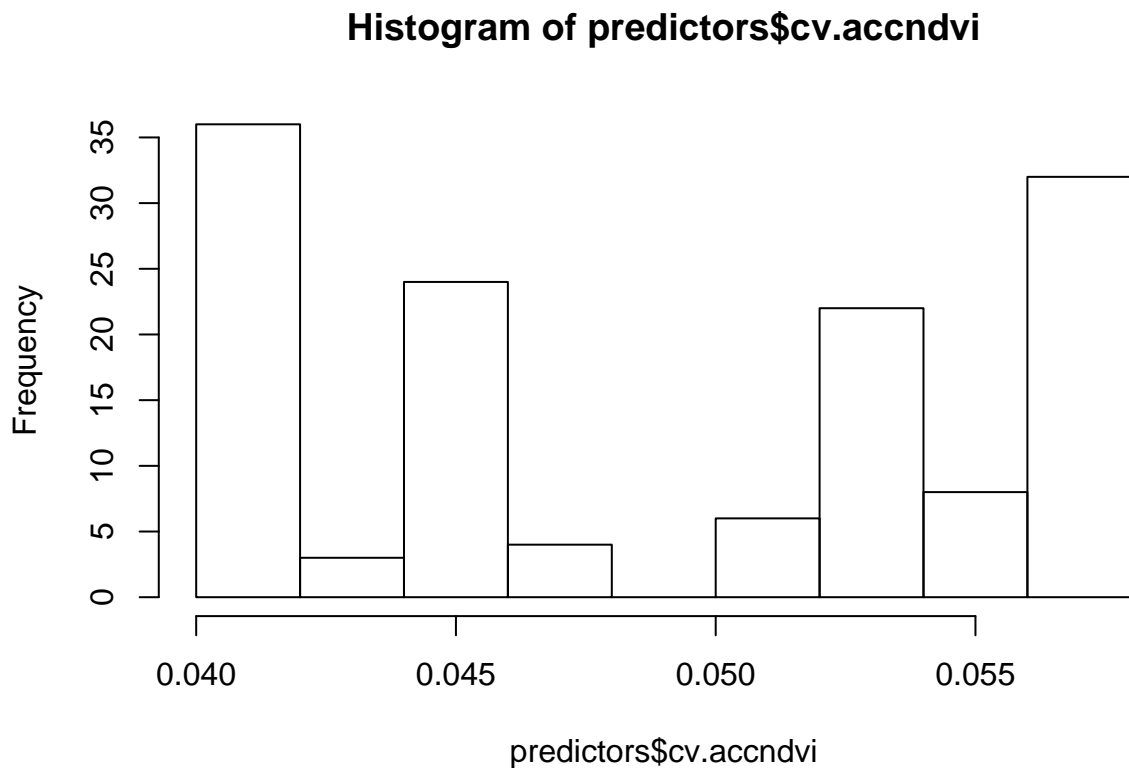


```
table(predictors$tsi.cat)
```

```
##
```

```
##      eutrophic hypereutrophic      mesotrophic      oligotrophic
##             55             3             48             29
```

```
hist(predictors$cv.accndvi)
```



```
predictors$log10_lake_area_ha<-log10(predictors$lake_area_ha)
predictors$log10_lake_perim_meters<-log10(predictors$lake_perim_meters)
predictors$log10_maxdepth<-log10(predictors$maxdepth)
predictors$log10_pct.ag<-log10(predictors$pct.ag+1)
predictors$log10_chla<-log10(predictors$chla)
```

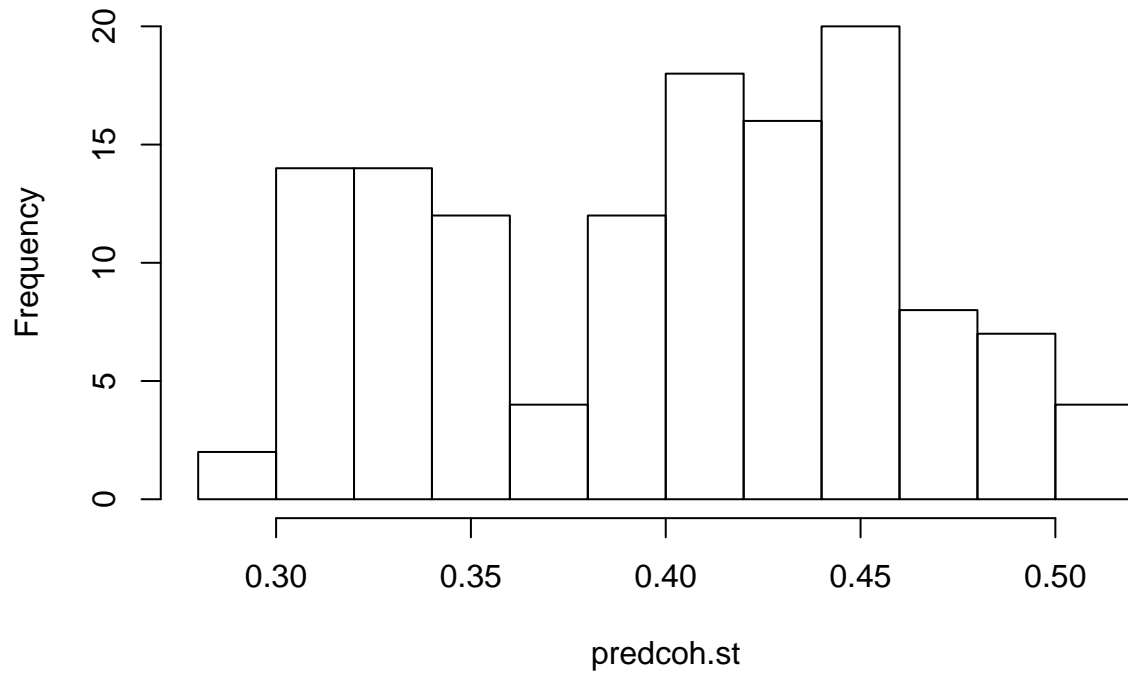
```
modvars.accndvi<-left_join(predictors, coh.chlaXaccndvi, by="lagoslakeid")
modvars.accndvi$nhd_ftype<-factor(modvars.accndvi$nhd_ftype)
modvars.accndvi$tsi.cat<-factor(modvars.accndvi$tsi.cat)
modvars.accndvi$tslength<-modvars.accndvi$end-modvars.accndvi$start + 1
```

```
modvars.accndvi<-modvars.accndvi[!is.na(modvars.accndvi$maxdepth),]
modvars.accndvi<-modvars.accndvi[!is.na(modvars.accndvi$pct.ag),]
```

```
modvars.accndvi.phist<-modvars.accndvi[modvars.accndvi$accndvip.ts1<0.3,]
modvars.accndvi.philt<-modvars.accndvi[modvars.accndvi$accndvip.ts2<0.3,]
```

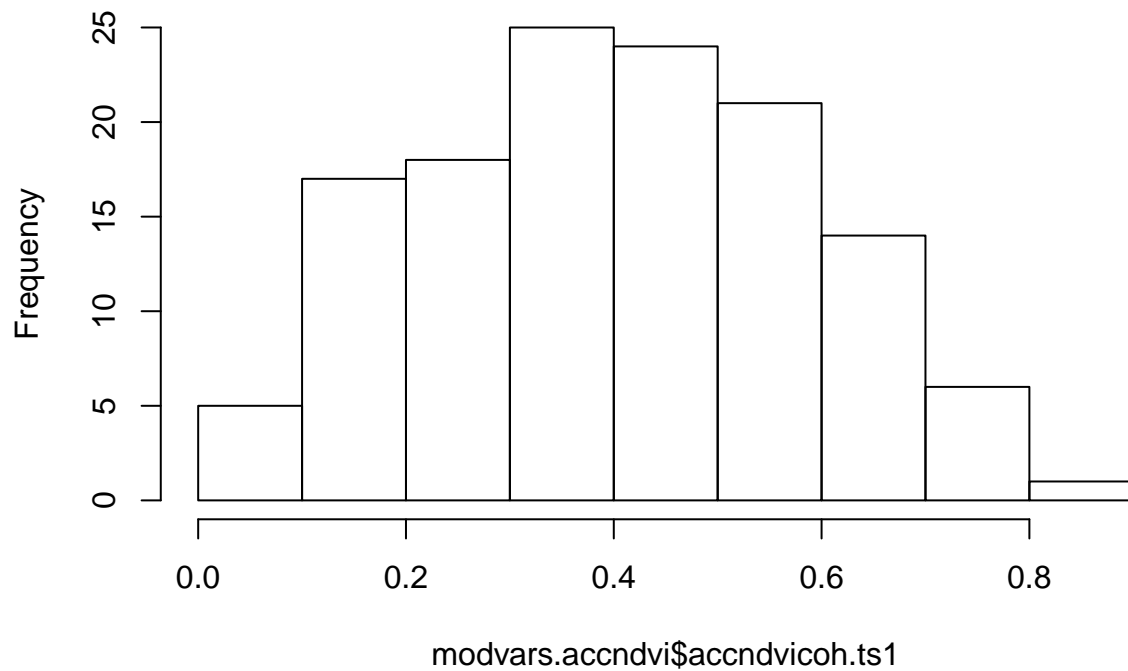
```
# cforest.st<-partykit::cforest(accndvicoh.ts1 ~ lake_area_ha + lake_perim_meters + maxdepth + pct.ag +
#                               data=modvars.accndvi, ntree=20000)
cforest.st<-party::cforest(accndvicoh.ts1 ~ lake_area_ha + lake_perim_meters + maxdepth + pct.ag + chla
                           data=modvars.accndvi, controls=cforest_control(ntree=80000))
predcoh.st<-predict(cforest.st, newdata=modvars.accndvi,type="response")
hist(predcoh.st)
```

Histogram of predcoh.st



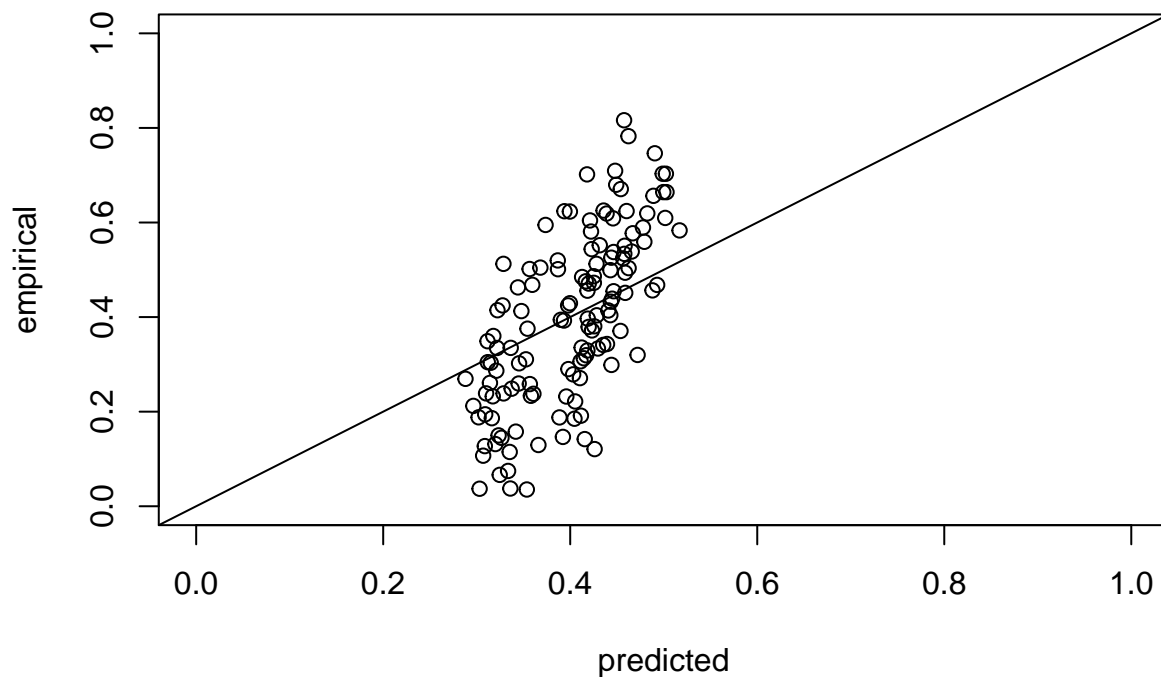
```
hist(modvars.accndvi$accndvicoh.ts1)
```

Histogram of modvars.accndvi\$accndvicoh.ts1



```
plot(predcoh.st, modvars.accndvi$accndvicoh.ts1, xlab="predicted", ylab="empirical", main="Coherence, s  
      xlim=c(0,1), ylim=c(0,1))  
abline(a=0,b=1)
```

Coherence, short ts



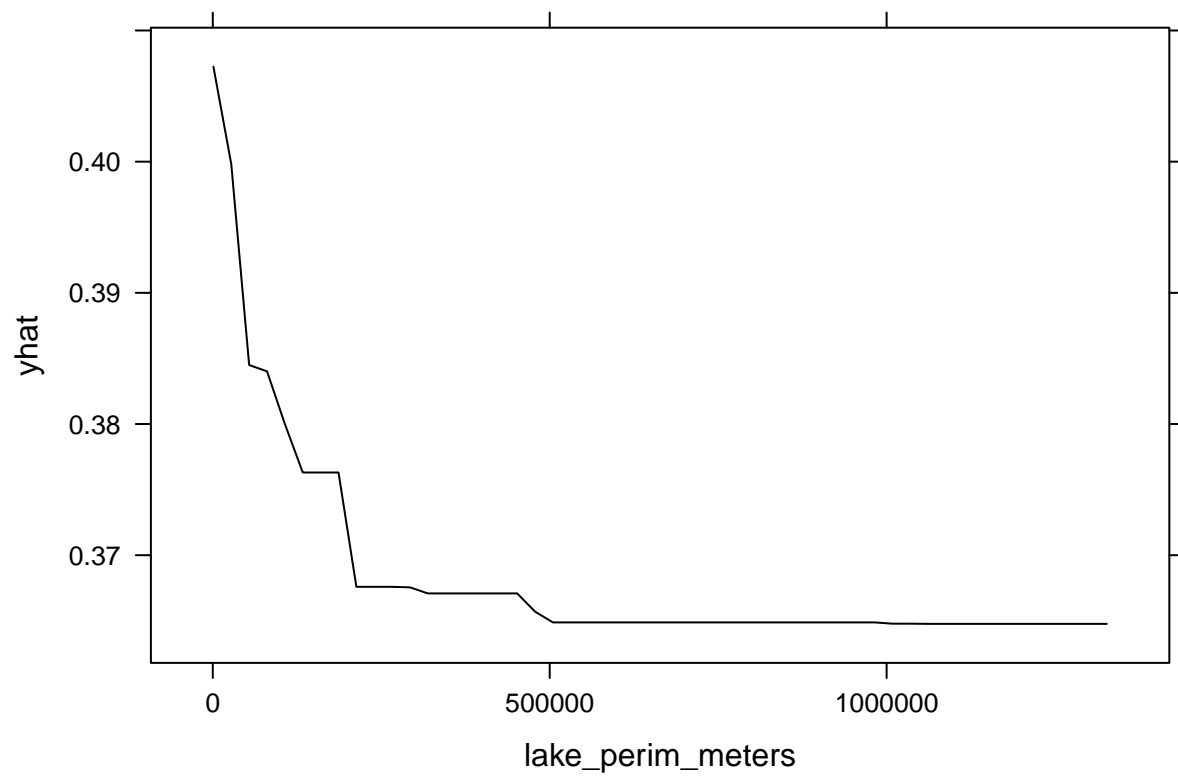
```
cor.test(predcoh.st,modvars.accndvi$accndvicoh.ts1)
```

```
##
## Pearson's product-moment correlation
##
## data: predcoh.st and modvars.accndvi$accndvicoh.ts1
## t = 10.771, df = 129, p-value < 2.2e-16
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
##  0.5857340 0.7688966
## sample estimates:
##      cor
## 0.6881247
```

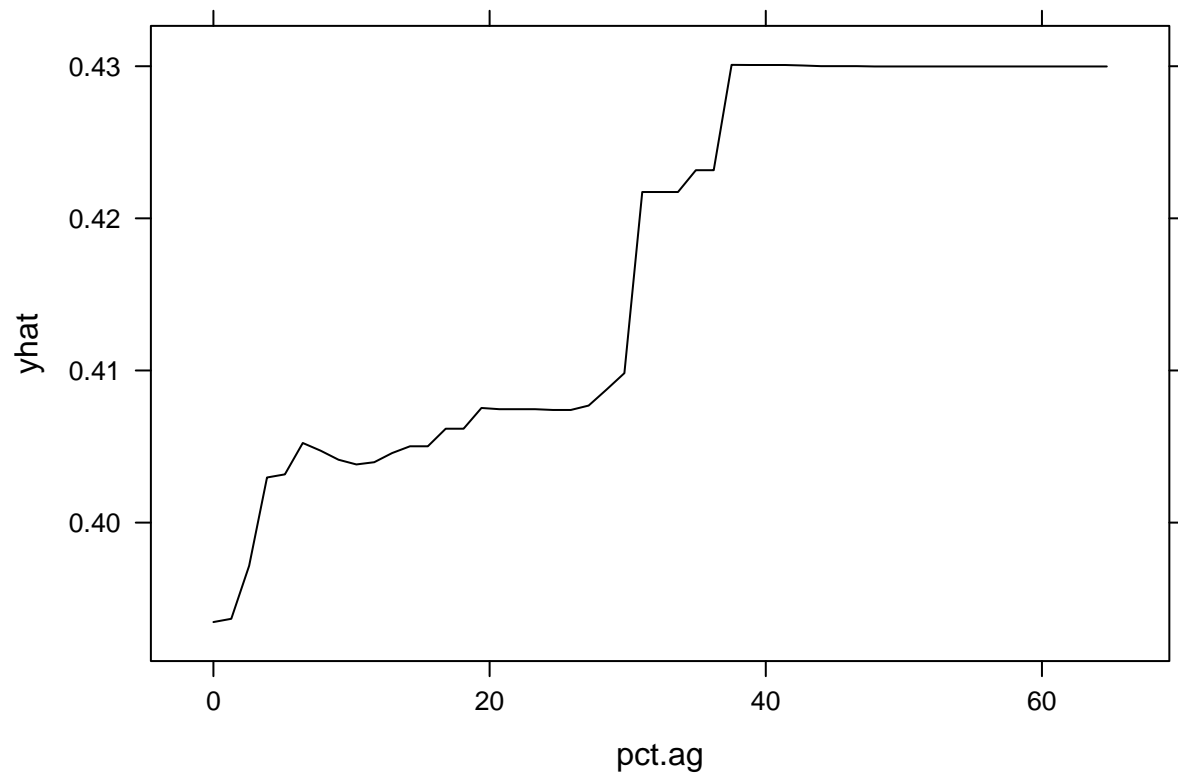
```
varimp.coh.st<-varimp(cforest.st)
print(varimp.coh.st[order(varimp.coh.st, decreasing=T)])
```

```
## lake_perim_meters      pct.ag      lake_area_ha      maxdepth
## 7.562703e-04      6.095637e-04      3.176892e-04      1.959366e-04
##          chla      cv.accndvi      tsi.cat      hu4_zoneid
## -2.003222e-05      -1.260899e-04      -5.684776e-04      -7.340727e-04
```

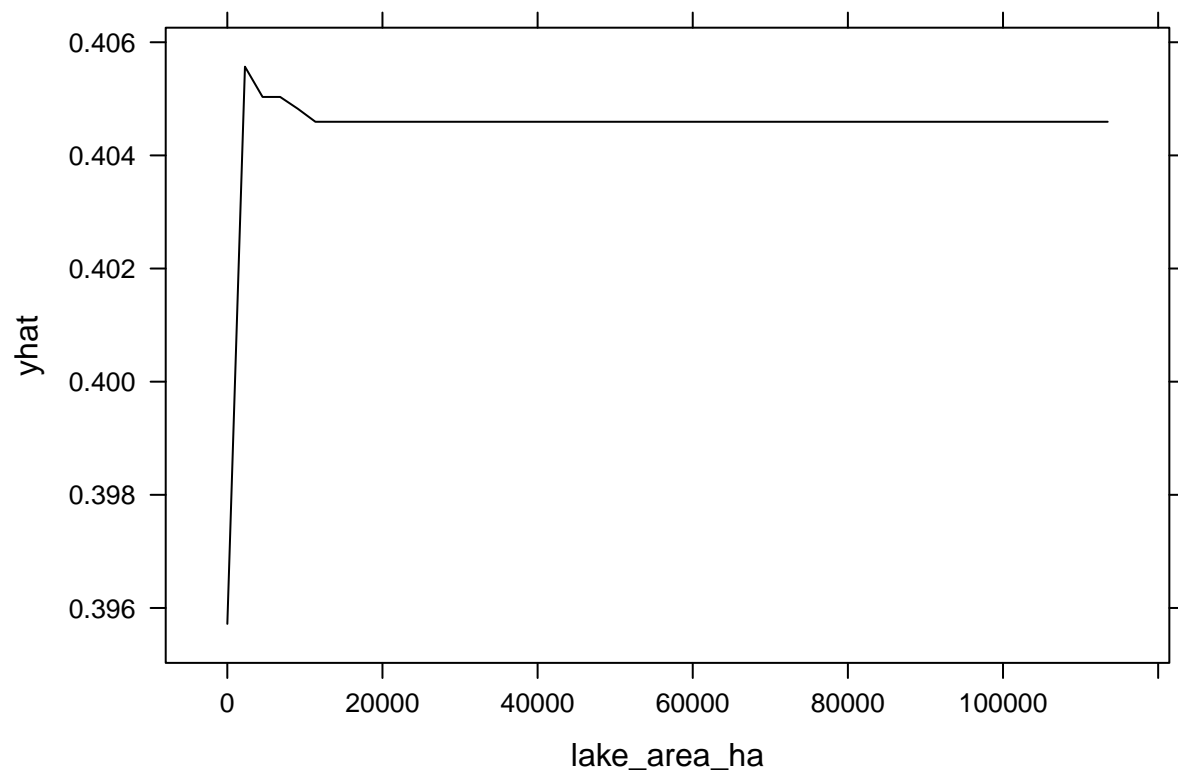
```
partial(cforest.st, pred.var="lake_perim_meters", train=modvars.accndvi, type="regression", plot=T)
```



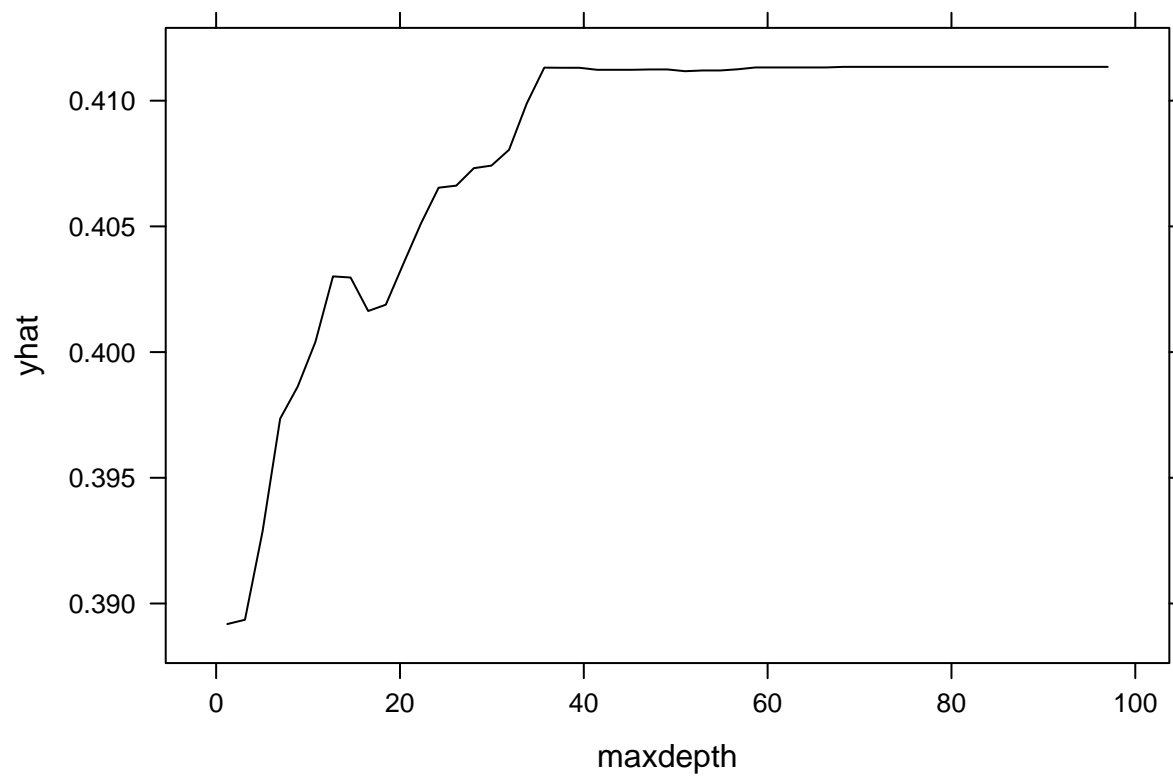
```
partial(cforest.st, pred.var="pct.ag", train=modvars.accndvi, type="regression", plot=T)
```



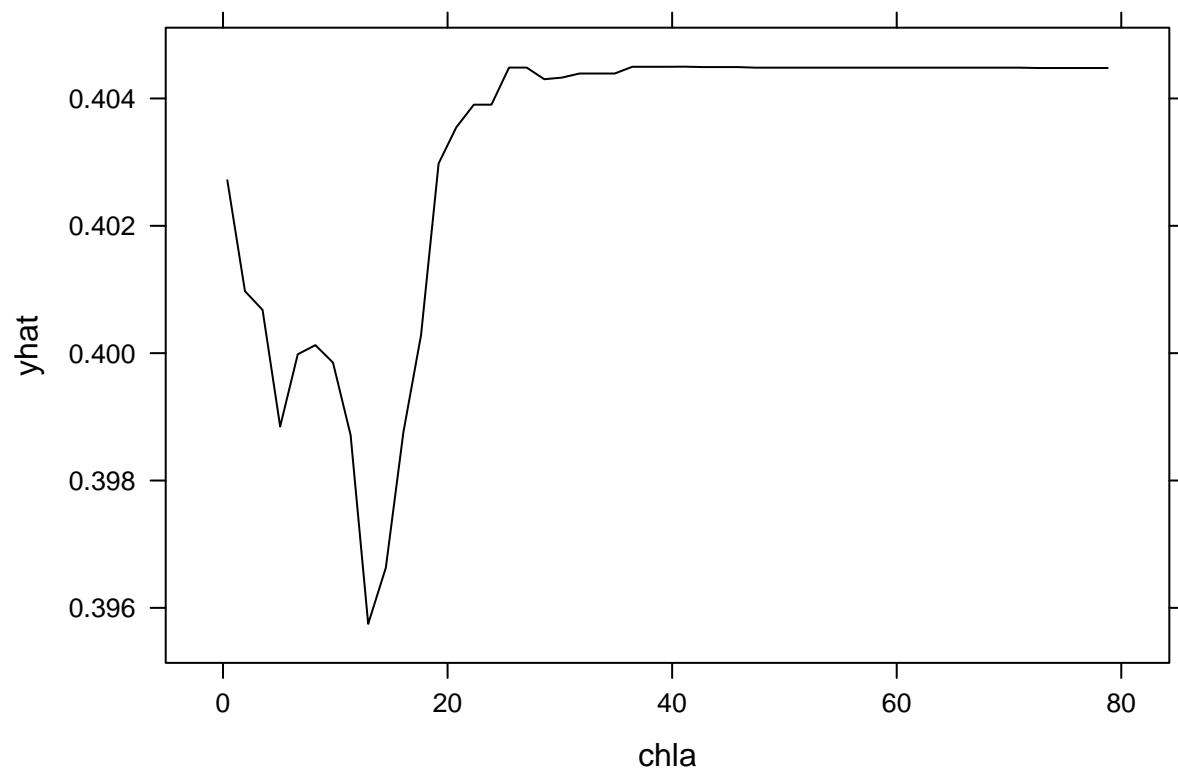
```
partial(cforest.st, pred.var="lake_area_ha", train=modvars.accndvi, type="regression", plot=T)
```



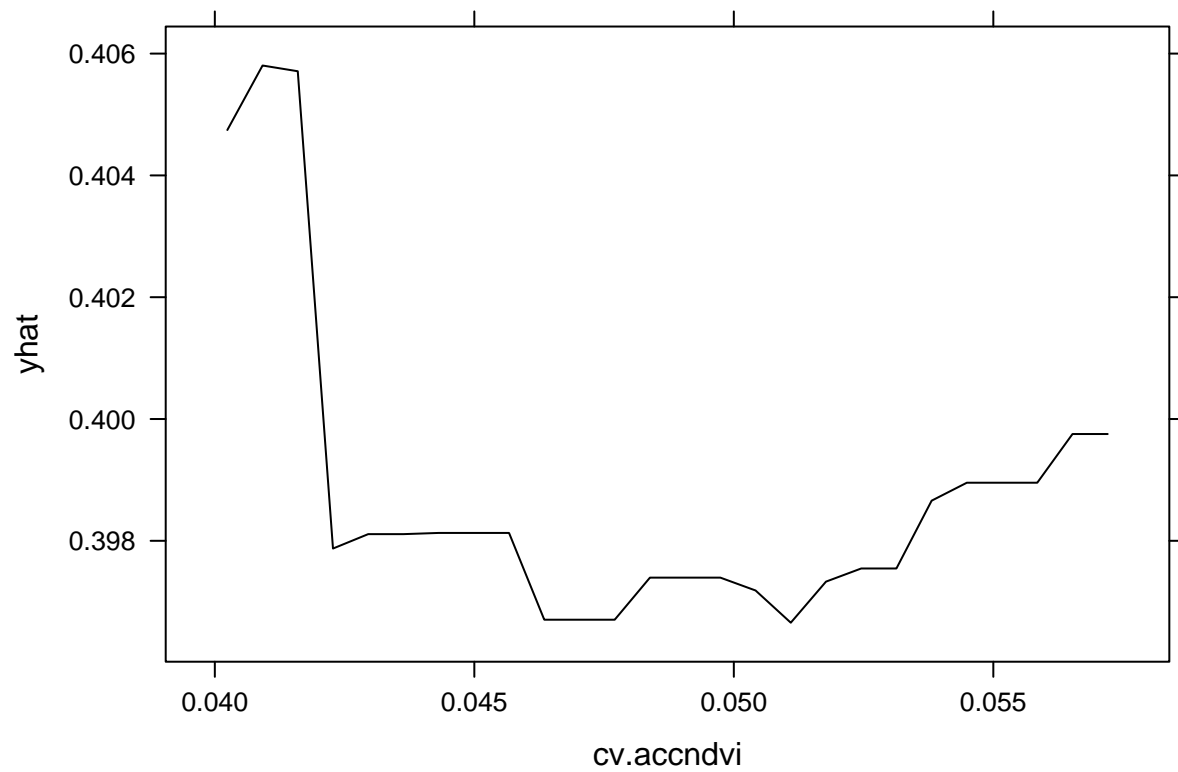
```
partial(cforest.st, pred.var="maxdepth", train=modvars.accndvi, type="regression", plot=T)
```



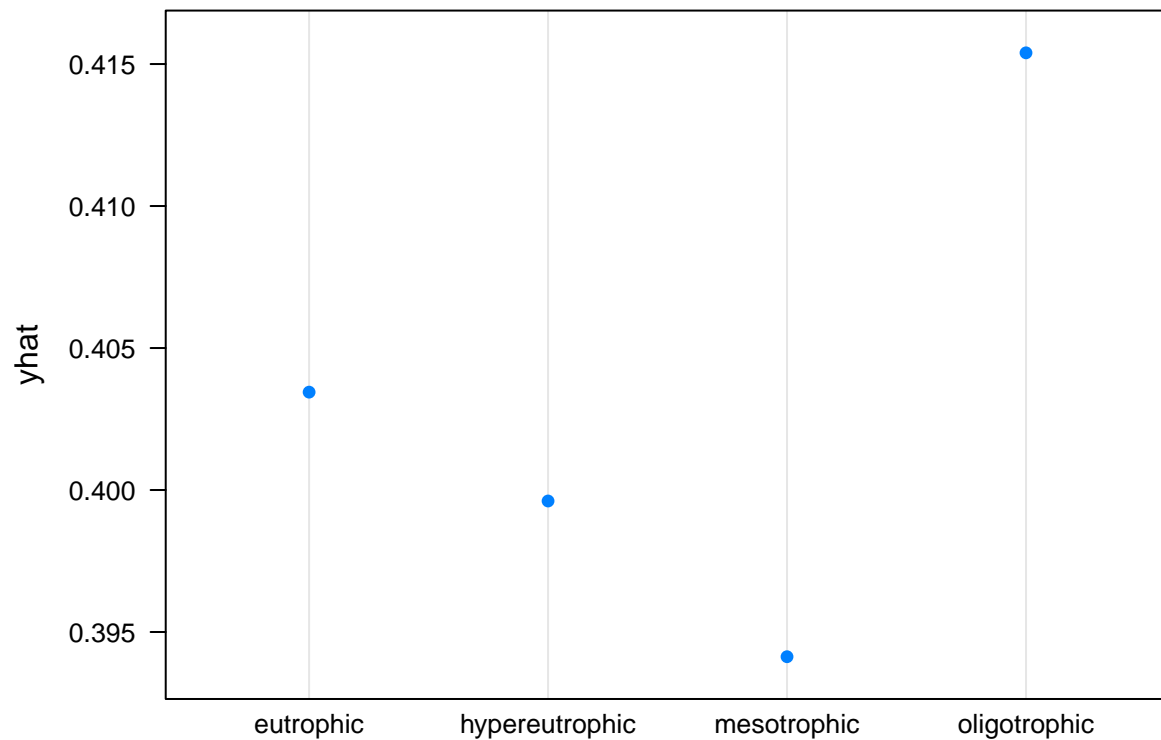
```
partial(cforest.st, pred.var="chla", train=modvars.accndvi, type="regression", plot=T)
```



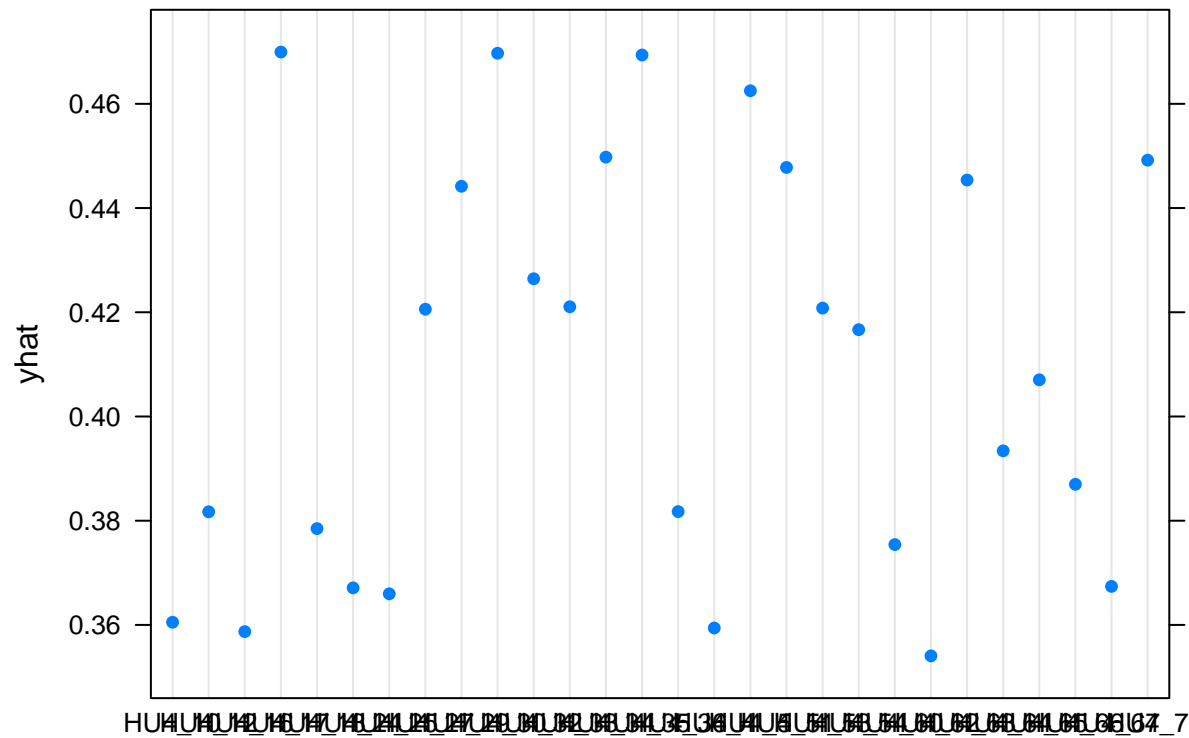
```
partial(cforest.st, pred.var="cv.accndvi", train=modvars.accndvi, type="regression", plot=T)
```



```
partial(cforest.st, pred.var="tsi.cat", train=modvars.accndvi, type="regression", plot=T)
```



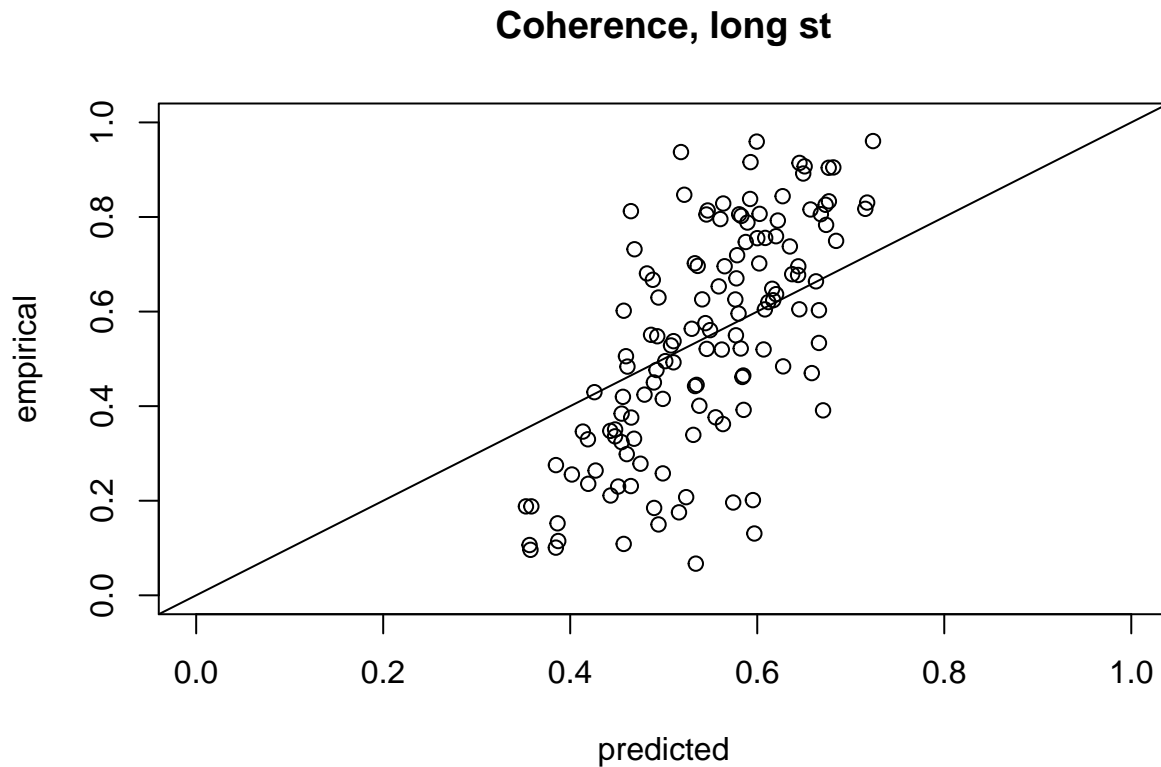
```
partial(cforest.st, pred.var="hu4_zoneid", train=modvars.accndvi, type="regression", plot=T)
```



```
cforest.lt<-party::cforest(accndvicoh.ts2 ~ lake_area_ha + lake_perim_meters + maxdepth + pct.ag + chla
                           data=modvars.accndvi, controls=cforest_control(ntree=80000))
predcoh.lt<-predict(cforest.lt, newdata=modvars.accndvi)
# hist(predcoh.lt)
# hist(modvars.accndvi$accndvicoh.ts2)
```



```
plot(predcoh.lt, modvars.accndvi$accndvicoh.ts2, xlab="predicted", ylab="empirical", main="Coherence, long st",
      xlim=c(0,1), ylim=c(0,1))
abline(a=0,b=1)
```



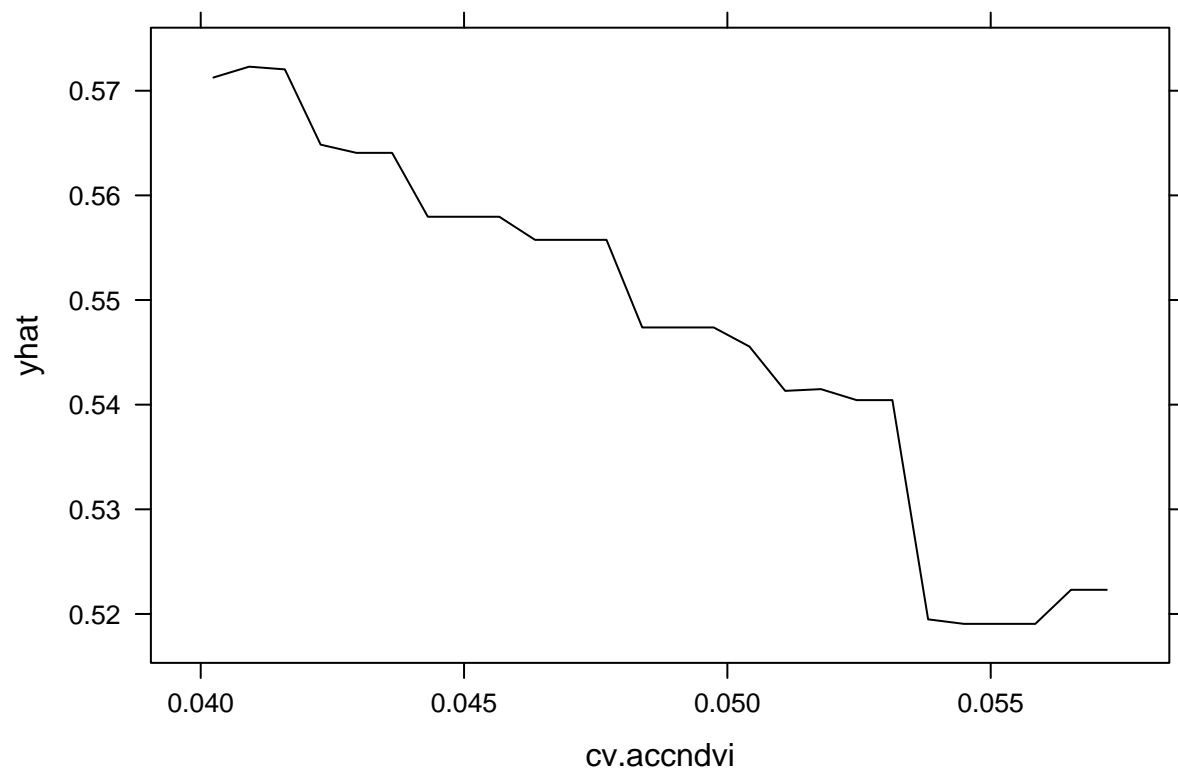
```
cor.test(predcoh.lt,modvars.accndvi$accndvicoh.ts2)
```

```
##
## Pearson's product-moment correlation
##
## data: predcoh.lt and modvars.accndvi$accndvicoh.ts2
## t = 10.735, df = 129, p-value < 2.2e-16
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
## 0.5841903 0.7679356
## sample estimates:
## cor
## 0.6868872
```

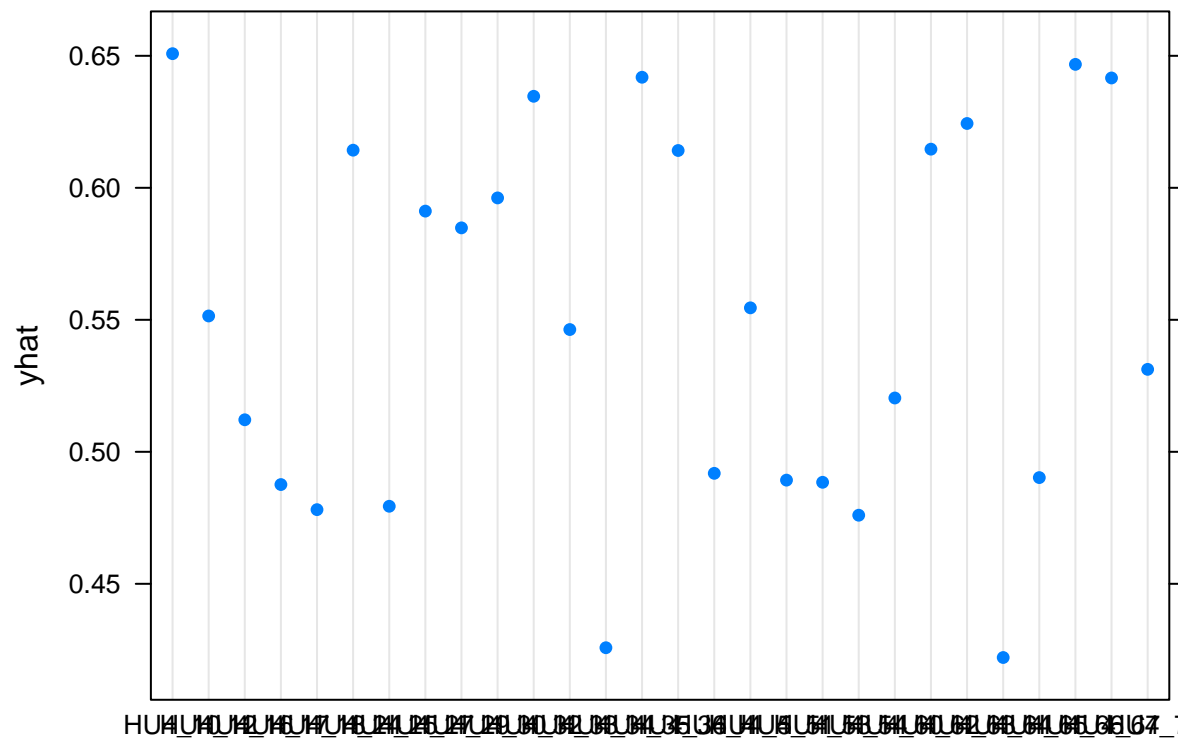
```
varimp.coh.lt<-varimp(cforest.lt)
print(varimp.coh.lt[order(varimp.coh.lt, decreasing=T)])
```

```
##      cv.accndvi      hu4_zoneid      lake_area_ha      pct.ag
## 9.854397e-04      3.224742e-04      -7.940386e-05      -8.382066e-05
## lake_perim_meters      chla      tsi.cat      maxdepth
## -8.584673e-05      -2.037749e-04      -4.266195e-04      -5.378430e-04
```

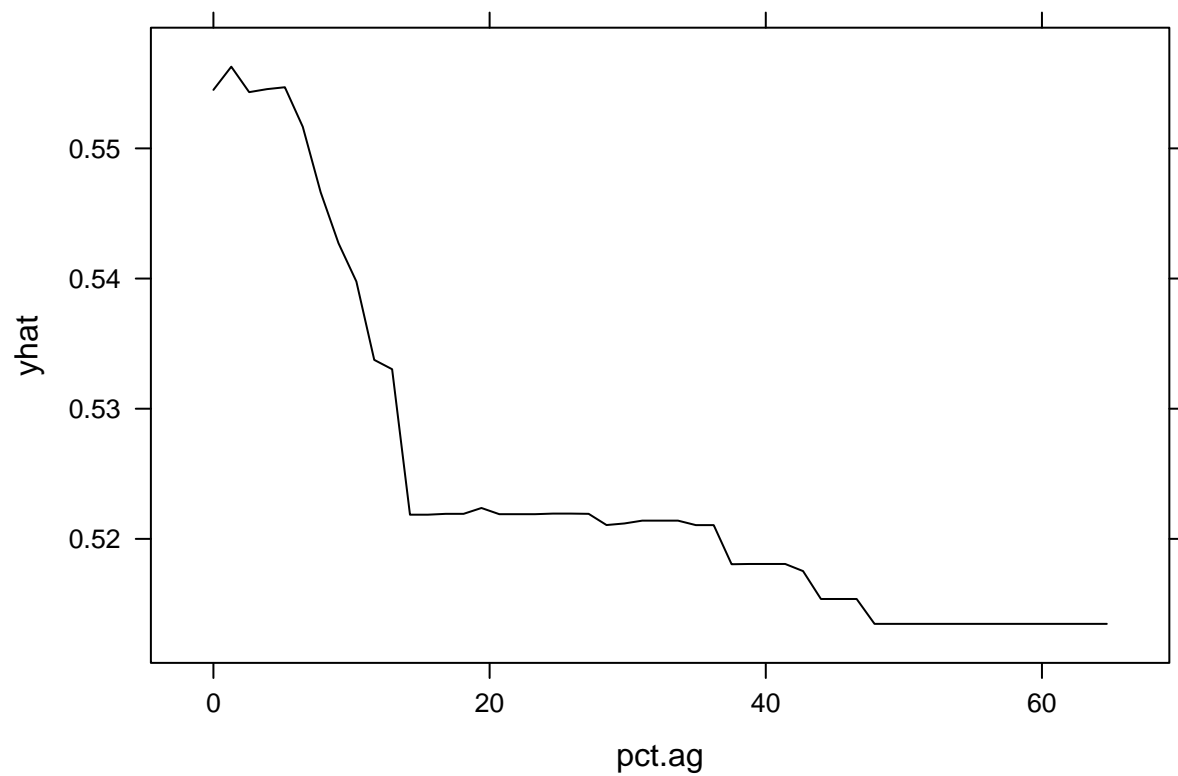
```
partial(cforest.lt, pred.var="cv.accndvi", train=modvars.accndvi, type="regression", plot=T)
```



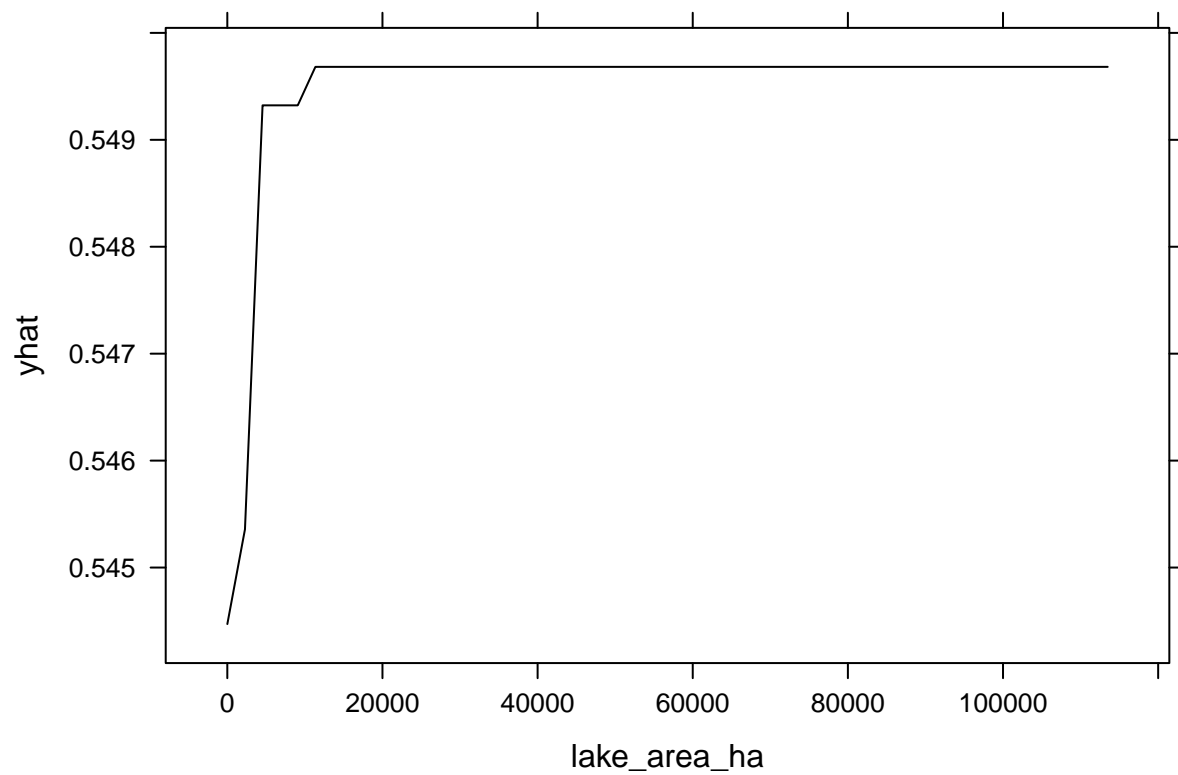
```
partial(cforest.lt, pred.var="hu4_zoneid", train=modvars.accndvi, type="regression", plot=T)
```



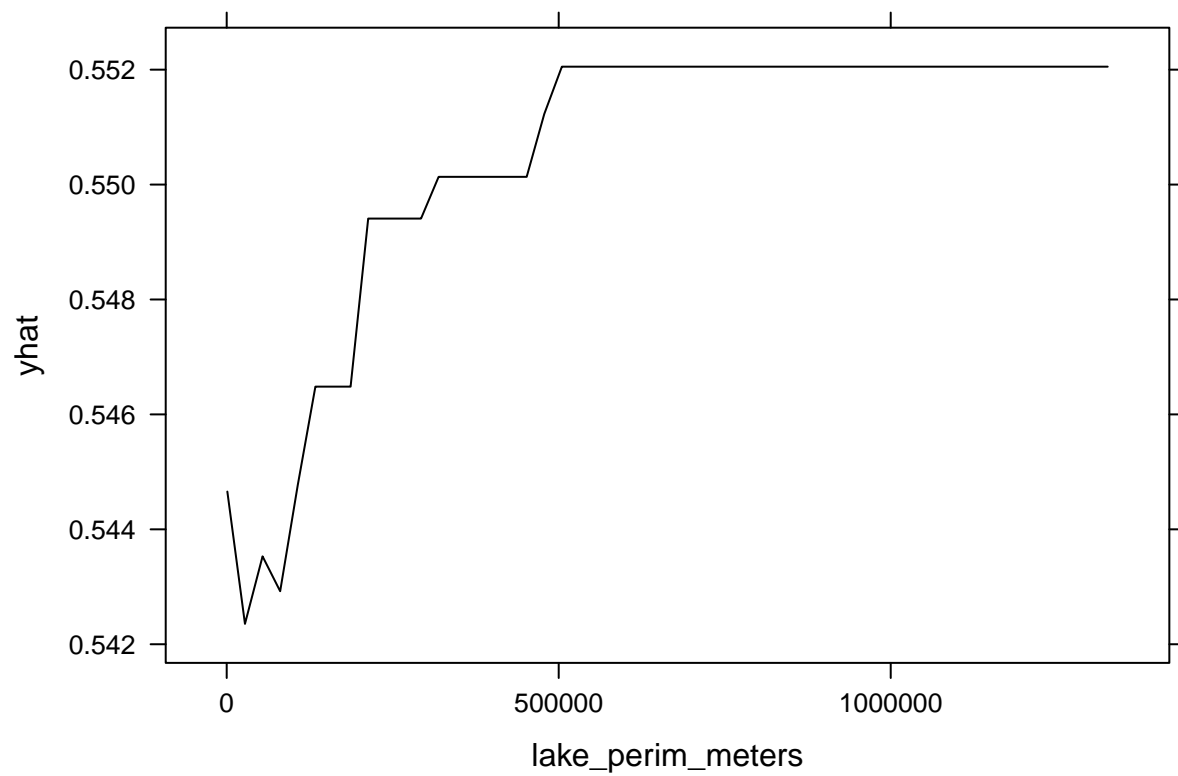
```
partial(cforest.lt, pred.var="pct.ag", train=modvars.accndvi, type="regression", plot=T)
```



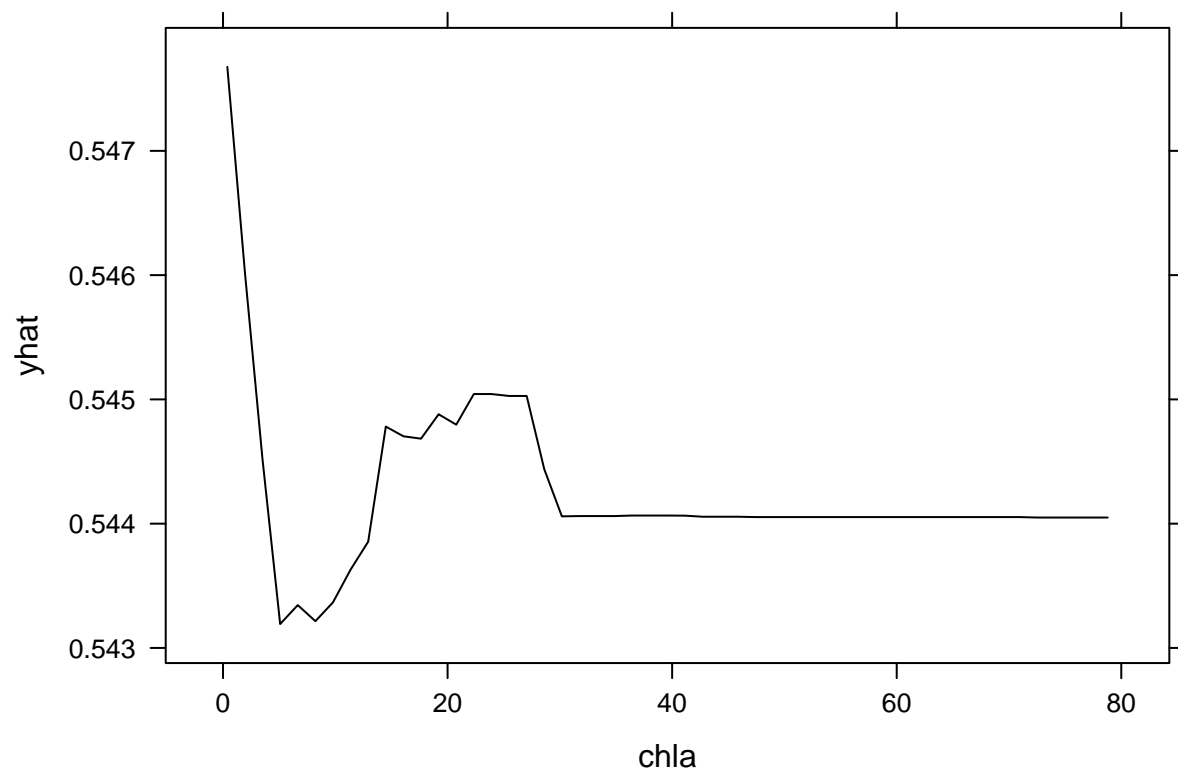
```
partial(cforest.lt, pred.var="lake_area_ha", train=modvars.accndvi, type="regression", plot=T)
```



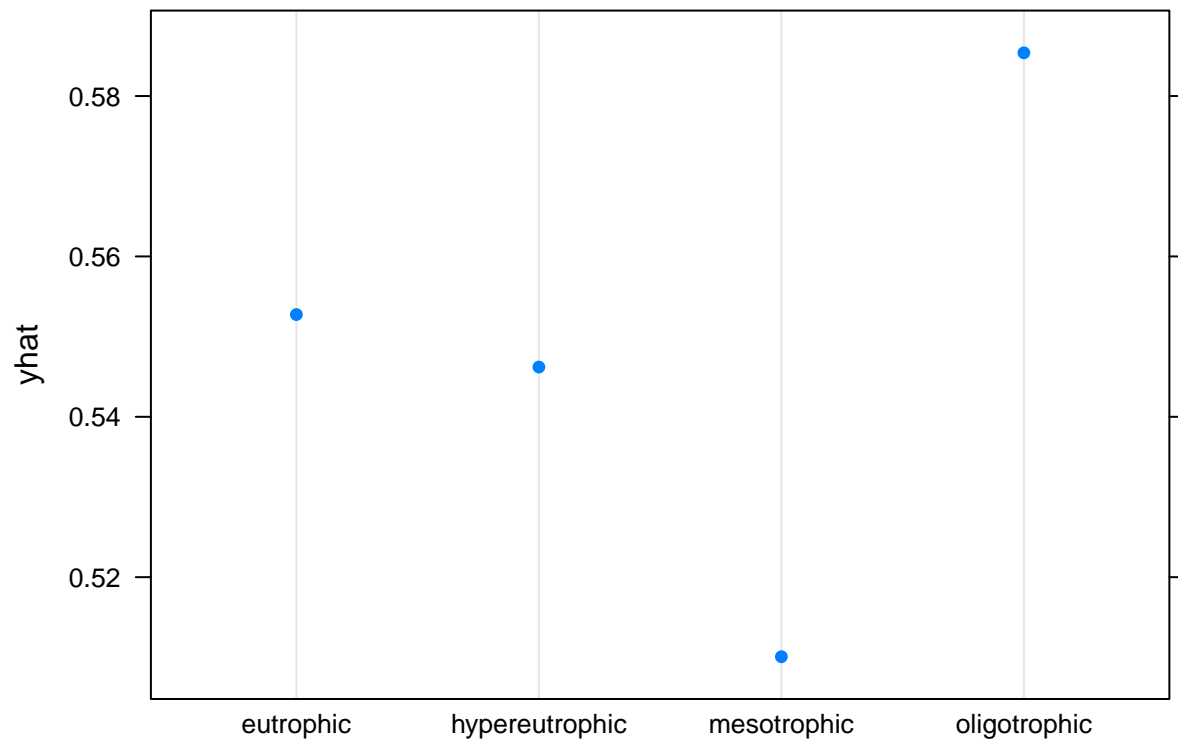
```
partial(cforest.lt, pred.var="lake_perim_meters", train=modvars.accndvi, type="regression", plot=T)
```



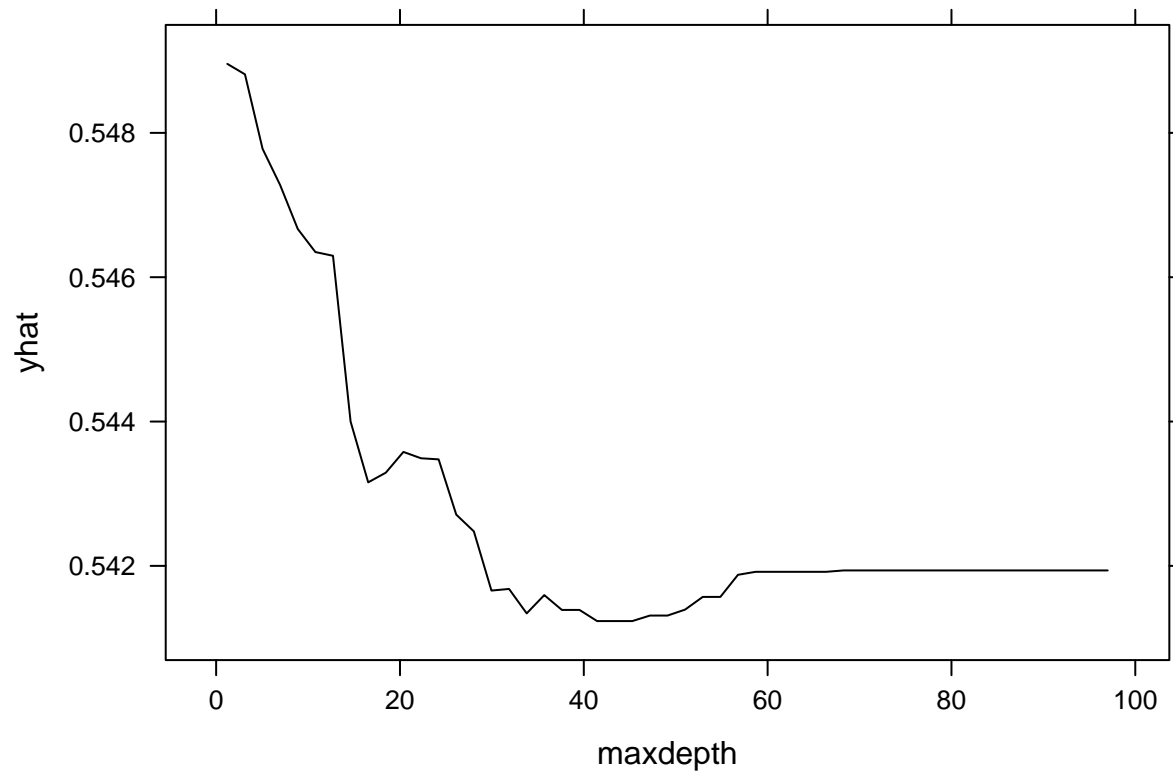
```
partial(cforest.lt, pred.var="chla", train=modvars.accndvi, type="regression", plot=T)
```



```
partial(cforest.lt, pred.var="tsi.cat", train=modvars.accndvi, type="regression", plot=T)
```

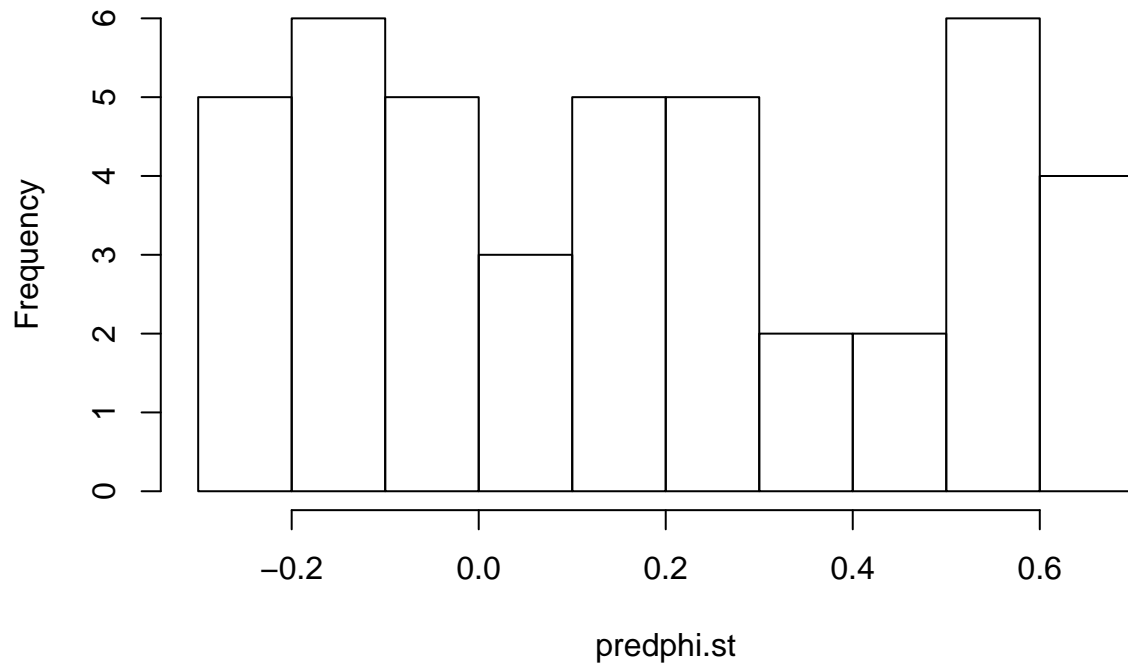


```
partial(cforest.lt, pred.var="maxdepth", train=modvars.accndvi, type="regression", plot=T)
```



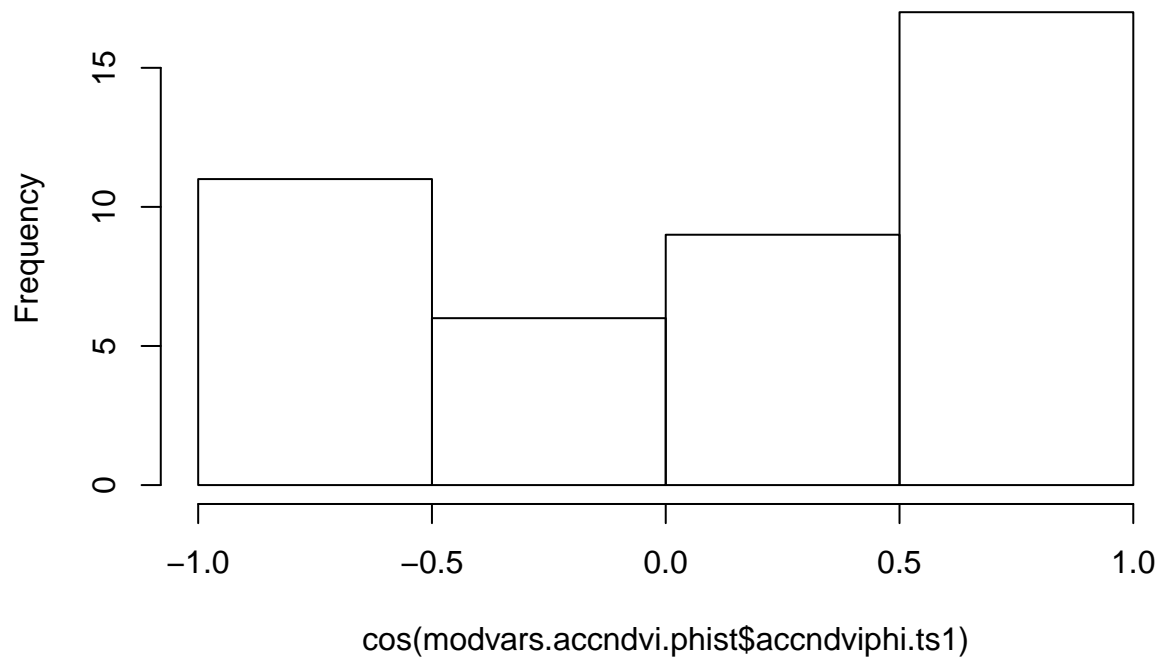
```
cforest.phi.st<-cforest(cos(accndviphi.ts1) ~ lake_area_ha + lake_perim_meters + maxdepth + pct.ag + ch.
                        data=modvars.accndvi.phist,controls=cforest_control(ntree=30000))
predphi.st<-predict(cforest.phi.st, newdata=modvars.accndvi.phist)
hist(predphi.st)
```

Histogram of predphi.st



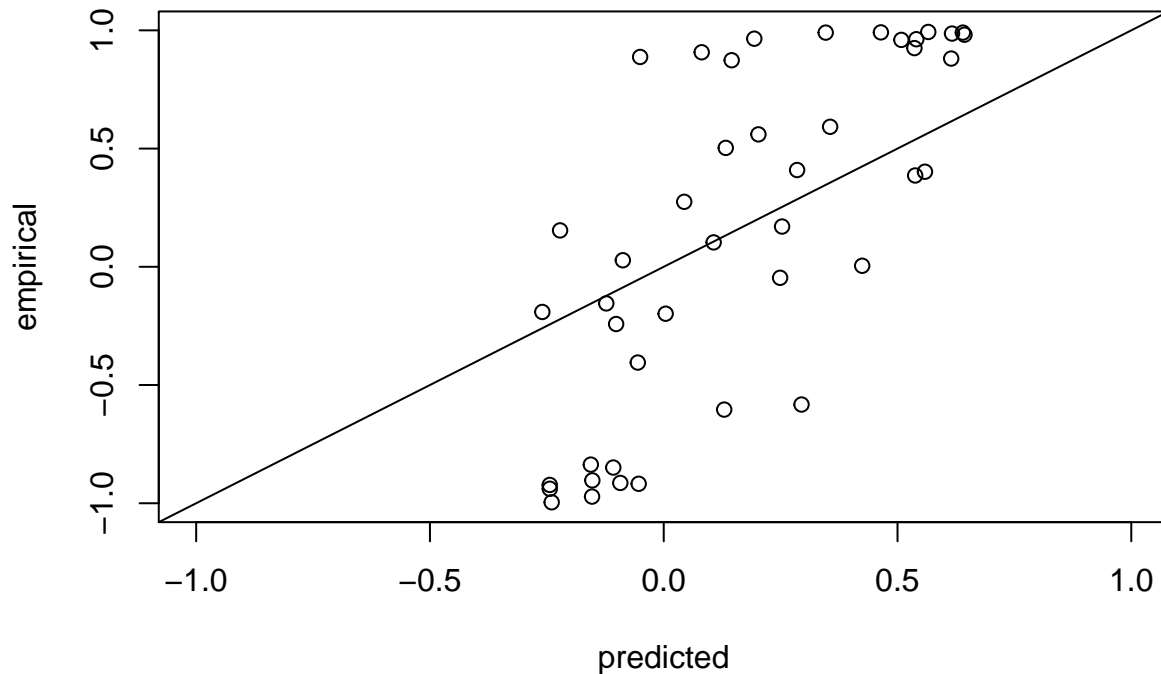
```
hist(cos(modvars.accndvi.phist$accndviphi.ts1))
```

Histogram of cos(modvars.accndvi.phist\$accndviphi.ts1)



```
plot(predphi.st, cos(modvars.accndvi.phist$accndviphi.ts1), xlab="predicted", ylab="empirical", main="cos")
xlim=c(-1,1), ylim=c(-1,1))
abline(a=0,b=1)
```

cos(phase), short ts



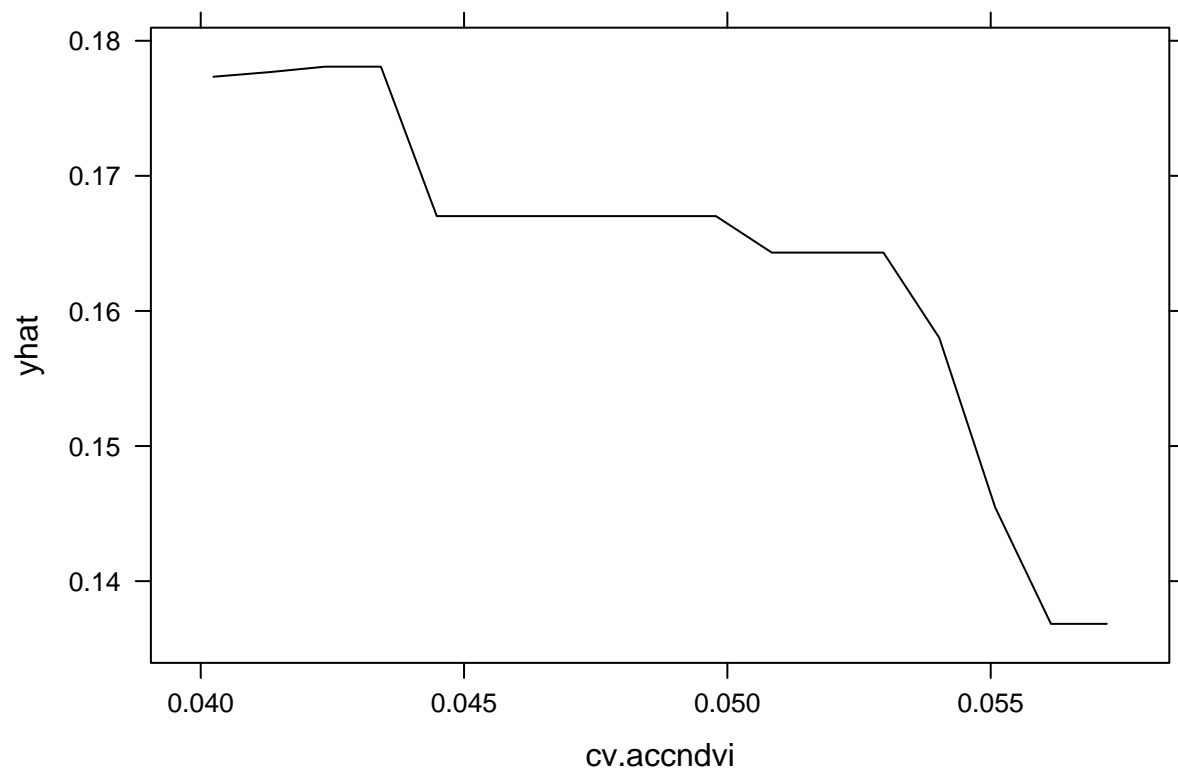
```
cor.test(predphi.st,cos(modvars.accndvi.phist$accndviphi.ts1))
```

```
##
## Pearson's product-moment correlation
##
## data: predphi.st and cos(modvars.accndvi.phist$accndviphi.ts1)
## t = 7.334, df = 41, p-value = 5.587e-09
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
##  0.5853958 0.8592385
## sample estimates:
##      cor
## 0.7532953
```

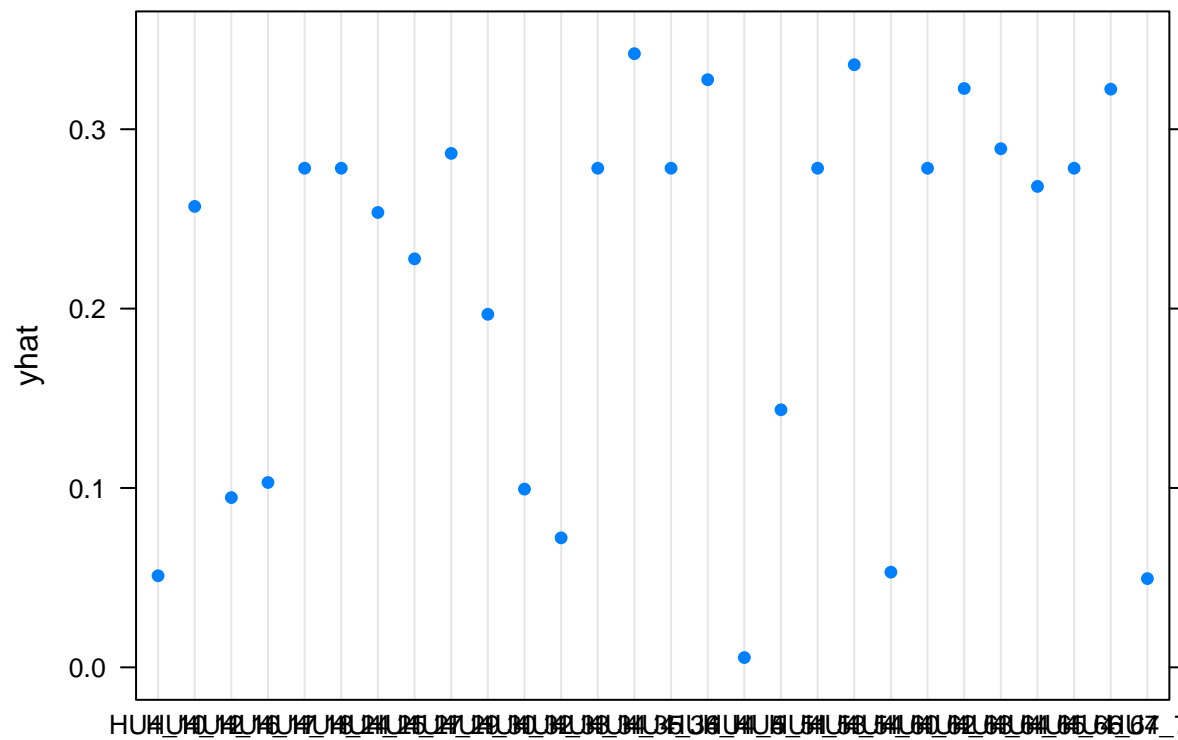
```
varimp.phi.st<-varimp(cforest.phi.st)
print(varimp.phi.st[order(varimp.phi.st,decreasing=TRUE)])
```

```
##      maxdepth      lake_area_ha lake_perim_meters      cv.accndvi
##  0.0832355040    0.0187470195    0.0068026866    -0.0005060651
##      chl_a      pct.ag      tsi.cat      hu4_zoneid
## -0.0053147437 -0.0070469587 -0.0110262249 -0.0314571005
```

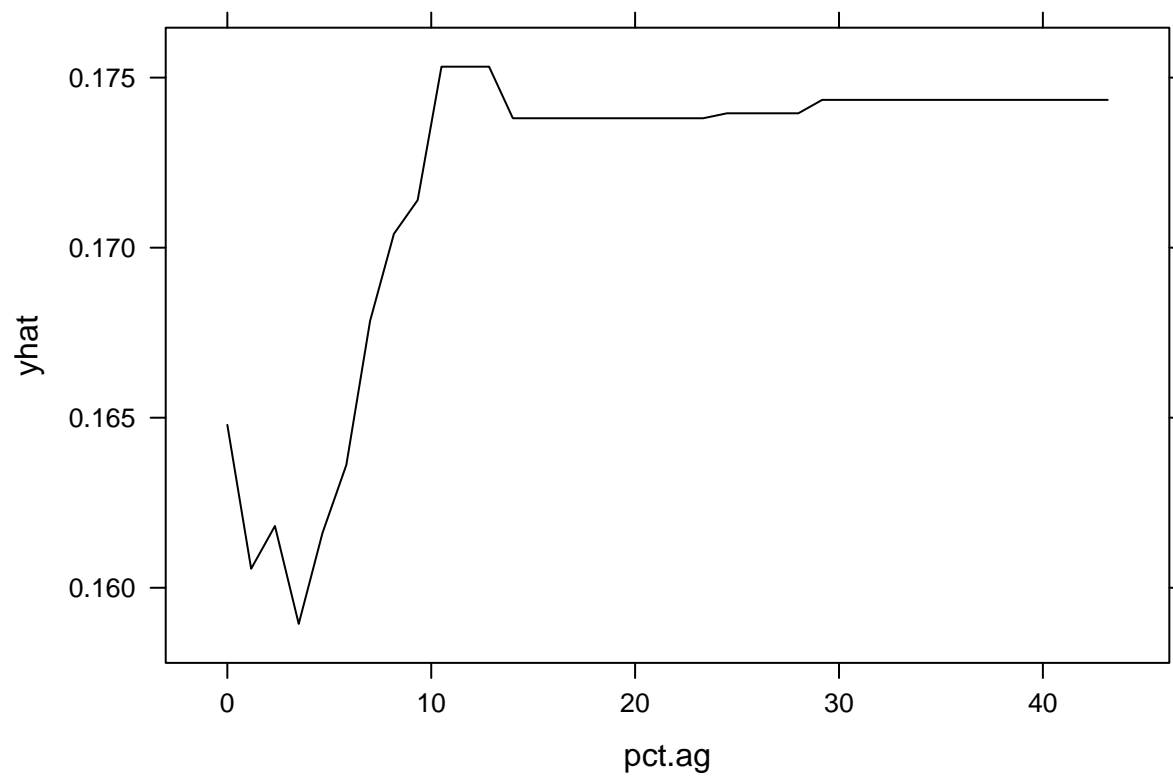
```
partial(cforest.phi.st, pred.var="cv.accndvi", train=modvars.accndvi.phist, type="regression", plot=T)
```



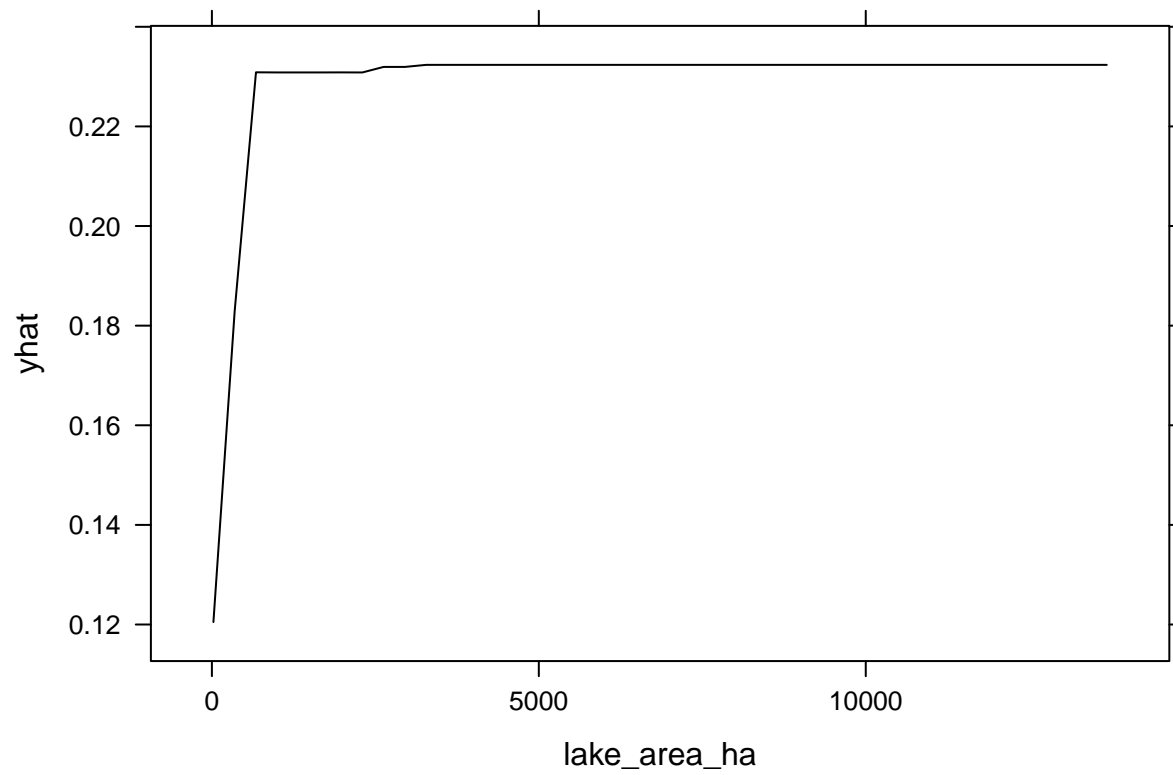
```
partial(cforest.phi.st, pred.var="hu4_zoneid", train=modvars.accdv.phist, type="regression", plot=T)
```



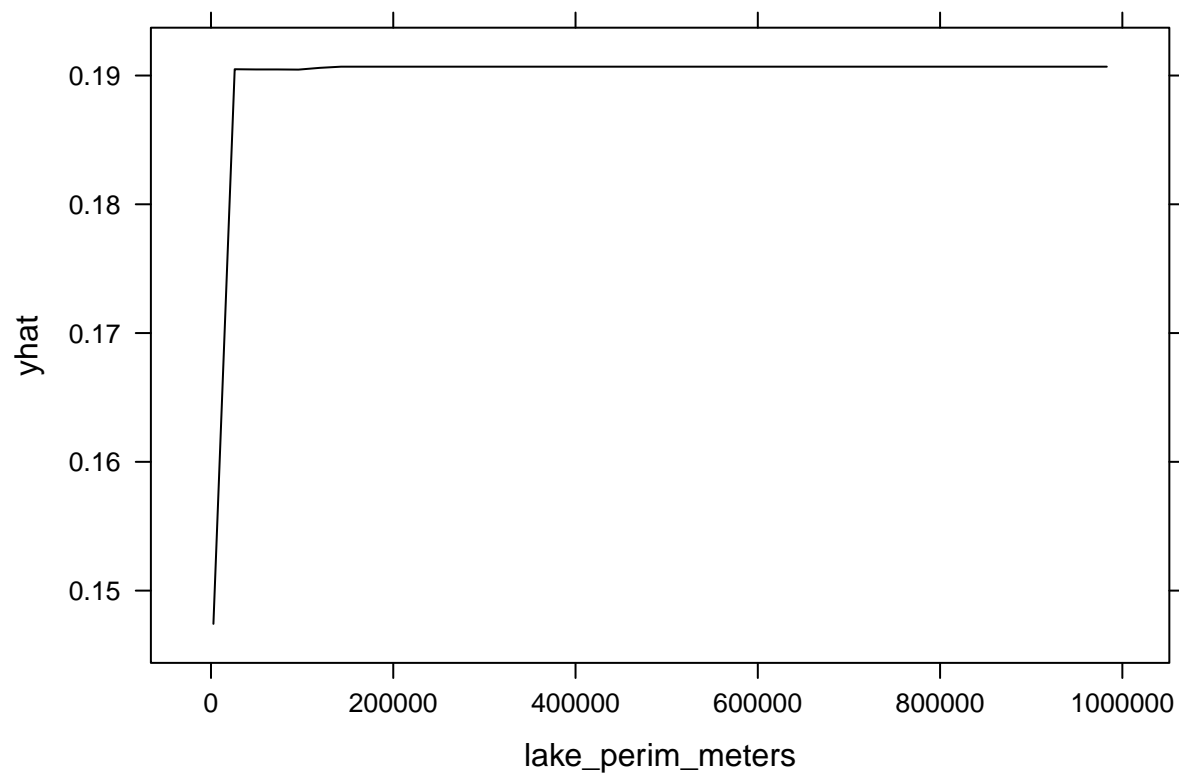
```
partial(cforest.phi.st, pred.var="pct.ag", train=modvars.accdv.phist, type="regression", plot=T)
```

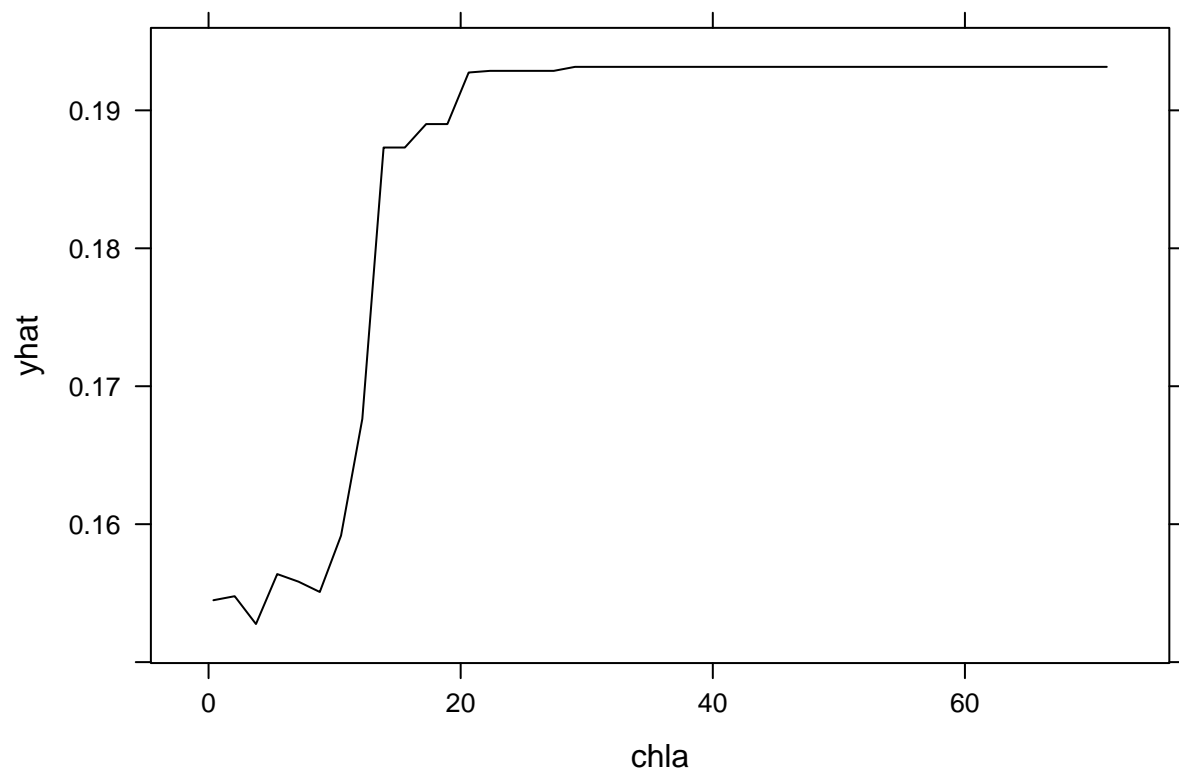
```
partial(cforest.phi.st, pred.var="lake_area_ha", train=modvars.accndvi.phist, type="regression", plot=T)
```



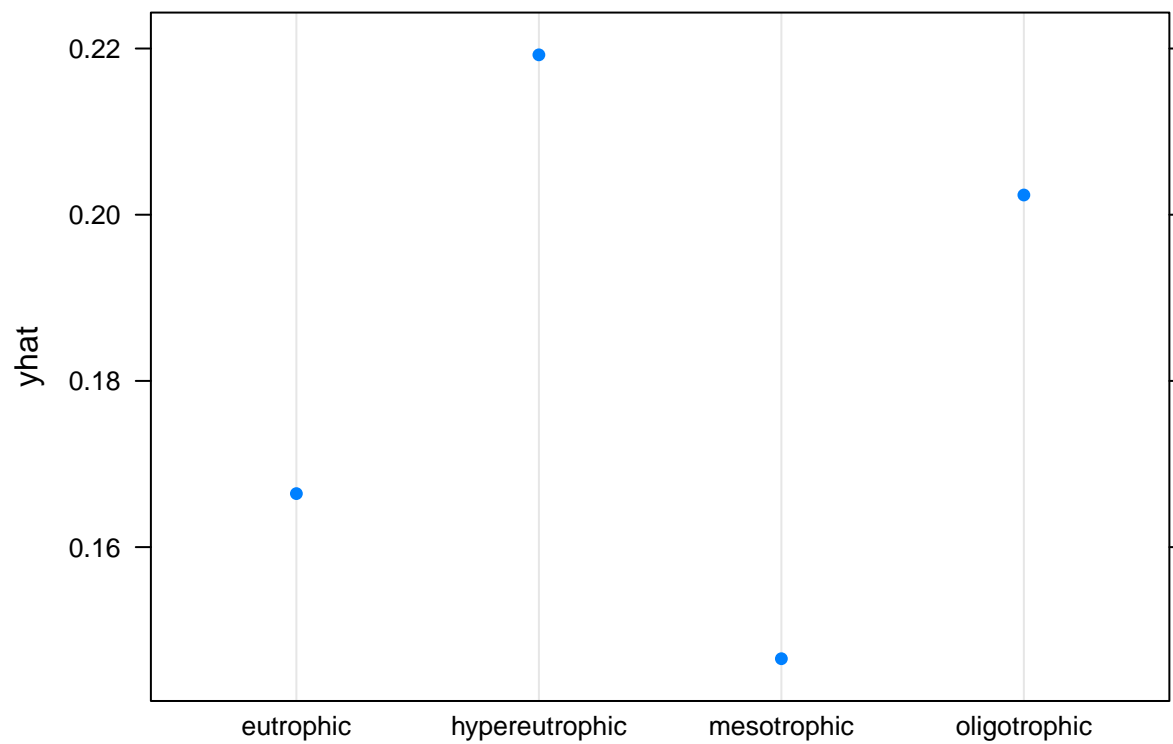
```
partial(cforest.phi.st, pred.var="lake_perim_meters", train=modvars.accndvi.phist, type="regression", p
```



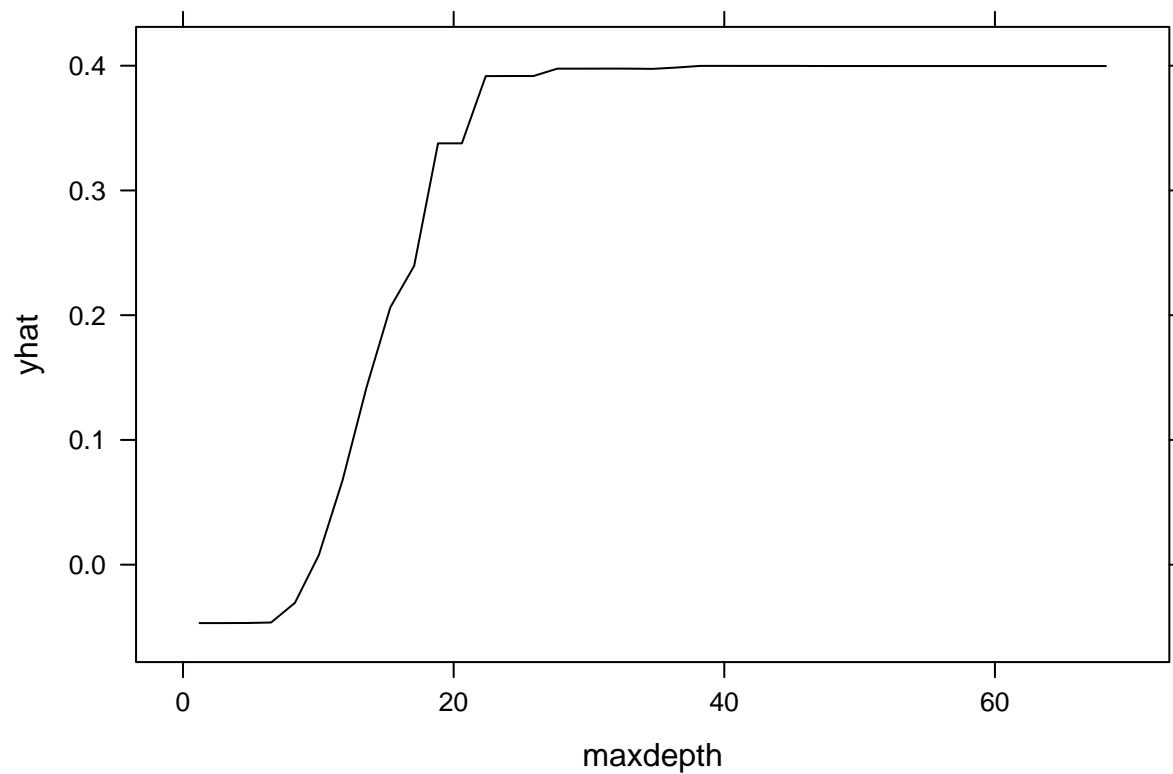
```
partial(cforest.phi.st, pred.var="chla", train=modvars.accndvi.phist, type="regression", plot=T)
```



```
partial(cforest.phi.st, pred.var="tsi.cat", train=modvars.accndvi.phist, type="regression", plot=T)
```

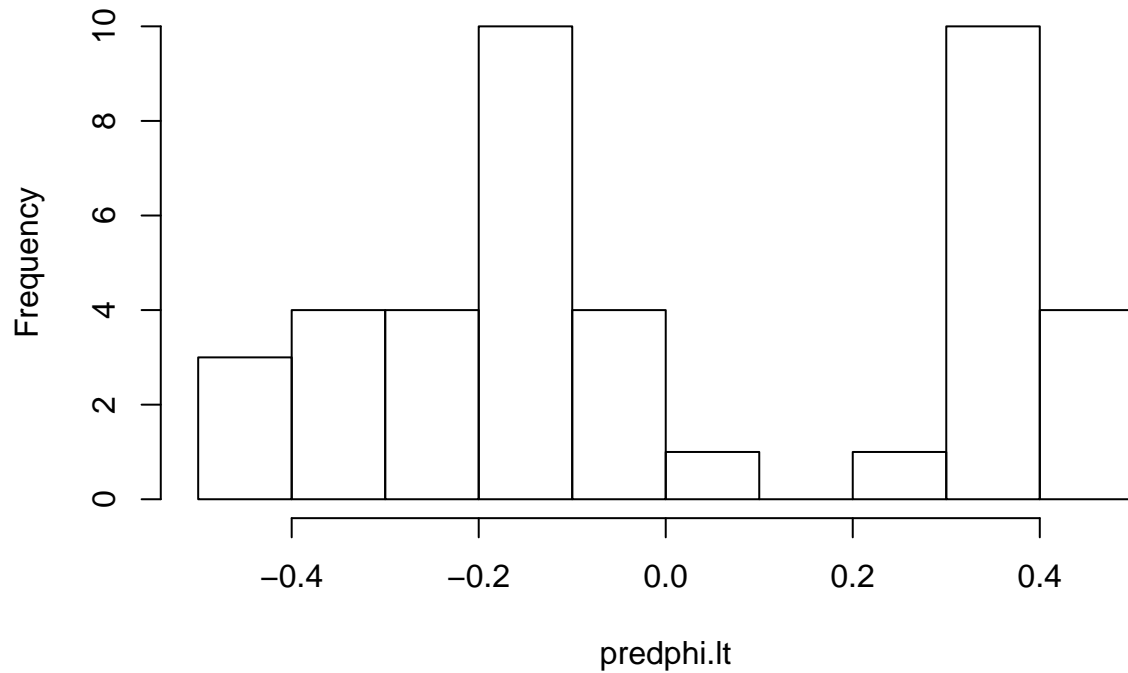


```
partial(cforest.phi.st, pred.var="maxdepth", train=modvars.accndvi.phist, type="regression", plot=T)
```



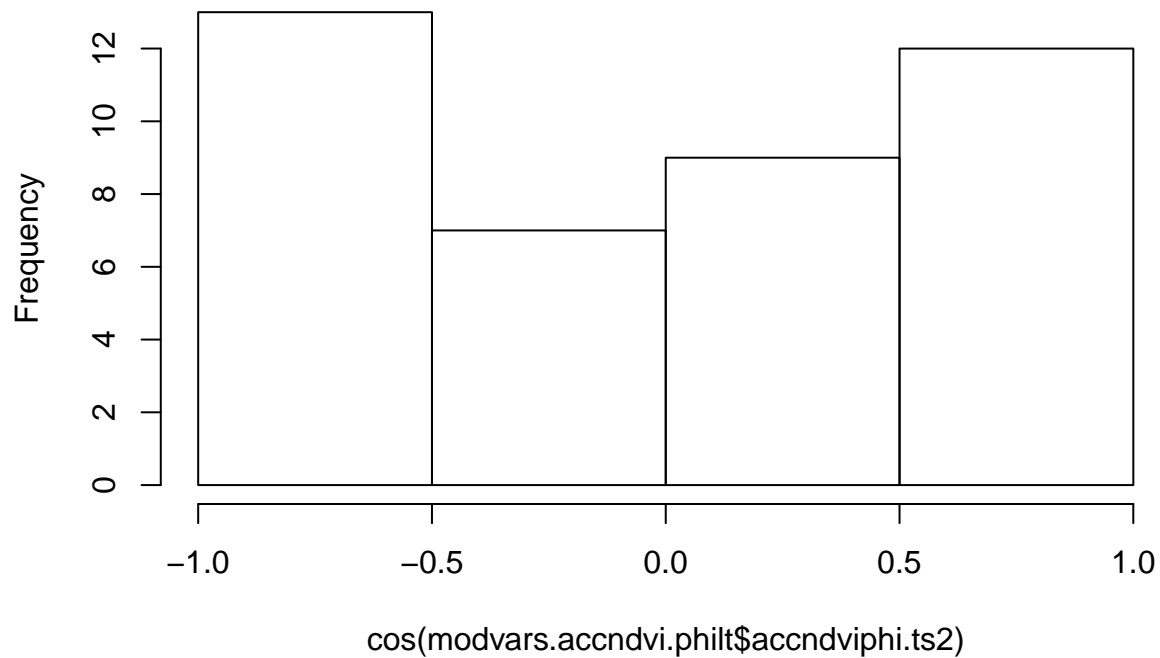
```
cforest.phi.lt<-cforest(cos(accndviphi.ts2) ~ lake_area_ha + lake_perim_meters + maxdepth + pct.ag + ch.
                        data=modvars.accndvi.philt,controls=cforest_control(ntree=30000))
predphi.lt<-predict(cforest.phi.lt, newdata=modvars.accndvi.philt)
hist(predphi.lt)
```

Histogram of predphi.lt



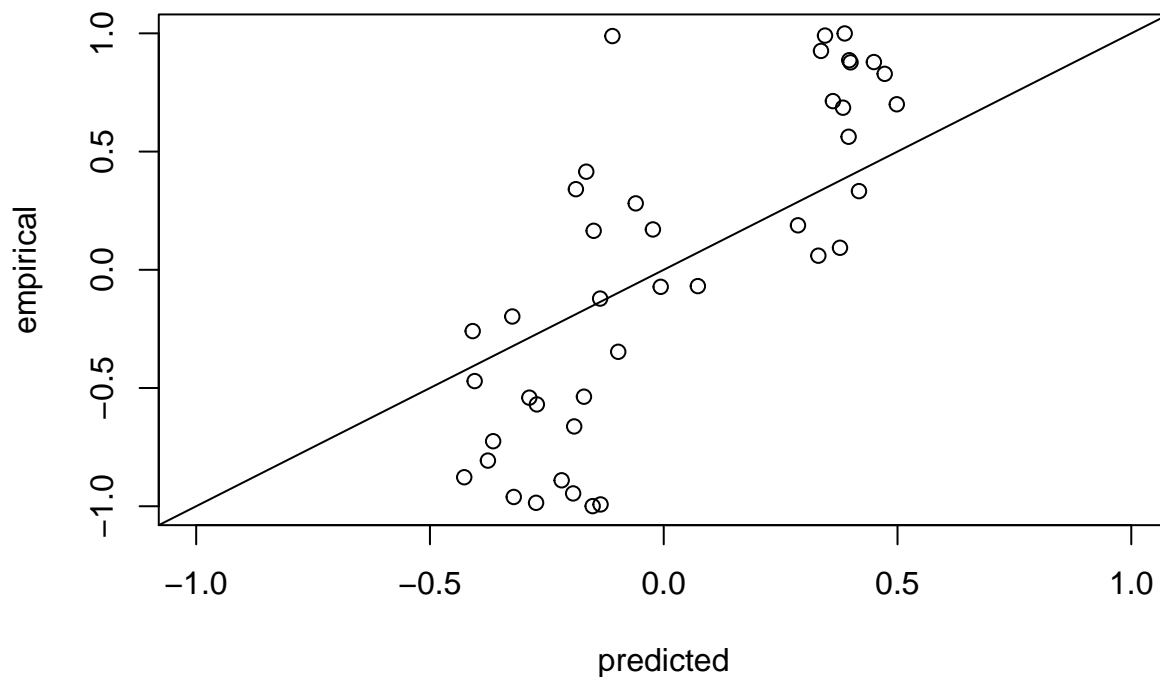
```
hist(cos(modvars.accndvi.philt$accndviphi.ts2))
```

Histogram of cos(modvars.accndvi.philt\$accndviphi.ts2)



```
plot(predphi.lt, cos(modvars.accndvi.philt$accndviphi.ts2), xlab="predicted", ylab="empirical", main="cos(modvars.accndvi.philt$accndviphi.ts2)")
xlim=c(-1,1), ylim=c(-1,1))
abline(a=0,b=1)
```

cos(phase), short ts



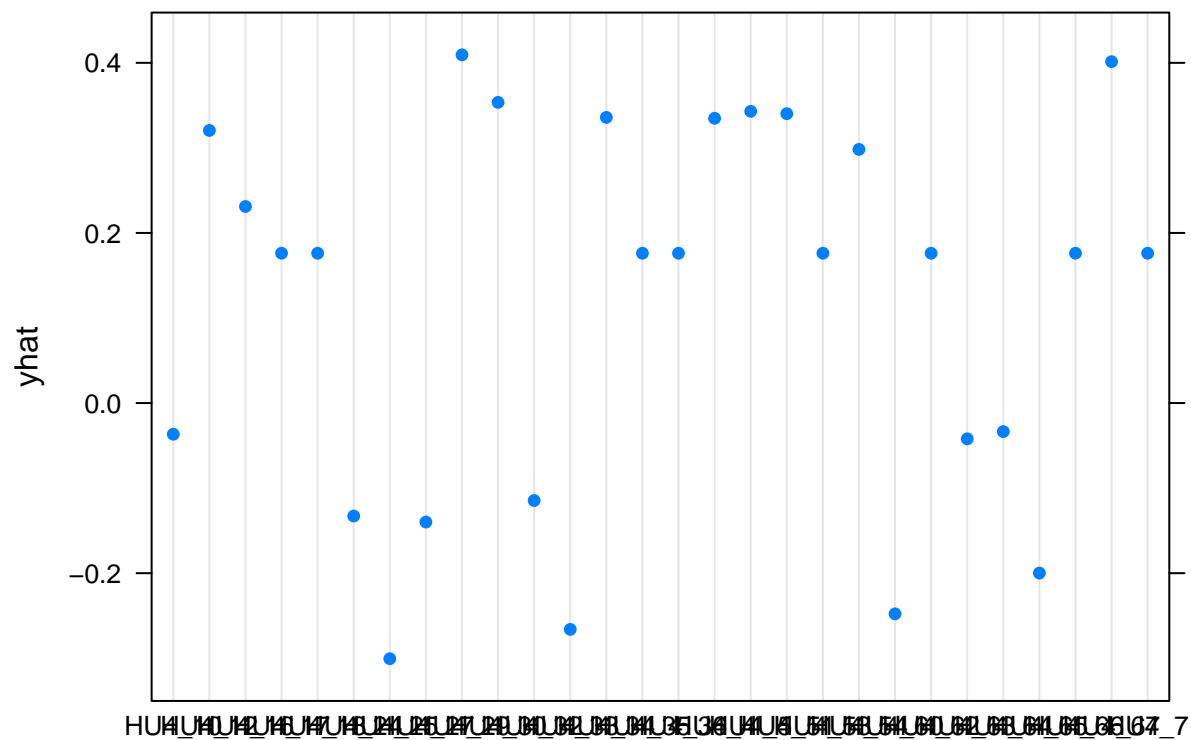
```
cor.test(predphi.lt,cos(modvars.accndvi.philt$accndviphi.ts2))
```

```
##
## Pearson's product-moment correlation
##
## data: predphi.lt and cos(modvars.accndvi.philt$accndviphi.ts2)
## t = 8.0993, df = 39, p-value = 6.98e-10
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
##  0.6402661 0.8841616
## sample estimates:
##      cor
## 0.7919247
```

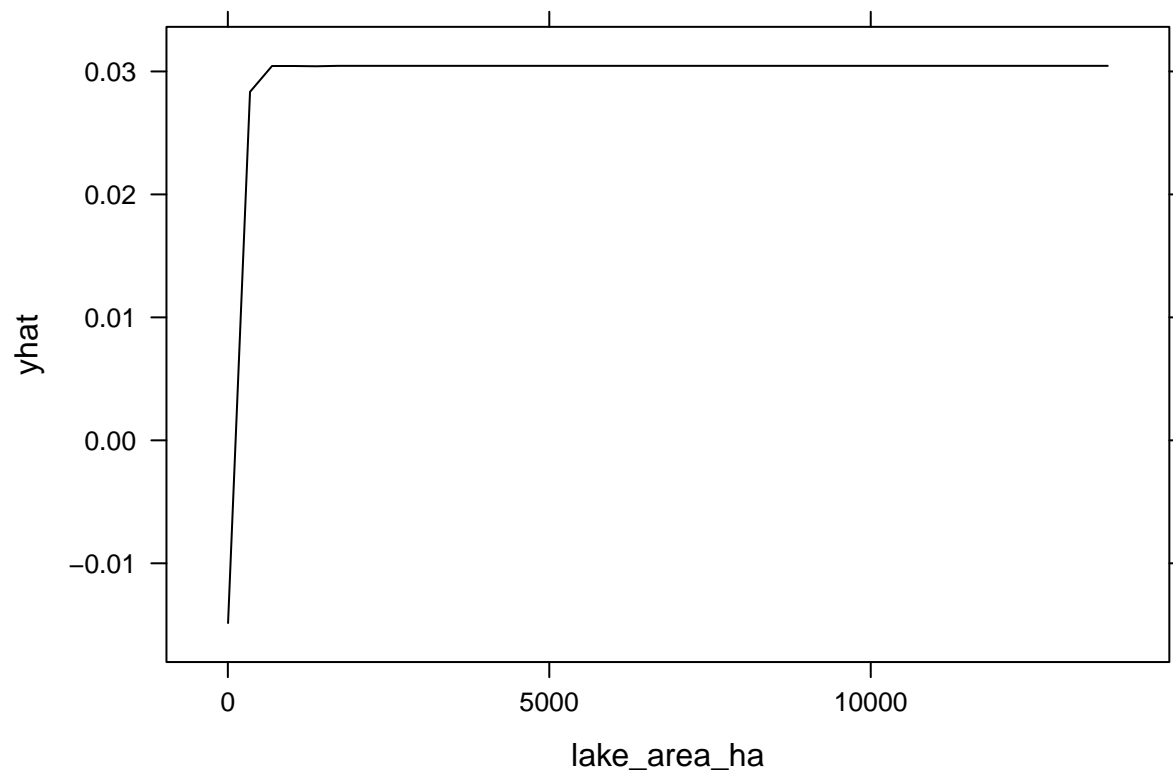
```
varimp.phi.lt<-varimp(cforest.phi.lt)
print(varimp.phi.lt[order(varimp.phi.lt, decreasing=TRUE)])
```

```
##      hu4_zoneid      lake_area_ha lake_perim_meters      cv.accndvi
##      0.131357442      0.006439426      0.001963842      -0.003344188
##      chl_a      maxdepth      pct.ag      tsi.cat
##      -0.003852368      -0.004980527      -0.007968329      -0.009684826
```

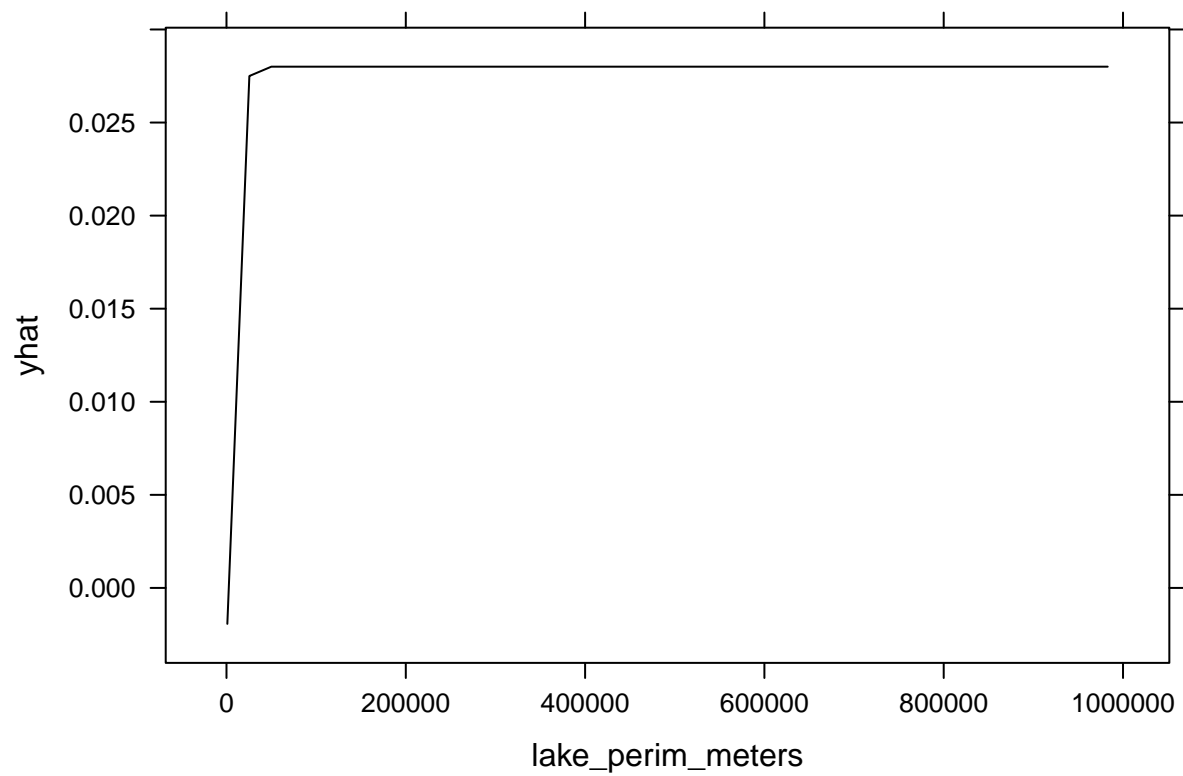
```
partial(cforest.phi.lt, pred.var="hu4_zoneid", train=modvars.accndvi.philt, type="regression", plot=T)
```



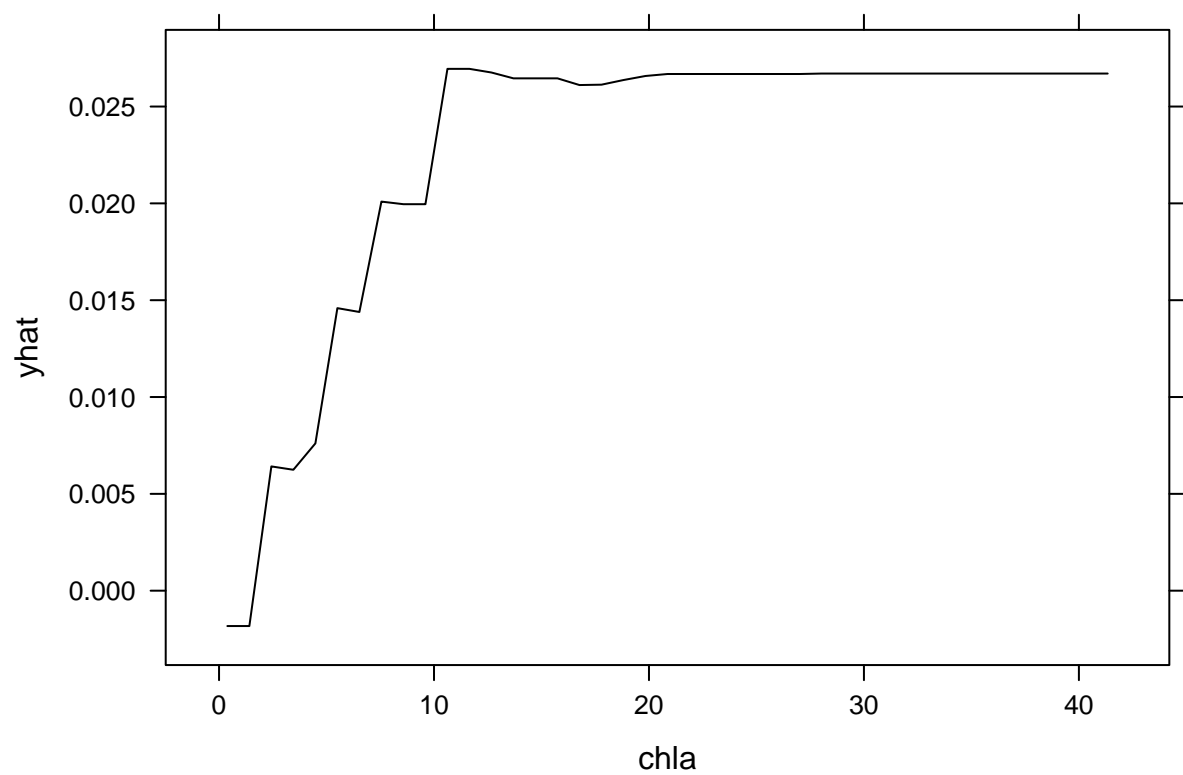
```
partial(cforest.phi.lt, pred.var="lake_area_ha", train=modvars.accndvi.philt, type="regression", plot=T)
```



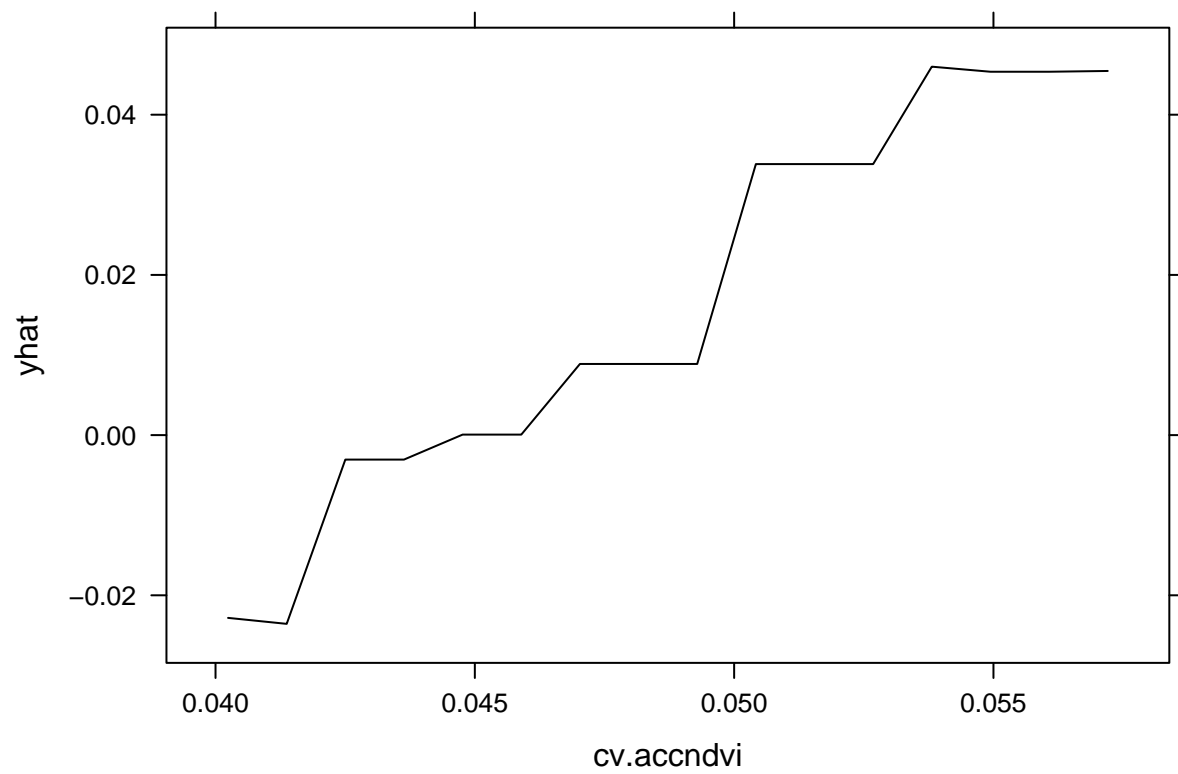
```
partial(cforest.phi.lt, pred.var="lake_perim_meters", train=modvars.accndvi.philt, type="regression", p
```



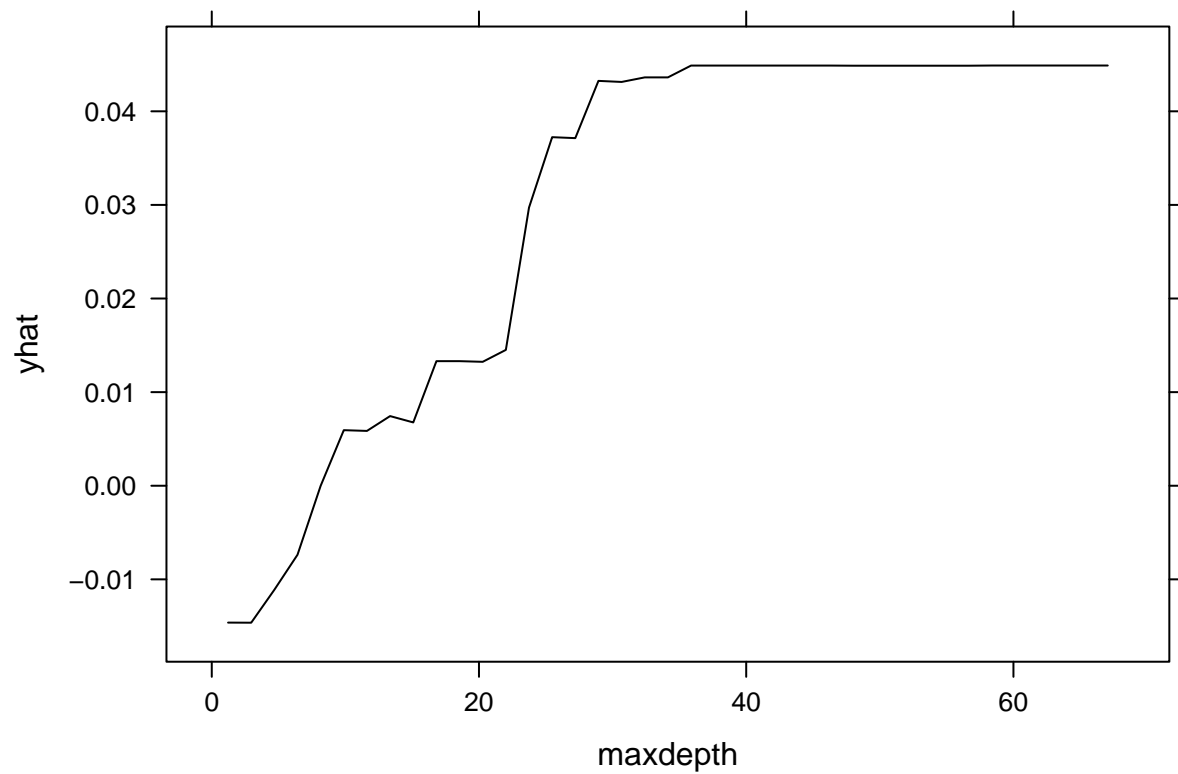
```
partial(cforest.phi.lt, pred.var="chla", train=modvars.accndvi.philt, type="regression", plot=T)
```



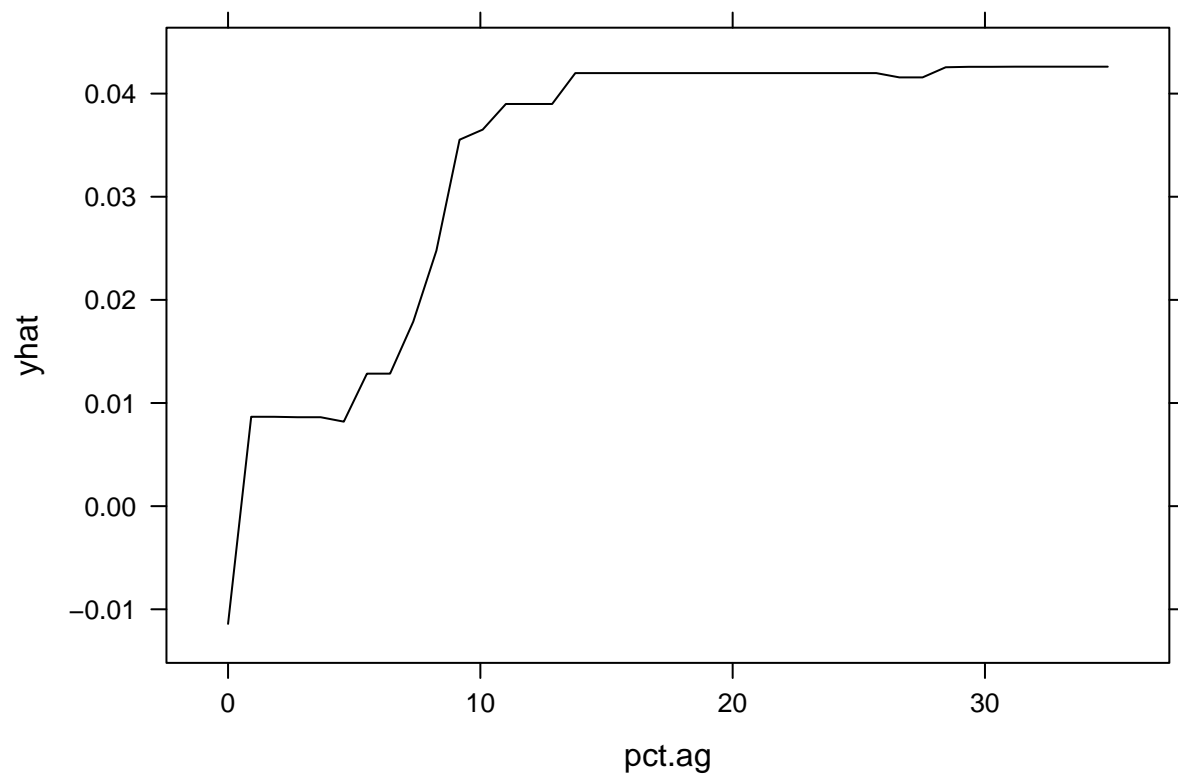
```
partial(cforest.phi.lt, pred.var="cv.accndvi", train=modvars.accndvi.philt, type="regression", plot=T)
```



```
partial(cforest.phi.lt, pred.var="maxdepth", train=modvars.accndvi.philt, type="regression", plot=T)
```



```
partial(cforest.phi.lt, pred.var="pct.ag", train=modvars.accndvi.philt, type="regression", plot=T)
```

```
partial(cforest.phi.lt, pred.var="tsi.cat", train=modvars.accndvi.philt, type="regression", plot=T)
```

