Assignment4\_Lobsters

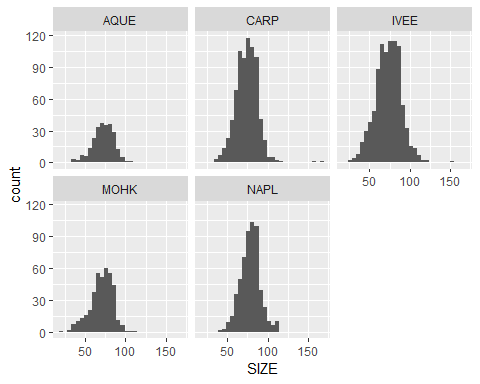
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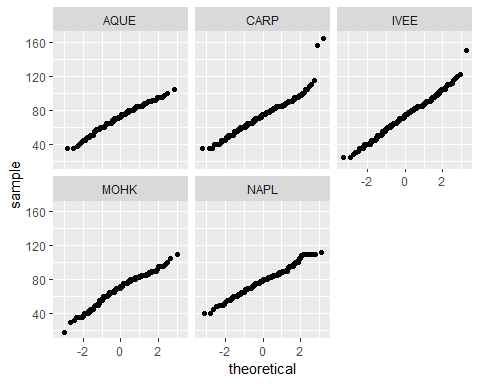
#2 Initial Data Editing  
lobster\_size\_edits <- lobster\_size %>%  
 filter(SIZE !=-99999) %>%  
 select(YEAR, SITE, SIZE, COUNT) # Edit Lobster Size data to remove -99999 data from size column and remove transect information  
   
  
lobster\_traps\_edits <- lobster\_traps %>%  
 filter(SITE=="AQUE" |SITE =="NAPL" |SITE=="MOHK" |SITE=="IVEE" |SITE=="CARP") %>%  
 select(YEAR, SITE, TRAPS)

#3 checking data for normality  
  
lobster\_size\_hist <- ggplot(lobster\_size\_edits, aes(x=SIZE)) +  
 geom\_histogram() +  
 facet\_wrap(~SITE)  
  
lobster\_size\_hist

## `stat\_bin()` using `bins = 30`. Pick better value with `binwidth`.

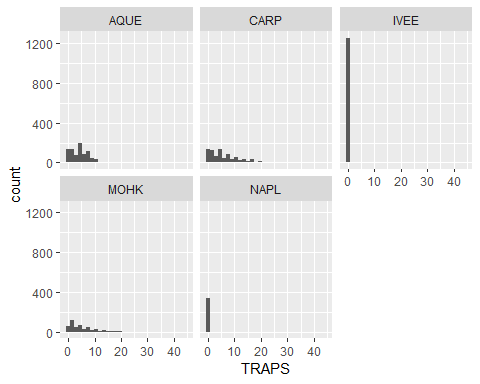


lobster\_size\_qq <- ggplot(lobster\_size\_edits, aes(sample=SIZE)) +  
 geom\_qq() +  
 facet\_wrap(~SITE)  
  
lobster\_size\_qq

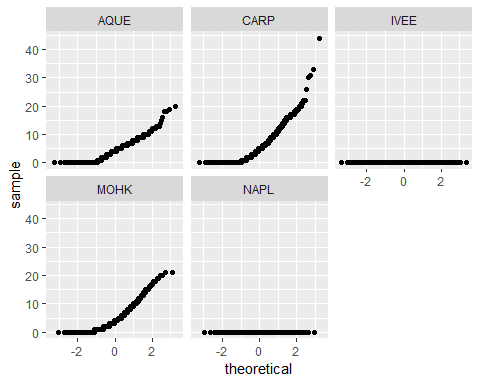


lobster\_traps\_hist <- ggplot(lobster\_traps\_edits, aes(x=TRAPS)) +  
 geom\_histogram() +  
 facet\_wrap(~SITE)  
  
lobster\_traps\_hist

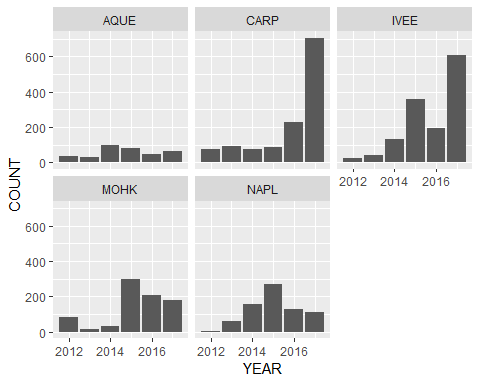
## `stat\_bin()` using `bins = 30`. Pick better value with `binwidth`.



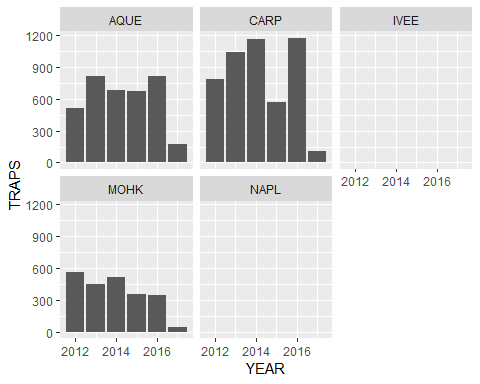
lobster\_traps\_qq <- ggplot(lobster\_traps\_edits, aes(sample=TRAPS)) +  
 geom\_qq() +  
 facet\_wrap(~SITE)  
  
lobster\_traps\_qq



#4 create graphs of abundance and fishing pressure from 2012-2017  
  
abundance\_col <- ggplot(lobster\_size\_edits, aes(x=YEAR, y=COUNT)) +  
 geom\_col() +  
 facet\_wrap(~SITE)  
  
abundance\_col



fishing\_pressure\_col <- ggplot(lobster\_traps\_edits, aes(x=YEAR, y=TRAPS)) +  
 geom\_col() +  
 facet\_wrap(~SITE)  
  
fishing\_pressure\_col



#5 put lobster size data into case format.  
  
lobster\_case\_format <- lobster\_size\_edits %>%  
 filter(YEAR=="2017") %>%  
 as.data.frame() %>%  
 expand.dft(freq="COUNT")

#6 run significance tests for lobster size  
lobster\_levene <- leveneTest(SIZE~SITE, data=lobster\_case\_format)  
lobster\_levene

## Levene's Test for Homogeneity of Variance (center = median)  
## Df F value Pr(>F)   
## group 4 8.3893 1.065e-06 \*\*\*  
## 1663   
## ---  
## Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

#Variences are not equal check to see if another need to do another test beside anova  
  
lobster\_aov <- aov(SIZE~SITE, data=lobster\_case\_format)  
summary(lobster\_aov)

## Df Sum Sq Mean Sq F value Pr(>F)   
## SITE 4 2355 588.6 3.424 0.0085 \*\*  
## Residuals 1663 285871 171.9   
## ---  
## Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

lobster\_tukeys <- TukeyHSD(lobster\_aov)  
lobster\_tukeys

## Tukey multiple comparisons of means  
## 95% family-wise confidence level  
##   
## Fit: aov(formula = SIZE ~ SITE, data = lobster\_case\_format)  
##   
## $SITE  
## diff lwr upr p adj  
## CARP-AQUE -1.6657352 -6.24294710 2.911477 0.8582355  
## IVEE-AQUE -2.4433772 -7.05292315 2.166169 0.5968998  
## MOHK-AQUE -1.8955224 -7.02720717 3.236162 0.8514711  
## NAPL-AQUE 2.3366205 -3.19311600 7.866357 0.7775633  
## IVEE-CARP -0.7776420 -2.76097123 1.205687 0.8216104  
## MOHK-CARP -0.2297872 -3.23309697 2.773523 0.9995765  
## NAPL-CARP 4.0023556 0.36042398 7.644287 0.0228728  
## MOHK-IVEE 0.5478548 -2.50450730 3.600217 0.9882889  
## NAPL-IVEE 4.7799976 1.09751057 8.462485 0.0037001  
## NAPL-MOHK 4.2321429 -0.08607271 8.550358 0.0579286