

# env\_and\_copies\_stats

This code runs uni-variate models on parameters of interest.

## Getting Started

### Libraries

### Data

## Cleaning

### Clean and Trim

```
dust<- dust0 %>%  
  filter(!is.na(date),  
         date %in% as.Date("2022-07-04"):as.Date("2022-07-19")) %>%  
  mutate(t7sum = log10(t7sum)) #Log transform Dust conc (as it is not normal)  
  
data<- data0 %>%  
  filter(date %in% as.Date("2022-07-04"):as.Date("2022-07-19"))
```

### Combine into one big data set

```
master0<-  
  dust %>% full_join(data) %>%  
  filter(hr_cst %in% "01") #While this is technically not for hour 1, this gets ride of al
```

Joining, by = "date"

```
master<- master0[c(1,8,14,34)] #Select columns of interest

#View new dataframe
master
```

```
# A tibble: 40 x 4
  date      t7sum site      copies_mL
  <date>    <dbl> <chr>      <dbl>
1 2022-07-04 -0.886 <NA>        NA
2 2022-07-05 -0.796 <NA>        NA
3 2022-07-06 -0.854 <NA>        NA
4 2022-07-07 -1      Blind Oso  25462.
5 2022-07-07 -1      Canals    21658.
6 2022-07-07 -1      Gulf      442.
7 2022-07-08 -1      Blind Oso  33586.
8 2022-07-08 -1      Canals    14290.
9 2022-07-08 -1      Gulf      2670.
10 2022-07-09 -0.886 Blind Oso  29093.
# ... with 30 more rows
```

## Make Site DF

Need to include 7/5 and 7/6 for previous dust data (will be used for lag analysis)

```
bo<- master %>%
  filter(site %in% "Blind Oso"| date %in% c( as.Date("2022-07-05"), as.Date("2022-07-06")))

c2<- master %>%
  filter(site %in% "Canals"| date %in% c( as.Date("2022-07-05"), as.Date("2022-07-06")))

rd<- master %>%
  filter(site %in% "Gulf"| date %in% c(as.Date("2022-07-05"), as.Date("2022-07-06")))
```

## DUST

Add Lag from 2\_cca.qmd

```

bo2<- bo %>% mutate(
  lag =Lag(bo$t7sum, shift = 1)) %>% #Make lag from ccf
filter(between(date, as.Date('2022-07-07'), as.Date('2022-07-19')))

c22<- c2%>% mutate(
  lag =Lag(c2$t7sum, shift = 2)) %>% #Make lag from ccf
filter(between(date, as.Date('2022-07-07'), as.Date('2022-07-19')))

rd2<- rd%>% mutate(
  lag =Lag(rd$t7sum, shift = 1)) %>% #Make lag from ccf
filter(between(date, as.Date('2022-07-07'), as.Date('2022-07-19')))

```

## Plot

```

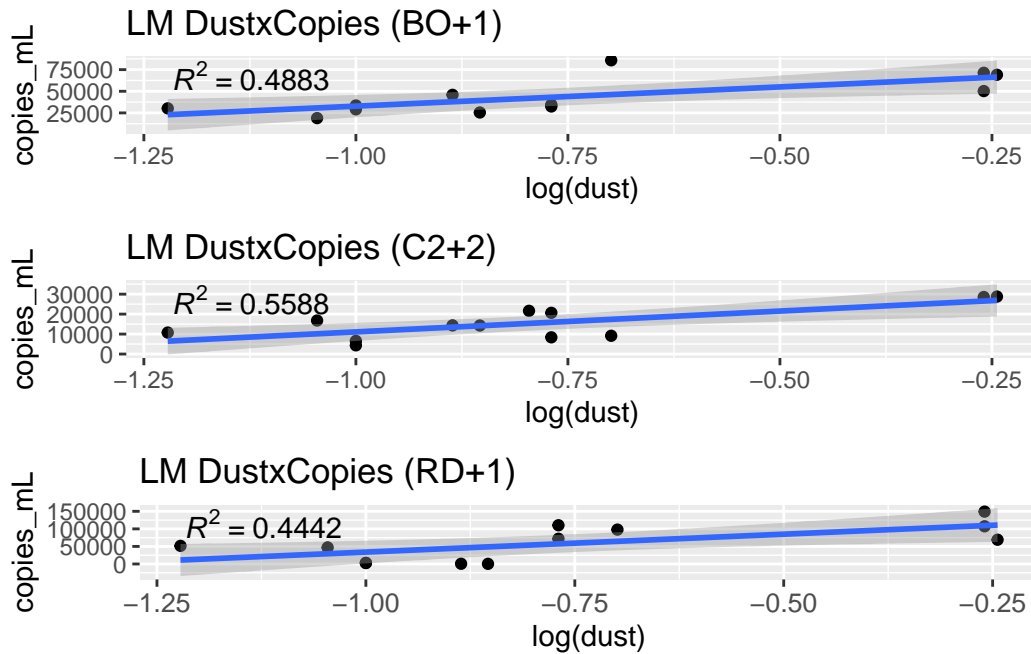
fig1<-
  bo2 %>% ggplot(aes(x = lag, y = copies_mL)) +
  geom_point() +
  stat_smooth(method = "lm")+
  stat_poly_eq(aes(x = lag, y = copies_mL),
               rr.digits = 4)+
  labs(title = "LM DustxCopies (B0+1)",
       x = "log(dust)")
fig2<-
  c22 %>% ggplot(aes(x = lag, y = copies_mL)) +
  geom_point() +
  stat_smooth(method = "lm")+
  stat_poly_eq(aes(x = lag, y = copies_mL),
               rr.digits = 4)+
  labs(title = "LM DustxCopies (C2+2)",
       x = "log(dust)")
fig3<-
  rd2 %>% ggplot(aes(lag, copies_mL)) +
  geom_point() +
  stat_smooth(method = "lm")+
  stat_poly_eq(aes(x = lag, y = copies_mL),
               rr.digits = 4)+
  labs(title = "LM DustxCopies (RD+1)",
       x = "log(dust)")

```

## Stac

```
ggarrange(fig1, fig2, fig3 + font("x.text", size = 10), ncol = 1, nrow = 3)
```

```
`geom_smooth()` using formula = 'y ~ x'
`geom_smooth()` using formula = 'y ~ x'
`geom_smooth()` using formula = 'y ~ x'
```



**Null Model:**

**5-Fold Cross Validation**

```
fold_bo <- vfold_cv(bo2, v = 5, repeats = 5) #Data was too small to stratify (<20)
fold_c2 <- vfold_cv(c22, v = 5, repeats = 5) #Data was too small to stratify (<20)
fold_rd <- vfold_cv(rd2, v = 5, repeats = 5) #Data was too small to stratify (<20)
```

**Create Null Model**

```
#Create Null Recipe for Growth at all three sites
null_bo<- recipe(copies_mL ~ 1, data = bo2)
null_c2<- recipe(copies_mL ~ 1, data = c22)
```

```

null_rd<- recipe(copies_mL ~ 1, data = rd2)

#Set up linear model
lm_mod<- linear_reg() %>%
  set_engine("lm") %>%
  set_mode("regression")

#Workflow that adds recipe to model for three sites
null_wflow_bo<-
  workflow() %>%
  add_model(lm_mod) %>%
  add_recipe(null_bo)

null_wflow_c2<-
  workflow() %>%
  add_model(lm_mod) %>%
  add_recipe(null_c2)

null_wflow_rd<-
  workflow() %>%
  add_model(lm_mod) %>%
  add_recipe(null_rd)

#Use workflow to fit model to data set
null_fit_bo<-
  fit_resamples(null_wflow_bo, resamples = fold_bo)

```

```

! Fold1, Repeat1: internal: A correlation computation is required, but `estimate` is constant
! Fold2, Repeat1: internal: A correlation computation is required, but `estimate` is constant
! Fold3, Repeat1: internal: A correlation computation is required, but `estimate` is constant
! Fold4, Repeat1: internal: A correlation computation is required, but `estimate` is constant
! Fold5, Repeat1: internal: A correlation computation is required, but `estimate` is constant
! Fold1, Repeat2: internal: A correlation computation is required, but `estimate` is constant
! Fold2, Repeat2: internal: A correlation computation is required, but `estimate` is constant

```

! Fold3, Repeat2: internal: A correlation computation is required, but `estimate` is constant

! Fold4, Repeat2: internal: A correlation computation is required, but `estimate` is constant

! Fold5, Repeat2: internal: A correlation computation is required, but `estimate` is constant

! Fold1, Repeat3: internal: A correlation computation is required, but `estimate` is constant

! Fold2, Repeat3: internal: A correlation computation is required, but `estimate` is constant

! Fold3, Repeat3: internal: A correlation computation is required, but `estimate` is constant

! Fold4, Repeat3: internal: A correlation computation is required, but `estimate` is constant

! Fold5, Repeat3: internal: A correlation computation is required, but `estimate` is constant

! Fold1, Repeat4: internal: A correlation computation is required, but `estimate` is constant

! Fold2, Repeat4: internal: A correlation computation is required, but `estimate` is constant

! Fold3, Repeat4: internal: A correlation computation is required, but `estimate` is constant

! Fold4, Repeat4: internal: A correlation computation is required, but `estimate` is constant

! Fold5, Repeat4: internal: A correlation computation is required, but `estimate` is constant

! Fold1, Repeat5: internal: A correlation computation is required, but `estimate` is constant

! Fold2, Repeat5: internal: A correlation computation is required, but `estimate` is constant

! Fold3, Repeat5: internal: A correlation computation is required, but `estimate` is constant

! Fold4, Repeat5: internal: A correlation computation is required, but `estimate` is constant

! Fold5, Repeat5: internal: A correlation computation is required, but `estimate` is constant

```
null_fit_c2<-  
  fit_resamples(null_wflow_c2, resamples = fold_c2)
```

```
! Fold1, Repeat1: internal: A correlation computation is required, but `estimate` is constant  
! Fold2, Repeat1: internal: A correlation computation is required, but `estimate` is constant  
! Fold3, Repeat1: internal: A correlation computation is required, but `estimate` is constant  
! Fold4, Repeat1: internal: A correlation computation is required, but `estimate` is constant  
! Fold5, Repeat1: internal: A correlation computation is required, but `estimate` is constant  
! Fold1, Repeat2: internal: A correlation computation is required, but `estimate` is constant  
! Fold2, Repeat2: internal: A correlation computation is required, but `estimate` is constant  
! Fold3, Repeat2: internal: A correlation computation is required, but `estimate` is constant  
! Fold4, Repeat2: internal: A correlation computation is required, but `estimate` is constant  
! Fold5, Repeat2: internal: A correlation computation is required, but `estimate` is constant  
! Fold1, Repeat3: internal: A correlation computation is required, but `estimate` is constant  
! Fold2, Repeat3: internal: A correlation computation is required, but `estimate` is constant  
! Fold3, Repeat3: internal: A correlation computation is required, but `estimate` is constant  
! Fold4, Repeat3: internal: A correlation computation is required, but `estimate` is constant  
! Fold5, Repeat3: internal: A correlation computation is required, but `estimate` is constant  
! Fold1, Repeat4: internal: A correlation computation is required, but `estimate` is constant  
! Fold2, Repeat4: internal: A correlation computation is required, but `estimate` is constant
```

! Fold3, Repeat4: internal: A correlation computation is required, but `estimate` is constant  
! Fold4, Repeat4: internal: A correlation computation is required, but `estimate` is constant  
! Fold5, Repeat4: internal: A correlation computation is required, but `estimate` is constant  
! Fold1, Repeat5: internal: A correlation computation is required, but `estimate` is constant  
! Fold2, Repeat5: internal: A correlation computation is required, but `estimate` is constant  
! Fold3, Repeat5: internal: A correlation computation is required, but `estimate` is constant  
! Fold4, Repeat5: internal: A correlation computation is required, but `estimate` is constant  
! Fold5, Repeat5: internal: A correlation computation is required, but `estimate` is constant

```
null_fit_rd<-  
  fit_resamples(null_wflow_rd, resamples = fold_rd)
```

! Fold1, Repeat1: internal: A correlation computation is required, but `estimate` is constant  
! Fold2, Repeat1: internal: A correlation computation is required, but `estimate` is constant  
! Fold3, Repeat1: internal: A correlation computation is required, but `estimate` is constant  
! Fold4, Repeat1: internal: A correlation computation is required, but `estimate` is constant  
! Fold5, Repeat1: internal: A correlation computation is required, but `estimate` is constant  
! Fold1, Repeat2: internal: A correlation computation is required, but `estimate` is constant  
! Fold2, Repeat2: internal: A correlation computation is required, but `estimate` is constant  
! Fold3, Repeat2: internal: A correlation computation is required, but `estimate` is constant  
! Fold4, Repeat2: internal: A correlation computation is required, but `estimate` is constant



! Fold5, Repeat2: internal: A correlation computation is required, but `estimate` is constant

! Fold1, Repeat3: internal: A correlation computation is required, but `estimate` is constant

! Fold2, Repeat3: internal: A correlation computation is required, but `estimate` is constant

! Fold3, Repeat3: internal: A correlation computation is required, but `estimate` is constant

! Fold4, Repeat3: internal: A correlation computation is required, but `estimate` is constant

! Fold5, Repeat3: internal: A correlation computation is required, but `estimate` is constant

! Fold1, Repeat4: internal: A correlation computation is required, but `estimate` is constant

! Fold2, Repeat4: internal: A correlation computation is required, but `estimate` is constant

! Fold3, Repeat4: internal: A correlation computation is required, but `estimate` is constant

! Fold4, Repeat4: internal: A correlation computation is required, but `estimate` is constant

! Fold5, Repeat4: internal: A correlation computation is required, but `estimate` is constant

! Fold1, Repeat5: internal: A correlation computation is required, but `estimate` is constant

! Fold2, Repeat5: internal: A correlation computation is required, but `estimate` is constant

! Fold3, Repeat5: internal: A correlation computation is required, but `estimate` is constant

! Fold4, Repeat5: internal: A correlation computation is required, but `estimate` is constant

! Fold5, Repeat5: internal: A correlation computation is required, but `estimate` is constant

## Get Metrics for Null

### RMSE

```

null_met_bo <-
  collect_metrics(null_fit_bo)

null_met_c2 <-
  collect_metrics(null_fit_c2)

null_met_rd <-
  collect_metrics(null_fit_rd)

null_met_bo

```

```

# A tibble: 2 x 6
  .metric .estimator  mean     n std_err .config
  <chr>   <chr>      <dbl> <int>   <dbl> <chr>
1 rmse    standard 20995.    25  1677. Preprocessor1_Model1
2 rsq     standard   NaN      0    NA  Preprocessor1_Model1

```

```

null_met_c2

```

```

# A tibble: 2 x 6
  .metric .estimator  mean     n std_err .config
  <chr>   <chr>      <dbl> <int>   <dbl> <chr>
1 rmse    standard  8449.    25   475. Preprocessor1_Model1
2 rsq     standard   NaN      0    NA  Preprocessor1_Model1

```

```

null_met_rd

```

```

# A tibble: 2 x 6
  .metric .estimator  mean     n std_err .config
  <chr>   <chr>      <dbl> <int>   <dbl> <chr>
1 rmse    standard 49889.    25  3928. Preprocessor1_Model1
2 rsq     standard   NaN      0    NA  Preprocessor1_Model1

```

```

rmnb<- 20927.53
rmnc<- 7934.59
rmnr<- 49831.58

```

## AIC

```
bo_null_mod<-
  lm(copies_mL ~ 1, data = bo2)
AIC(bo_null_mod)
```

```
[1] 275.9331
```

```
c2_null_mod<-
  lm(copies_mL ~ 1, data = c22)
AIC(c2_null_mod)
```

```
[1] 253.1876
```

```
rd_null_mod<-
  lm(copies_mL ~ 1, data = rd2)
AIC(rd_null_mod)
```

```
[1] 296.9821
```

```
aicnb<- 275.93
aicnc<- 253.19
aicnr<- 296.98
```

We will come back to these metrics once we run the models on our actual data.

## Actual Data for Model

### Create Linear Model

#### BLIND OSO:

```
#Create Recipe for Growth
growth_recipe_bo<- recipe(copies_mL ~ lag, data = bo2)

#Set up linear model
lm_mod<- linear_reg() %>%
  set_engine("lm") %>%
  set_mode("regression")
```

```

#Workflow that adds recipe to model
Growth_wflow<-
  workflow() %>%
  add_model(lm_mod) %>%
  add_recipe(growth_recipe_bo)

#Use workflow to fit model to data set
growth_fit_bo<- Growth_wflow %>%
  fit(data = bo2)

#View as Tibble
growth_fit_bo %>%
  extract_fit_parsnip() %>%
  tidy()

```

```

# A tibble: 2 x 5
  term      estimate std.error statistic  p.value
<chr>      <dbl>     <dbl>     <dbl>    <dbl>
1 (Intercept) 77277.    11730.      6.59 0.0000617
2 lag        44471.    14395.      3.09 0.0115

```

## CANALS:

```

#Create Recipe for Growth
growth_recipe_c2<- recipe(copies_mL ~ lag, data = c22)

#Set up linear model
lm_mod<- linear_reg() %>%
  set_engine("lm") %>%
  set_mode("regression")

#Workflow that adds recipe to model
Growth_wflow<-
  workflow() %>%
  add_model(lm_mod) %>%
  add_recipe(growth_recipe_c2)

#Use workflow to fit model to data set
growth_fit_c2<- Growth_wflow %>%
  fit(data = c22)

```

```
#View as Tibble
growth_fit_c2 %>%
  extract_fit_parsnip() %>%
  tidy()
```

```
# A tibble: 2 x 5
  term      estimate std.error statistic  p.value
<chr>      <dbl>    <dbl>    <dbl>    <dbl>
1 (Intercept) 31954.    4942.     6.47 0.0000721
2 lag        20859.    5861.     3.56 0.00519
```

## GULF:

```
#Create Recipe for Growth
growth_recipe_rd<- recipe(copies_mL ~ lag, data = rd2)

#Set up linear model
lm_mod<- linear_reg() %>%
  set_engine("lm") %>%
  set_mode("regression")

#Workflow that adds recipe to model
Growth_wflow<-
  workflow() %>%
  add_model(lm_mod) %>%
  add_recipe(growth_recipe_rd)

#Use workflow to fit model to data set
growth_fit_rd<- Growth_wflow %>%
  fit(data = rd2)

#View as Tibble
growth_fit_rd %>%
  extract_fit_parsnip() %>%
  tidy()
```

```
# A tibble: 2 x 5
  term      estimate std.error statistic  p.value
<chr>      <dbl>    <dbl>    <dbl>    <dbl>
1 (Intercept) 135834.    29387.     4.62 0.000947
```

```
2 lag          101960.    36063.    2.83 0.0179
```

## Metrics

### RMSE and R2

```
aug_test_bo <- augment(growth_fit_bo, bo2)
rmse <- aug_test_bo %>% rmse(truth = copies_mL, .pred)
rsq <- aug_test_bo %>% rsq(truth = copies_mL, .pred)
bo_metrics<- full_join(rmse, rsq)
```

```
Joining, by = c(".metric", ".estimator", ".estimate")
```

```
bo_metrics
```

```
# A tibble: 2 x 3
  .metric .estimator .estimate
  <chr>   <chr>       <dbl>
1 rmse    standard    14423.
2 rsq     standard      0.488
```

```
aug_test_c2 <- augment(growth_fit_c2, c22)
rmse <- aug_test_c2 %>% rmse(truth = copies_mL, .pred)
rsq <- aug_test_c2 %>% rsq(truth = copies_mL, .pred)
c2_metrics<- full_join(rmse, rsq)
```

```
Joining, by = c(".metric", ".estimator", ".estimate")
```

```
c2_metrics
```

```
# A tibble: 2 x 3
  .metric .estimator .estimate
  <chr>   <chr>       <dbl>
1 rmse    standard     5191.
2 rsq     standard      0.559
```

```

aug_test_rd <- augment(growth_fit_rd, rd2)
rmse <- aug_test_rd %>% rmse(truth = copies_mL, .pred)
rsq <- aug_test_rd %>% rsq(truth = copies_mL, .pred)
rd_metrics<- full_join(rmse, rsq)

```

Joining, by = c(".metric", ".estimator", ".estimate")

```
rd_metrics
```

```

# A tibble: 2 x 3
  .metric .estimator .estimate
  <chr>   <chr>       <dbl>
1 rmse    standard    36132.
2 rsq     standard      0.444

```

```

rmb<- 14422.91
rmc<- 5191.22
rmr<- 36132.11

```

## AIC

Approaches based on such selection criteria essentially try to guess how the model would perform if it were to be fit to new data, without actually trying to do it (in contrast to CV). For AIC, a rule of thumb is that a difference of 10 between 2 models is meaningful.

```

bo_mod<- lm(copies_mL ~ lag, data = bo2)
AIC(bo_mod)

```

```
[1] 269.8923
```

```

c2_mod<- lm(copies_mL ~ lag, data = c22)
AIC(c2_mod)

```

```
[1] 245.3679
```

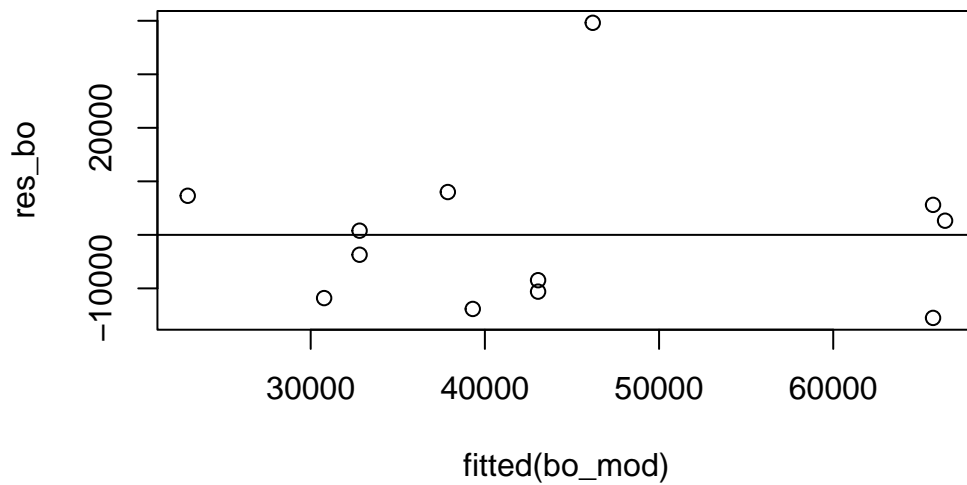
```
rd_mod<- lm(copies_mL ~ lag, data = rd2)
AIC(rd_mod)
```

```
[1] 291.933
```

```
aicb<- 269.89
aicc<- 245.37
aicr<- 291.93
```

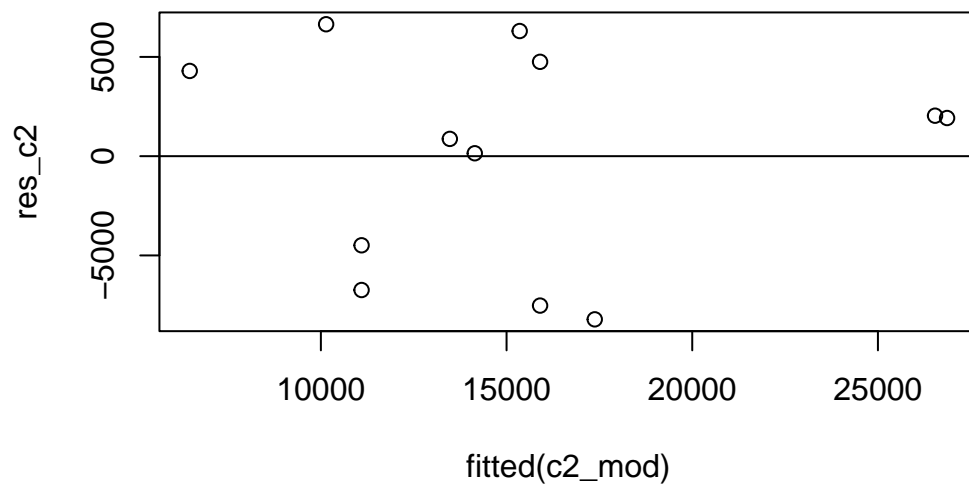
### Plot Residuals

```
res_bo<- resid(bo_mod)
plot(fitted(bo_mod), res_bo)
abline(0,0)
```

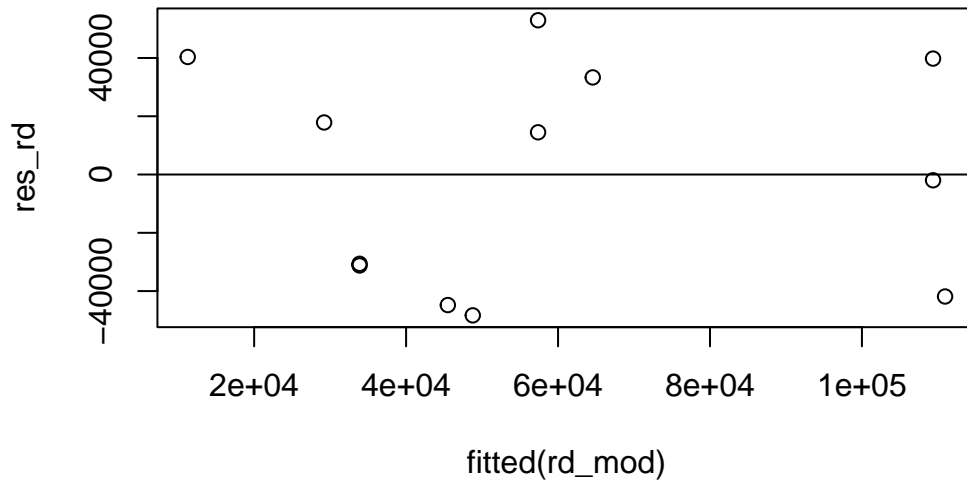


```
res_c2<- resid(c2_mod)
plot(fitted(c2_mod), res_c2)
abline(0,0)
```





```
res_rd<- resid(rd_mod)
plot(fitted(rd_mod), res_rd)
abline(0,0)
```



## Conclusion

Site	Model Type	RMSE	AIC
BO	Null	20472.82	275.9331
BO	Actual	14422.91	269.8923
C2	Null	8472.898	253.1876
C2	Actual	5191.22	245.3679
RD	Null	50386.56	296.9821
RD	Actual	36132.11	291.933

AIC and RMSE are lower (better performance) than the null.

## Make Table

```
kable(data.frame(Model = c("Null RMSE", "Actual RMSE", "Null AIC ", "Actual AIC"),
  Blind_Oso = c(rmnb, rmb, aicnb, aicb),
  Canals = c(rmnc, rmc, aicnc, aicc),
  Gulf = c(rmnr, rmr, aicnr, aicr)),
  caption = "Linear Regression Model Metrics")
```

Table 2: Linear Regression Model Metrics

Model	Blind_Oso	Canals	Gulf
Null RMSE	20927.53	7934.59	49831.58
Actual RMSE	14422.91	5191.22	36132.11
Null AIC	275.93	253.19	296.98
Actual AIC	269.89	245.37	291.93

## OTHER ENVIRONMENTAL PARAMETERS

### Create New Master Data set

```

master2<-
  master0 %>%
    select(copies_mL,sal,temp,site, amm, orthop, doc, don, tn, do_mgl, nn, sil, din_dip, tdn)

#Make Site Specific DF
b<-
  master2 %>%
    filter(site %in% "Blind Oso")

c<-
  master2 %>%
    filter(site %in% "Canals")

r<-
  master2 %>%
    filter(site %in% "Gulf")

```

### Plot

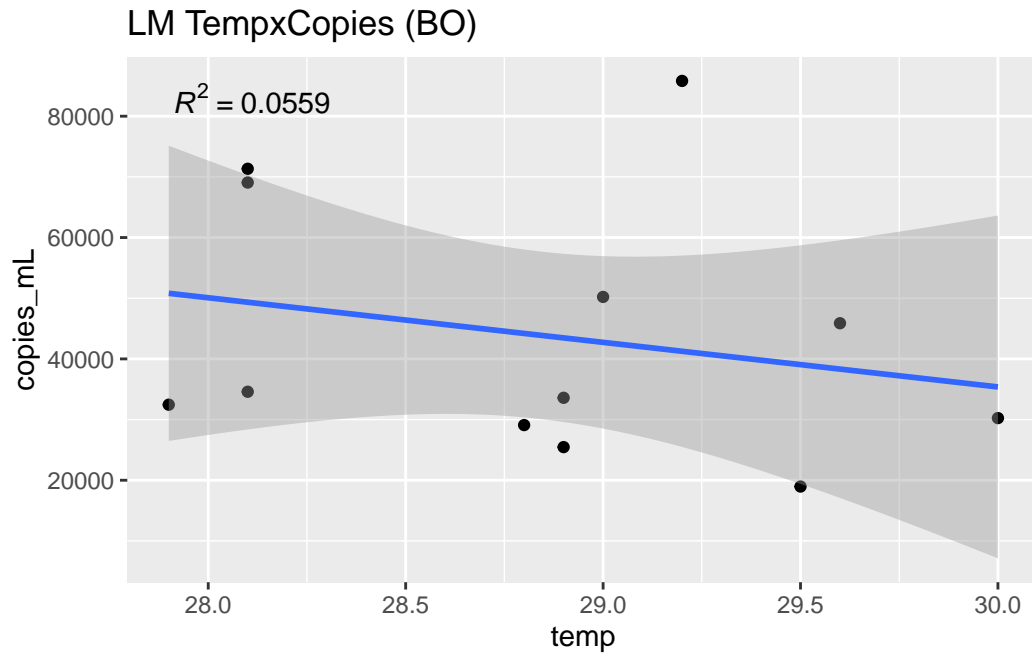
#### Temperature

```

b %>% ggplot(aes(x = temp, y = copies_mL)) +
  geom_point() +
  stat_smooth(method = "lm")+
  stat_poly_eq(aes(x = temp, y = copies_mL),
    rr.digits = 4) +
  labs(title = "LM TempxCopies (B0)")

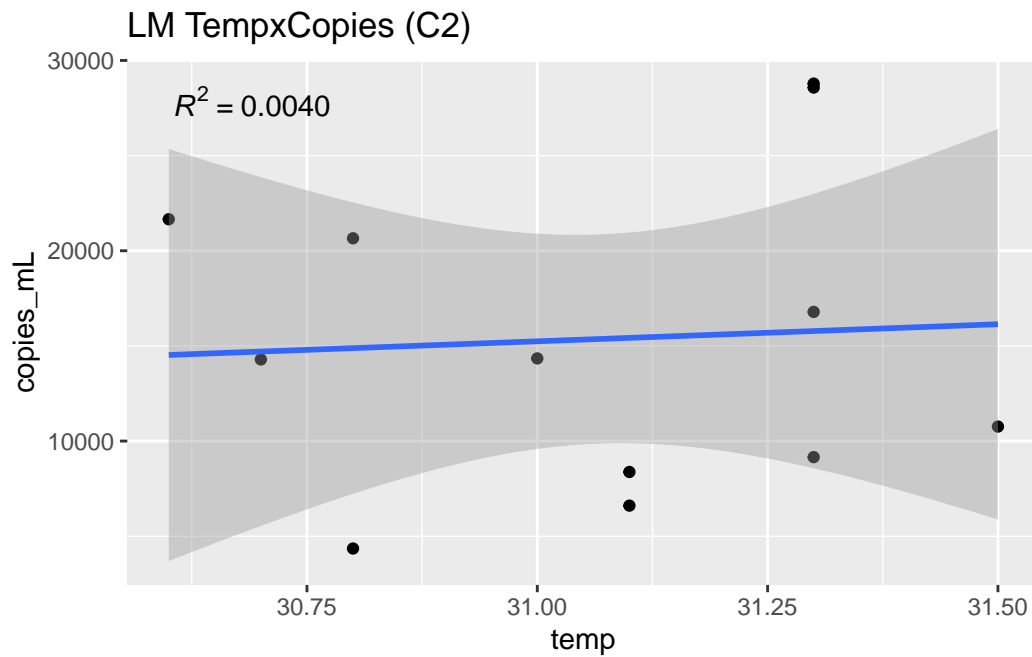
```

```
`geom_smooth()` using formula = 'y ~ x'
```



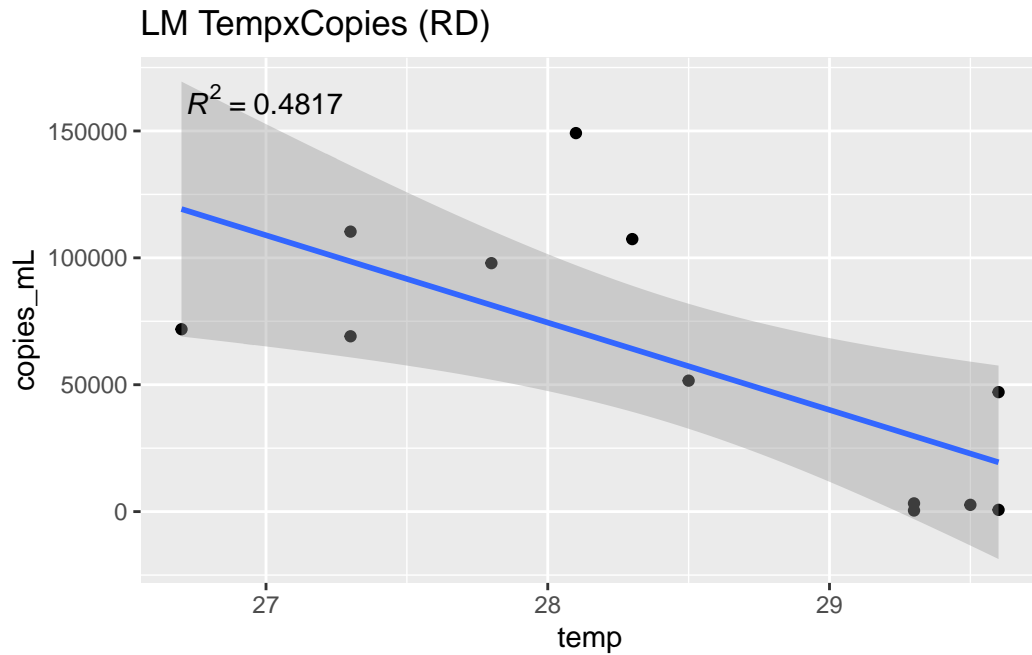
```
c %>% ggplot(aes(x = temp, y = copies_mL)) +  
  geom_point() +  
  stat_smooth(method = "lm")+  
  stat_poly_eq(aes(x = temp, y = copies_mL),  
    rr.digits = 4) +  
  labs(title = "LM TempxCopies (C2)")
```

```
`geom_smooth()` using formula = 'y ~ x'
```



```
r %>% ggplot(aes(x = temp, y = copies_mL)) +
  geom_point() +
  stat_smooth(method = "lm")+
  stat_poly_eq(aes(x = temp, y = copies_mL),
               rr.digits = 4) +
  labs(title = "LM TempxCopies (RD)")
```

`geom\_smooth()` using formula = 'y ~ x'

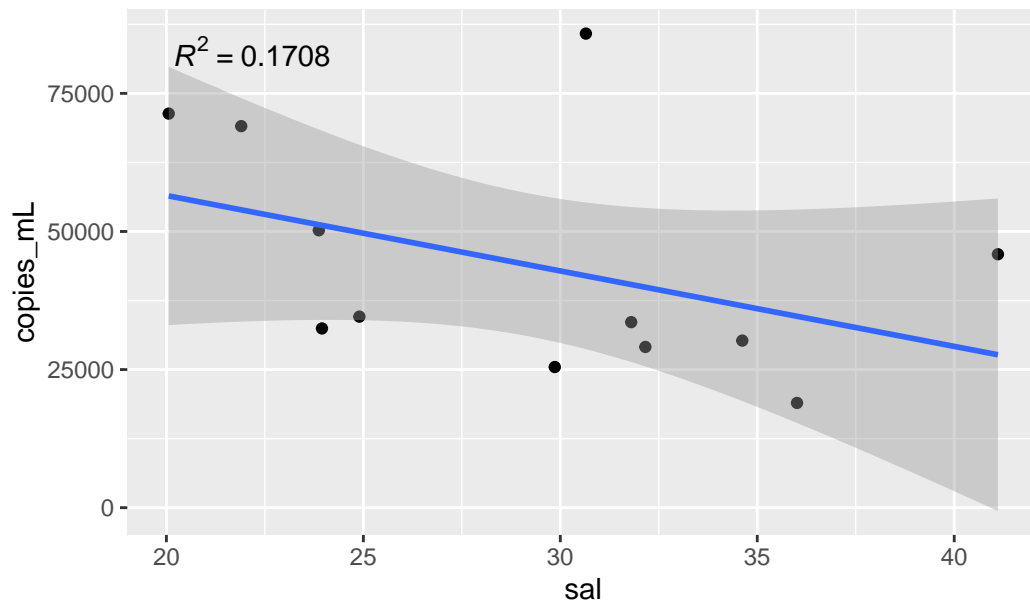


### Salinity

```
b %>% ggplot(aes(x = sal, y = copies_mL)) +
  geom_point() +
  stat_smooth(method = "lm")+
  stat_poly_eq(aes(x = sal, y = copies_mL),
    rr.digits = 4) +
  labs(title = "LM salxCopies (B0)")
```

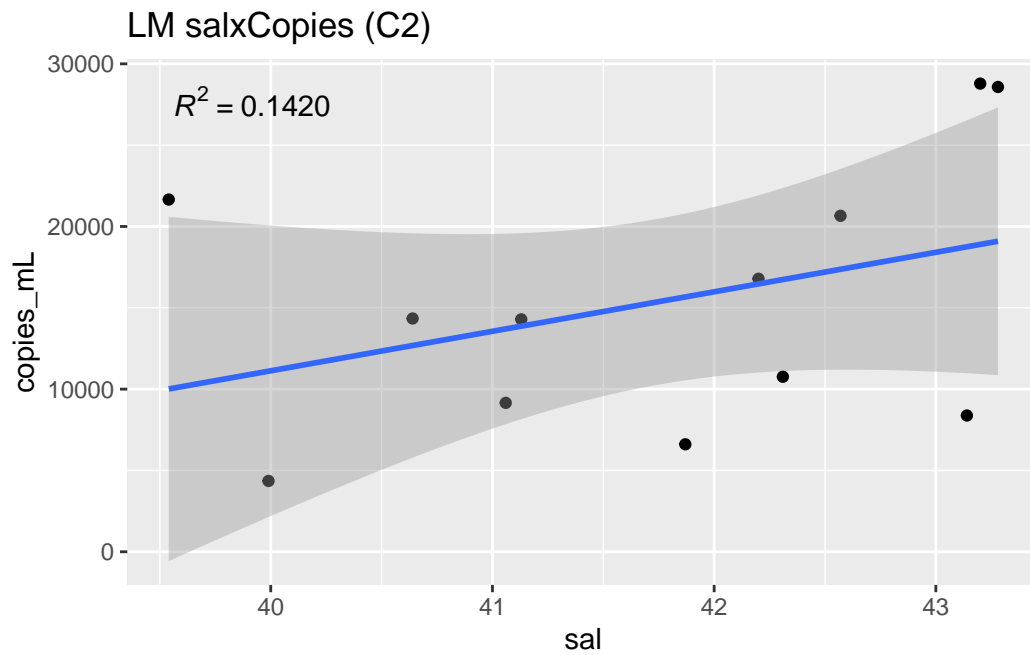
`geom\_smooth()` using formula = 'y ~ x'

LM salxCopies (BO)



```
c %>% ggplot(aes(x = sal, y = copies_mL)) +
  geom_point() +
  stat_smooth(method = "lm")+
  stat_poly_eq(aes(x = sal, y = copies_mL),
               rr.digits = 4) +
  labs(title = "LM salxCopies (C2)")
```

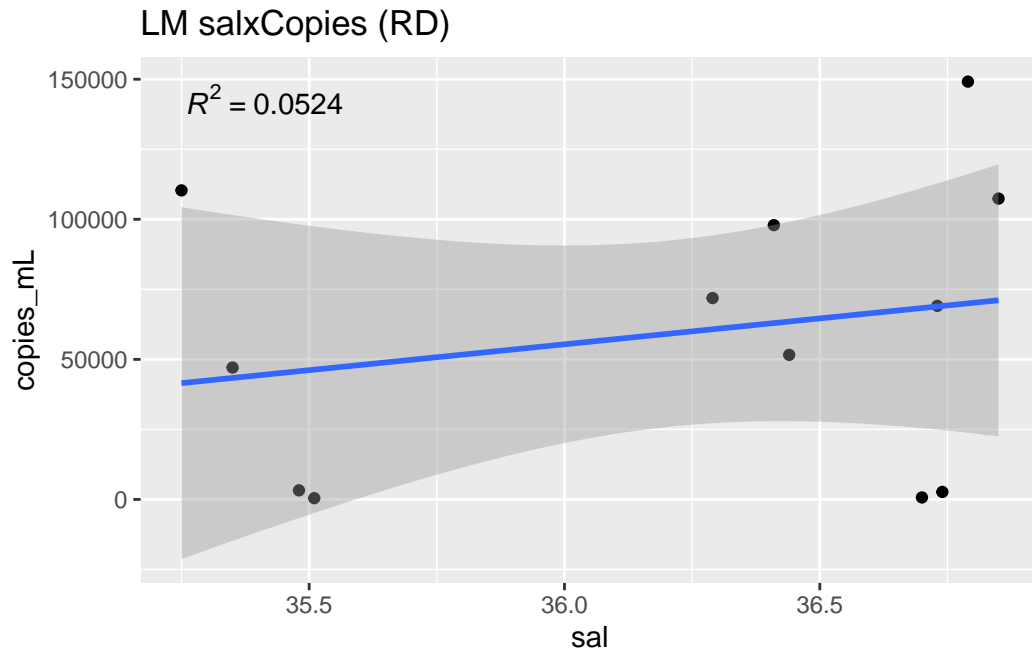
`geom\_smooth()` using formula = 'y ~ x'



```
r %>% ggplot(aes(x = sal, y = copies_mL)) +
  geom_point() +
  stat_smooth(method = "lm")+
  stat_poly_eq(aes(x = sal, y = copies_mL),
               rr.digits = 4) +
  labs(title = "LM salxCopies (RD)")
```

`geom\_smooth()` using formula = 'y ~ x'



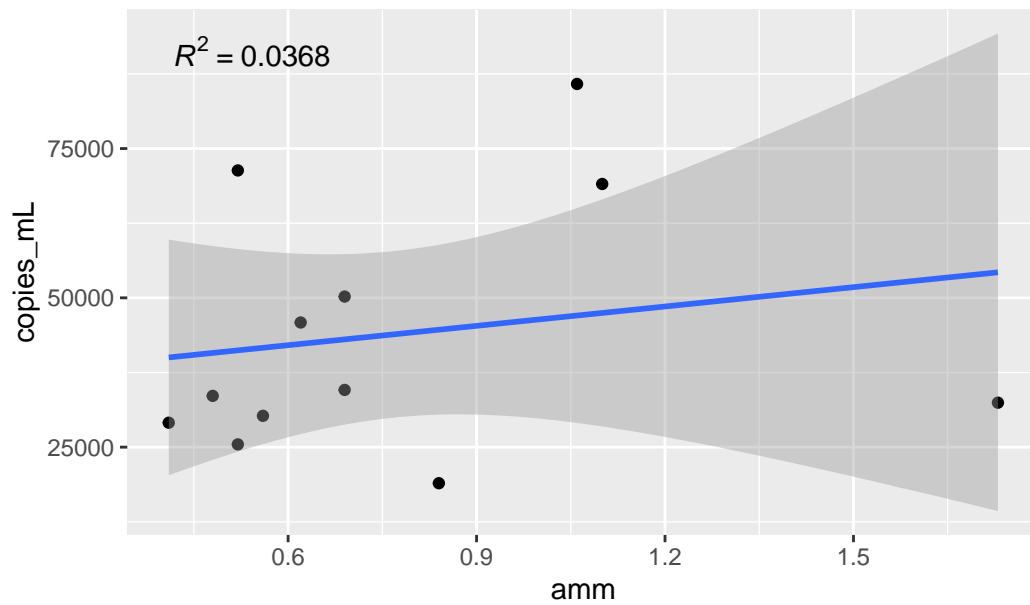


### Ammonia

```
b %>% ggplot(aes(x = amm, y = copies_mL)) +
  geom_point() +
  stat_smooth(method = "lm")+
  stat_poly_eq(aes(x = amm, y = copies_mL),
    rr.digits = 4) +
  labs(title = "LM ammxCopies (B0)")
```

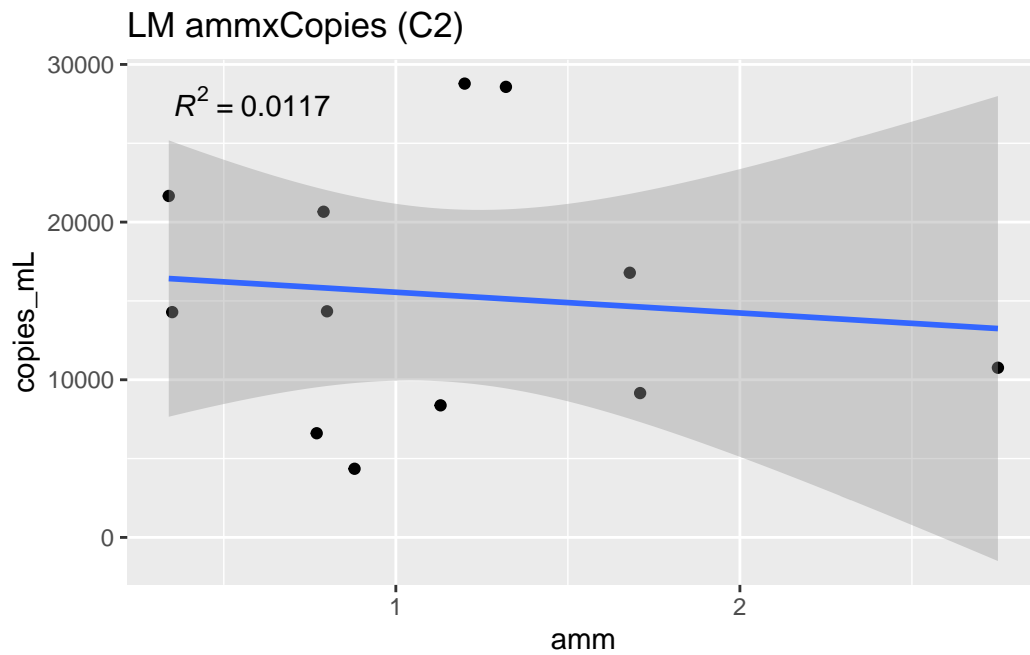
`geom\_smooth()` using formula = 'y ~ x'

LM ammxCopies (BO)



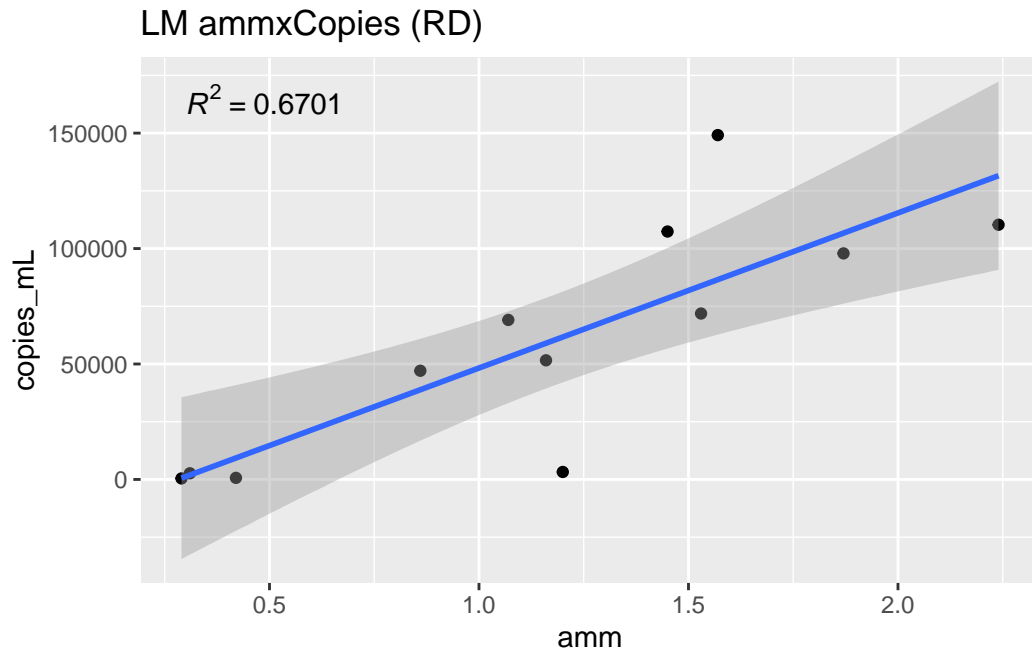
```
c %>% ggplot(aes(x = amm, y = copies_mL)) +
  geom_point() +
  stat_smooth(method = "lm")+
  stat_poly_eq(aes(x = amm, y = copies_mL),
               rr.digits = 4) +
  labs(title = "LM ammxCopies (C2)")
```

`geom\_smooth()` using formula = 'y ~ x'



```
r %>% ggplot(aes(x = amm, y = copies_mL)) +
  geom_point() +
  stat_smooth(method = "lm")+
  stat_poly_eq(aes(x = amm, y = copies_mL),
               rr.digits = 4) +
  labs(title = "LM ammxCopies (RD)")
```

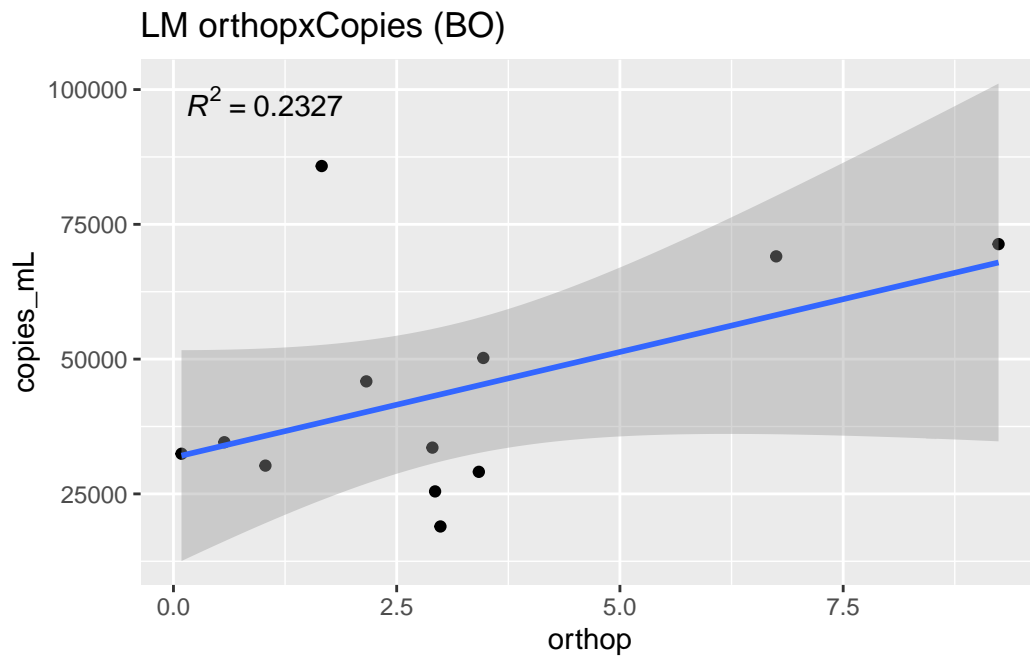
`geom\_smooth()` using formula = 'y ~ x'



### Orthophosphate

```
b %>% ggplot(aes(x = orthop, y = copies_mL)) +
  geom_point() +
  stat_smooth(method = "lm")+
  stat_poly_eq(aes(x = orthop, y = copies_mL),
    rr.digits = 4) +
  labs(title = "LM orthopxCopies (B0)")
```

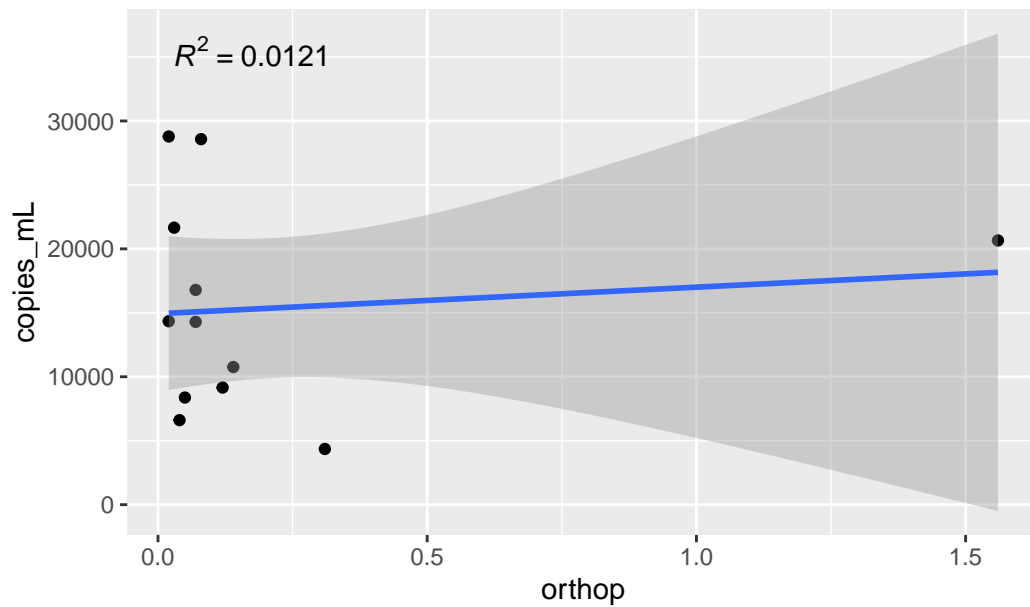
`geom\_smooth()` using formula = 'y ~ x'



```
c %>% ggplot(aes(x = orthop, y = copies_mL)) +
  geom_point() +
  stat_smooth(method = "lm")+
  stat_poly_eq(aes(x = orthop, y = copies_mL),
               rr.digits = 4) +
  labs(title = "LM orthopxCopies (C2)")
```

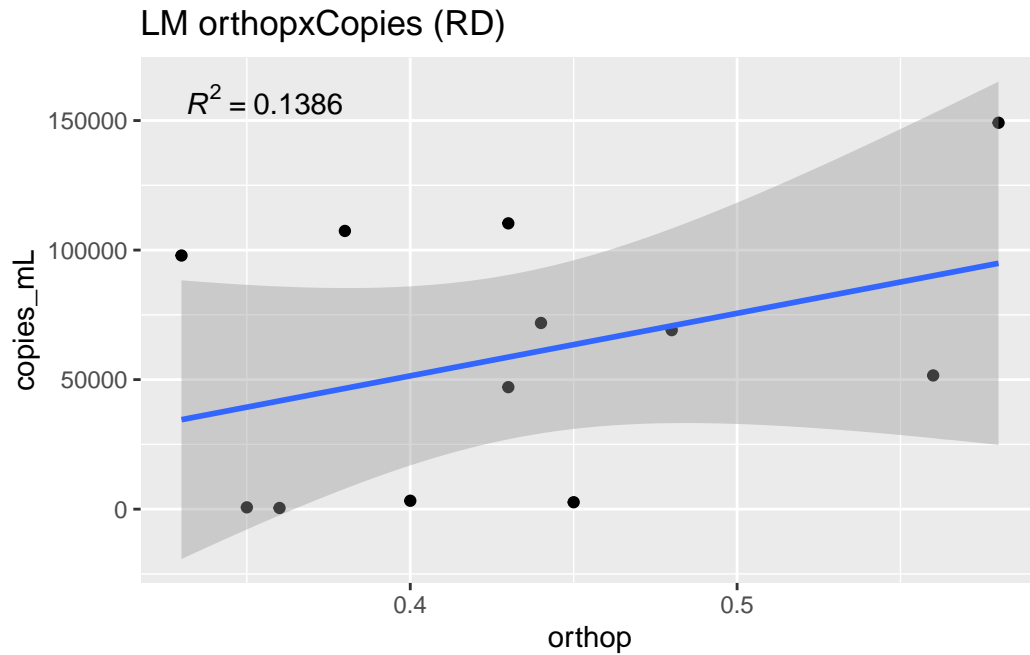
`geom\_smooth()` using formula = 'y ~ x'

LM orthopxCopies (C2)



```
r %>% ggplot(aes(x = orthop, y = copies_mL)) +
  geom_point() +
  stat_smooth(method = "lm")+
  stat_poly_eq(aes(x = orthop, y = copies_mL),
               rr.digits = 4) +
  labs(title = "LM orthopxCopies (RD)")
```

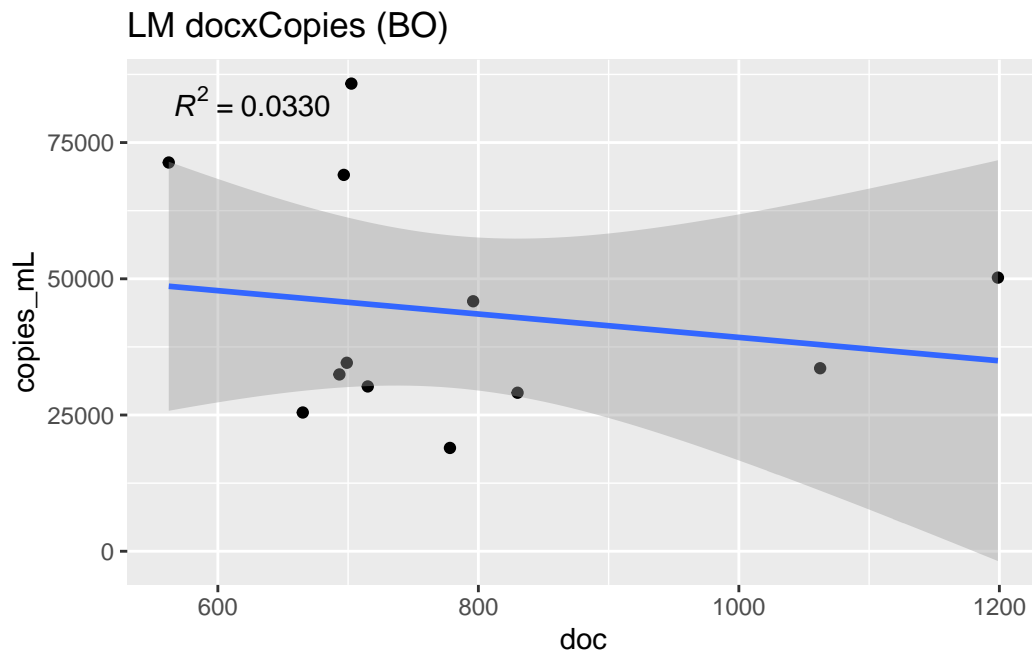
`geom\_smooth()` using formula = 'y ~ x'



## DOC

```
b %>% ggplot(aes(x = doc, y = copies_mL)) +
  geom_point() +
  stat_smooth(method = "lm")+
  stat_poly_eq(aes(x = doc, y = copies_mL),
    rr.digits = 4) +
  labs(title = "LM docxCopies (B0)")
```

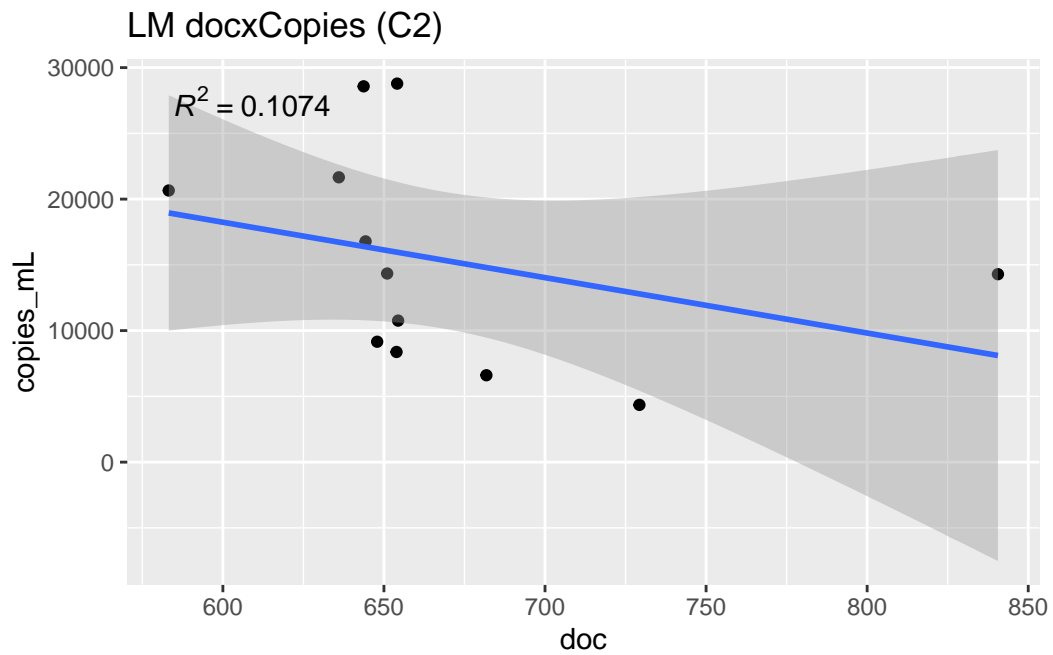
`geom\_smooth()` using formula = 'y ~ x'



```
c %>% ggplot(aes(x = doc, y = copies_mL)) +  
  geom_point() +  
  stat_smooth(method = "lm")+  
  stat_poly_eq(aes(x = doc, y = copies_mL),  
               rr.digits = 4) +  
  labs(title = "LM docxCopies (C2)")
```

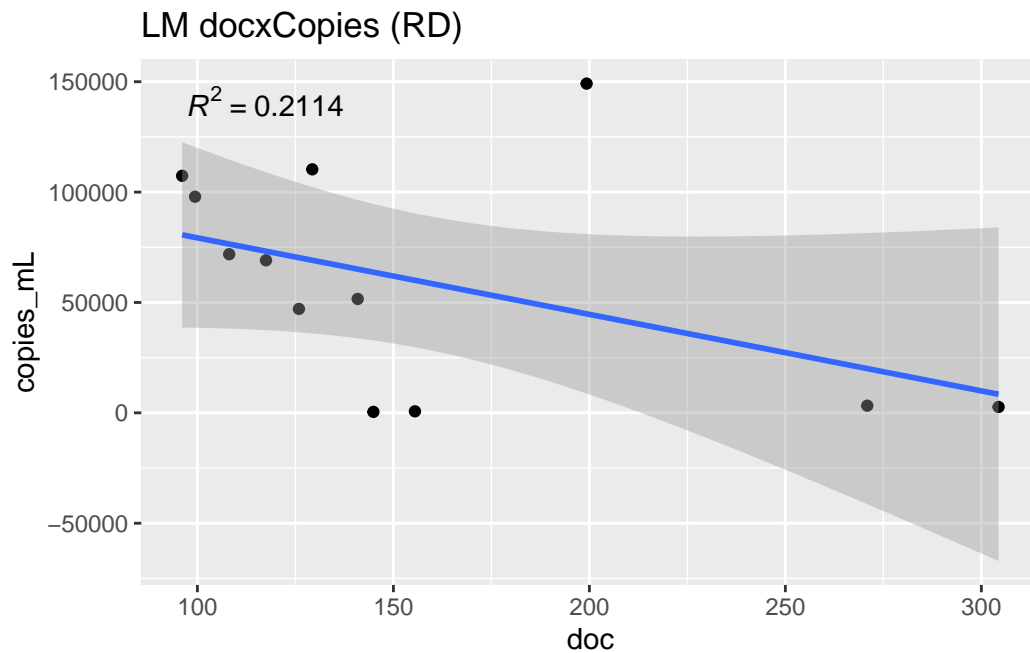
`geom\_smooth()` using formula = 'y ~ x'





```
r %>% ggplot(aes(x = doc, y = copies_mL)) +
  geom_point() +
  stat_smooth(method = "lm")+
  stat_poly_eq(aes(x = doc, y = copies_mL),
               rr.digits = 4) +
  labs(title = "LM docxCopies (RD)")
```

`geom\_smooth()` using formula = 'y ~ x'

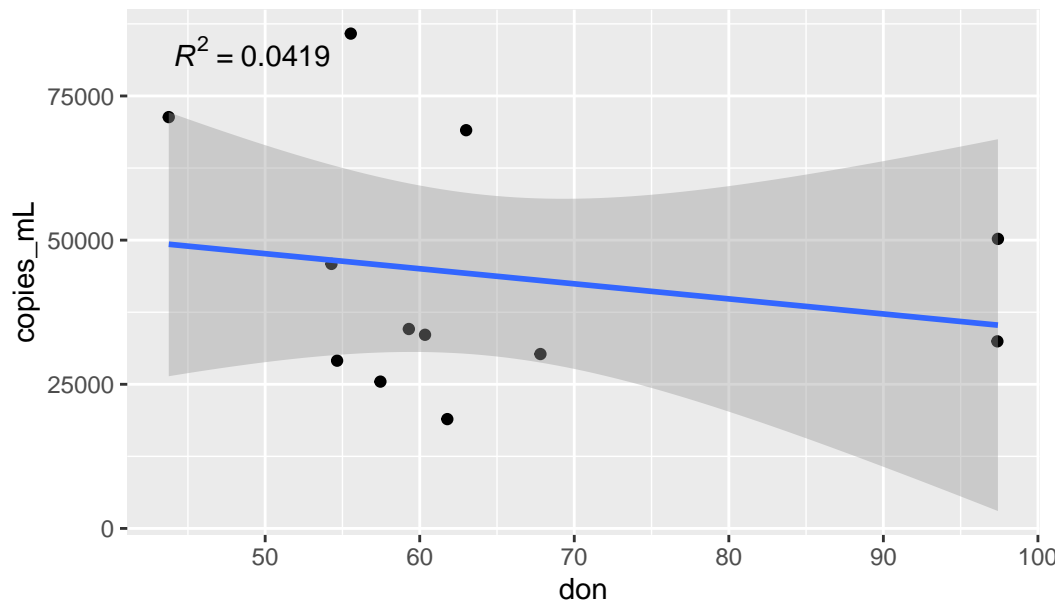


## DON

```
b %>% ggplot(aes(x = don, y = copies_mL)) +
  geom_point() +
  stat_smooth(method = "lm")+
  stat_poly_eq(aes(x = don, y = copies_mL),
    rr.digits = 4) +
  labs(title = "LM donxCopies (B0)")
```

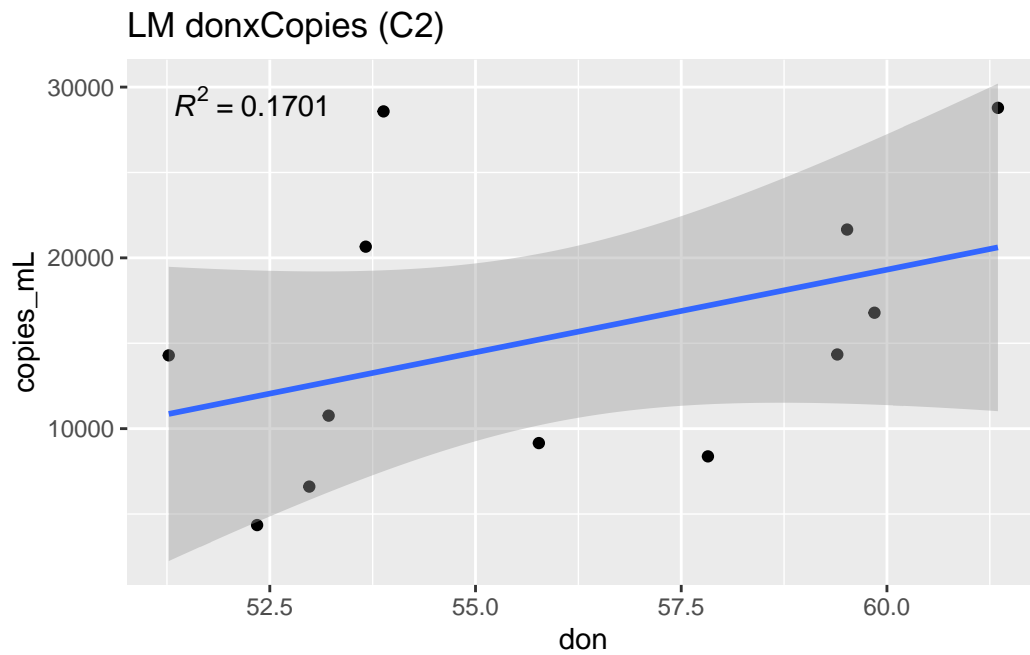
`geom\_smooth()` using formula = 'y ~ x'

LM donxCopies (BO)



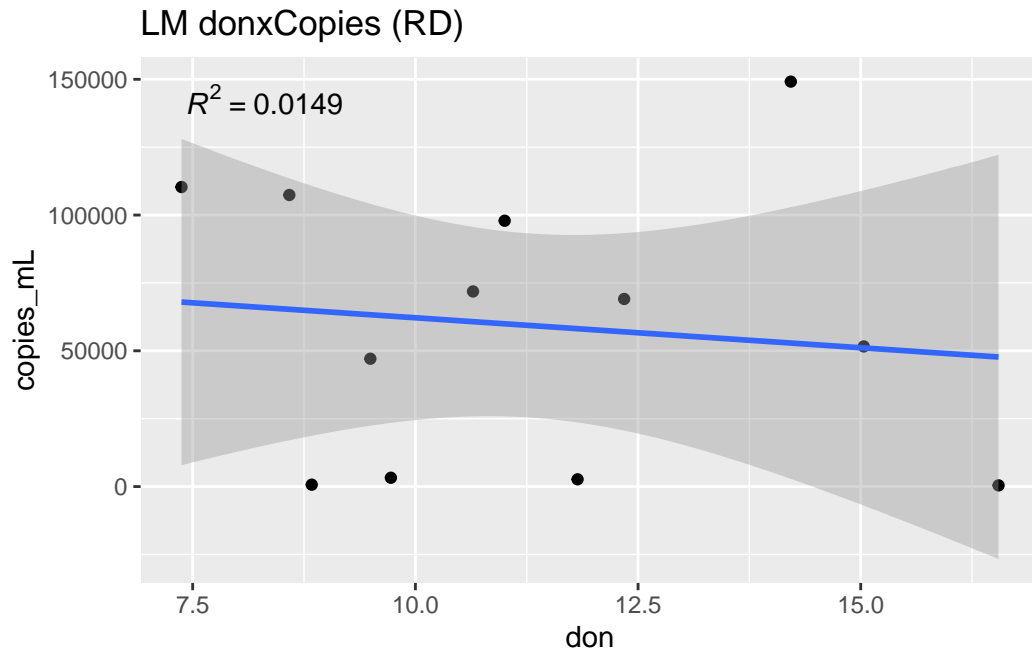
```
c %>% ggplot(aes(x = don, y = copies_mL)) +
  geom_point() +
  stat_smooth(method = "lm")+
  stat_poly_eq(aes(x = don, y = copies_mL),
               rr.digits = 4) +
  labs(title = "LM donxCopies (C2)")
```

`geom\_smooth()` using formula = 'y ~ x'



```
r %>% ggplot(aes(x = don, y = copies_mL)) +  
  geom_point() +  
  stat_smooth(method = "lm")+  
  stat_poly_eq(aes(x = don, y = copies_mL),  
               rr.digits = 4) +  
  labs(title = "LM donxCopies (RD)")
```

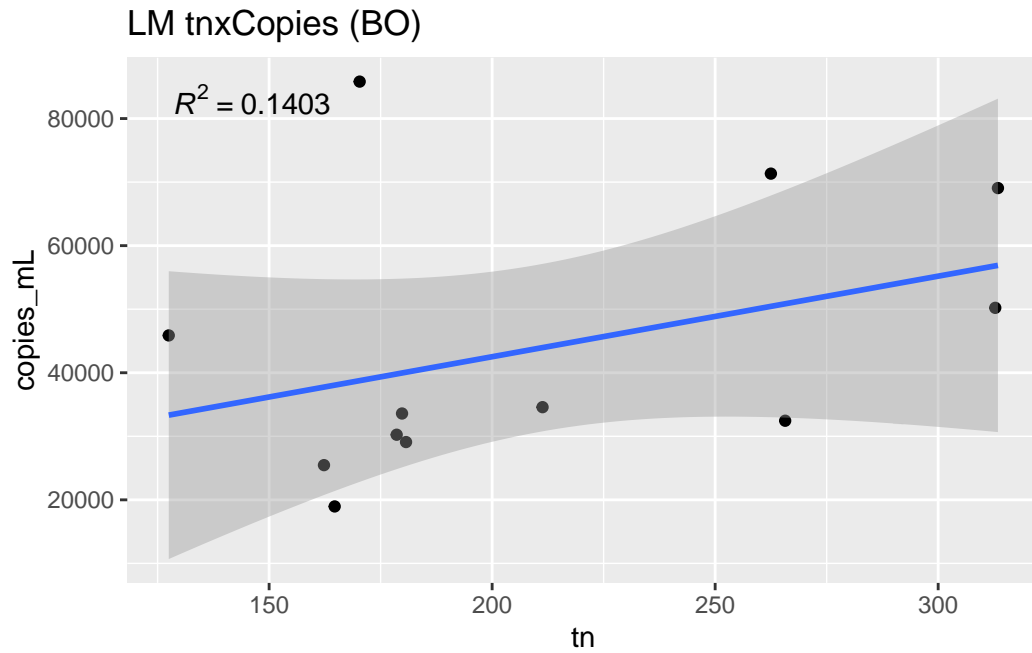
`geom\_smooth()` using formula = 'y ~ x'



### Total Nitrogen

```
b %>% ggplot(aes(x = tn, y = copies_mL)) +
  geom_point() +
  stat_smooth(method = "lm")+
  stat_poly_eq(aes(x = tn, y = copies_mL),
    rr.digits = 4) +
  labs(title = "LM tnxCopies (B0)")
```

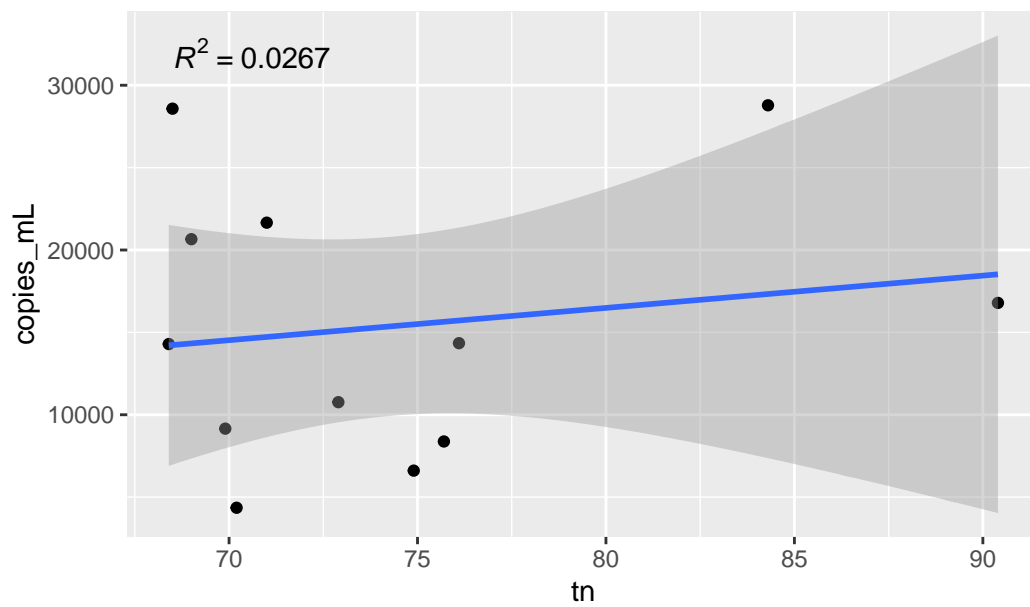
`geom\_smooth()` using formula = 'y ~ x'



```
c %>% ggplot(aes(x = tn, y = copies_mL)) +
  geom_point() +
  stat_smooth(method = "lm")+
  stat_poly_eq(aes(x = tn, y = copies_mL),
               rr.digits = 4) +
  labs(title = "LM tnxCopies (C2)")
```

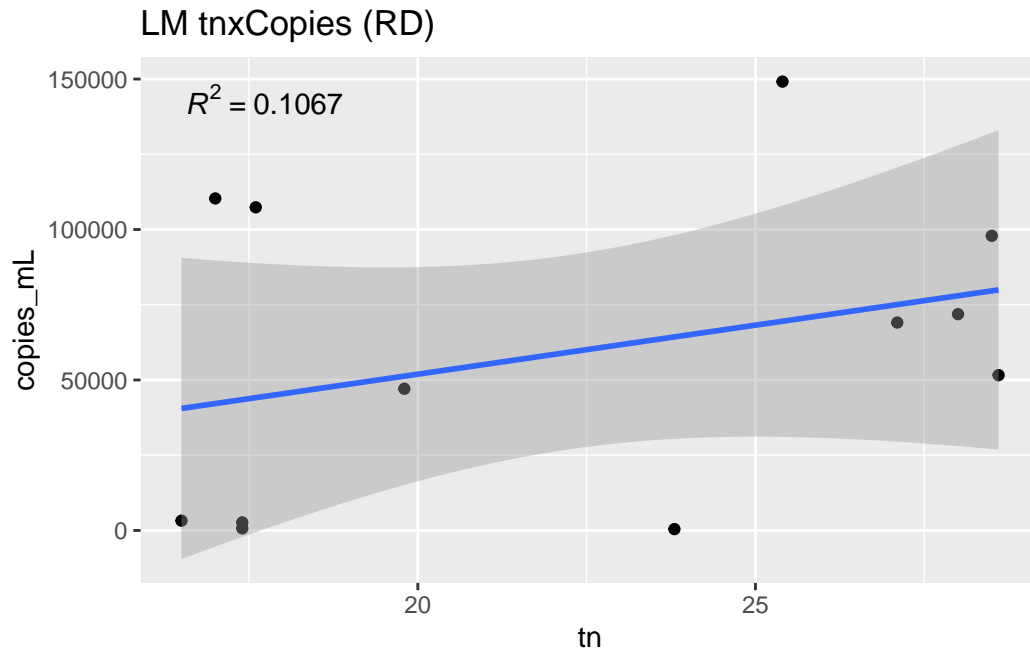
`geom\_smooth()` using formula = 'y ~ x'

LM tnxCopies (C2)



```
r %>% ggplot(aes(x = tn, y = copies_mL)) +
  geom_point() +
  stat_smooth(method = "lm")+
  stat_poly_eq(aes(x = tn, y = copies_mL),
               rr.digits = 4) +
  labs(title = "LM tnxCopies (RD)")
```

`geom\_smooth()` using formula = 'y ~ x'



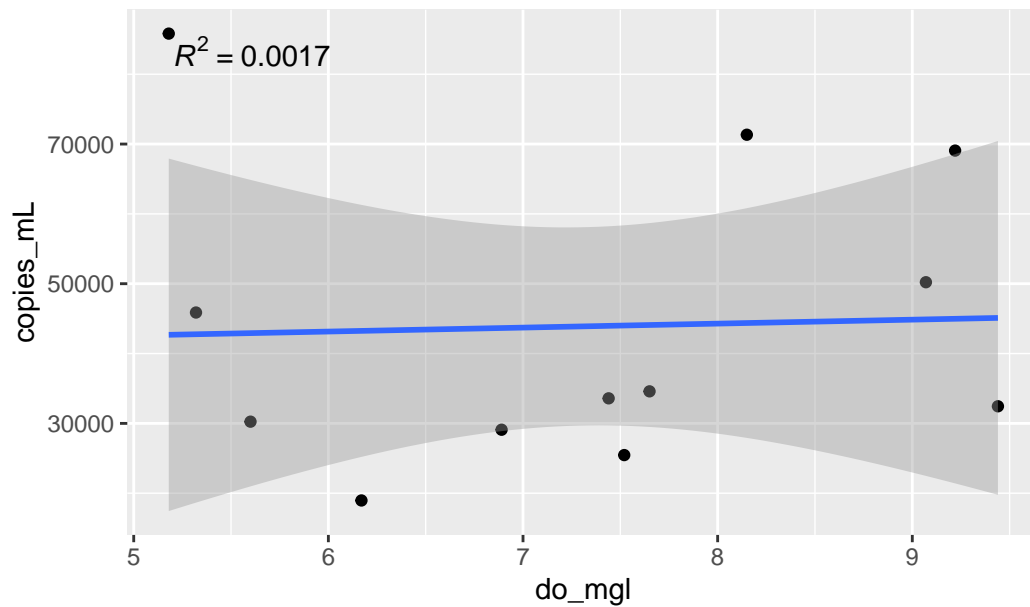
DO

```
b %>% ggplot(aes(x = do_mgl, y = copies_mL)) +
  geom_point() +
  stat_smooth(method = "lm")+
  stat_poly_eq(aes(x = do_mgl, y = copies_mL),
    rr.digits = 4) +
  labs(title = "LM do_mglxCopies (B0)")
```

`geom\_smooth()` using formula = 'y ~ x'

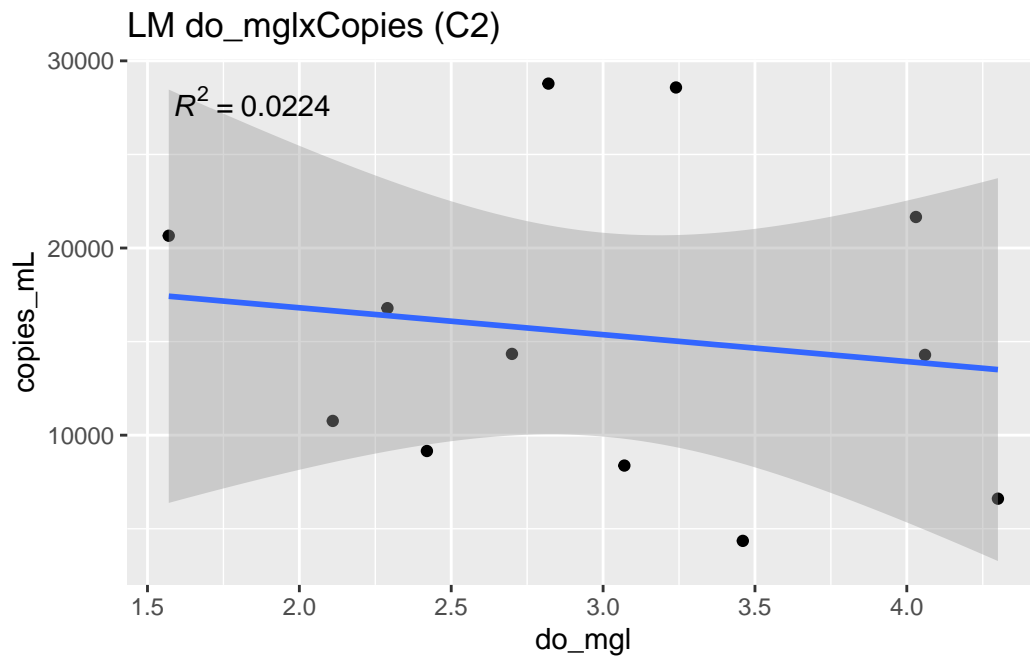


LM do\_mglxCopies (BO)



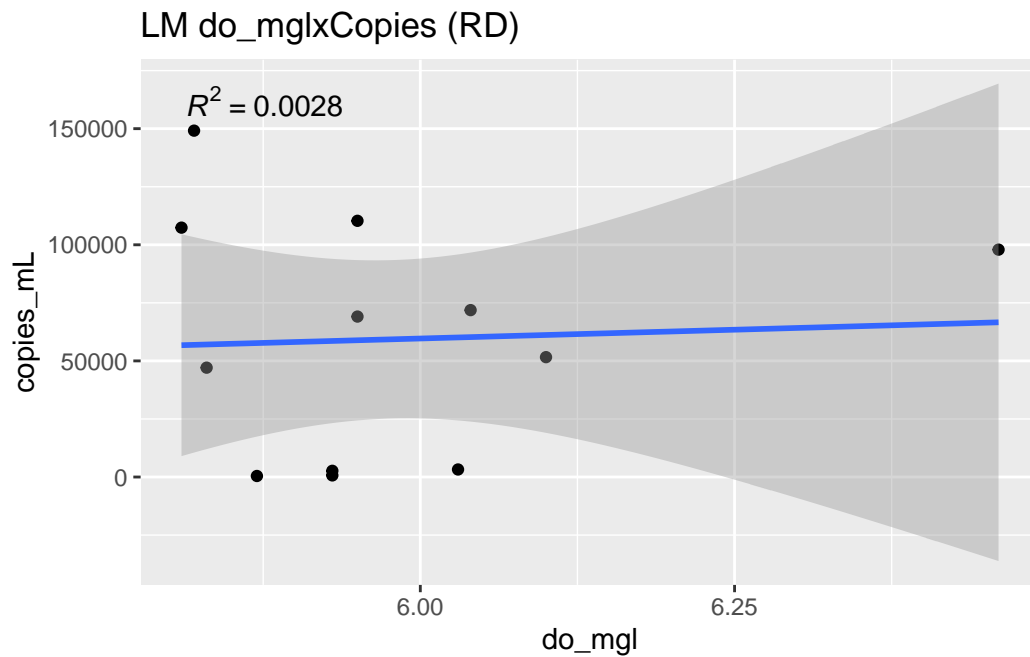
```
c %>% ggplot(aes(x = do_mgl, y = copies_mL)) +
  geom_point() +
  stat_smooth(method = "lm")+
  stat_poly_eq(aes(x = do_mgl, y = copies_mL),
               rr.digits = 4) +
  labs(title = "LM do_mglxCopies (C2)")
```

`geom\_smooth()` using formula = 'y ~ x'



```
r %>% ggplot(aes(x = do_mgl, y = copies_mL)) +  
  geom_point() +  
  stat_smooth(method = "lm")+  
  stat_poly_eq(aes(x = do_mgl, y = copies_mL),  
               rr.digits = 4) +  
  labs(title = "LM do_mglxCopies (RD)")
```

`geom\_smooth()` using formula = 'y ~ x'

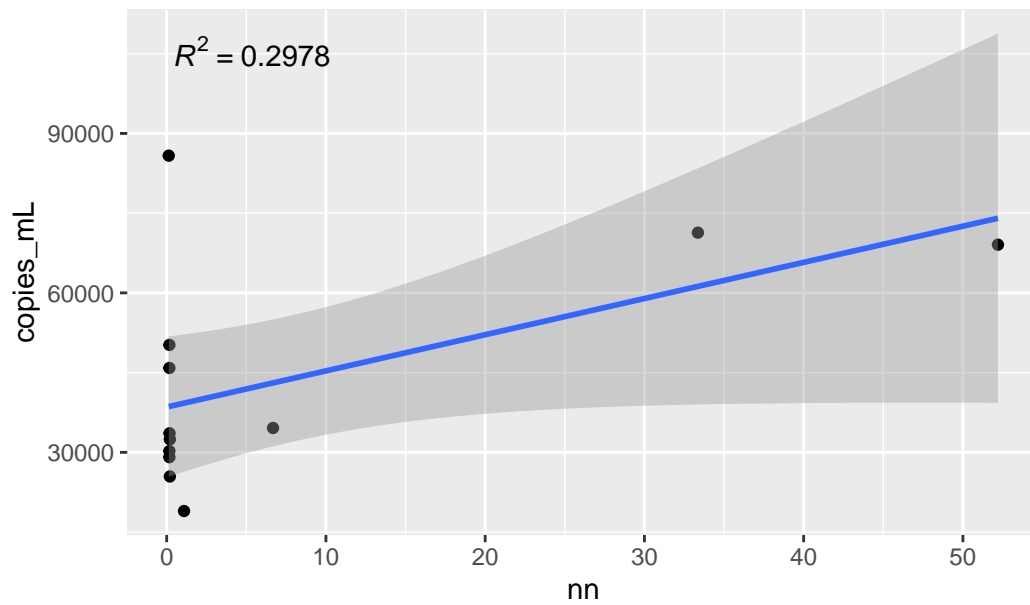


NN

```
b %>% ggplot(aes(x = nn, y = copies_mL)) +
  geom_point() +
  stat_smooth(method = "lm")+
  stat_poly_eq(aes(x = nn, y = copies_mL),
    rr.digits = 4) +
  labs(title = "LM nnxCopies (B0)")
```

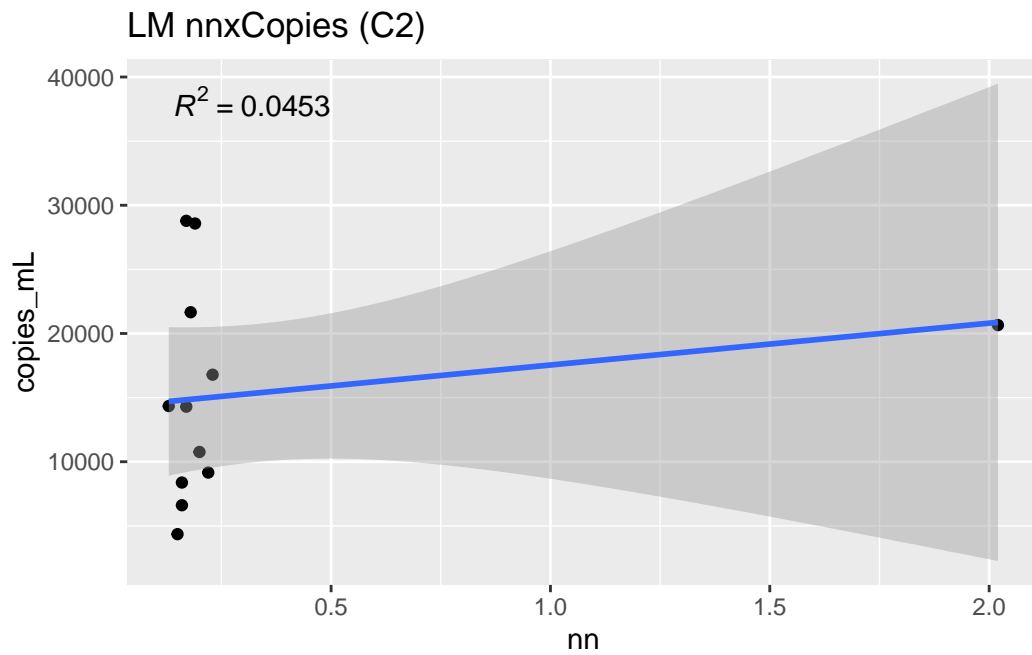
`geom\_smooth()` using formula = 'y ~ x'

LM nnxCopies (BO)



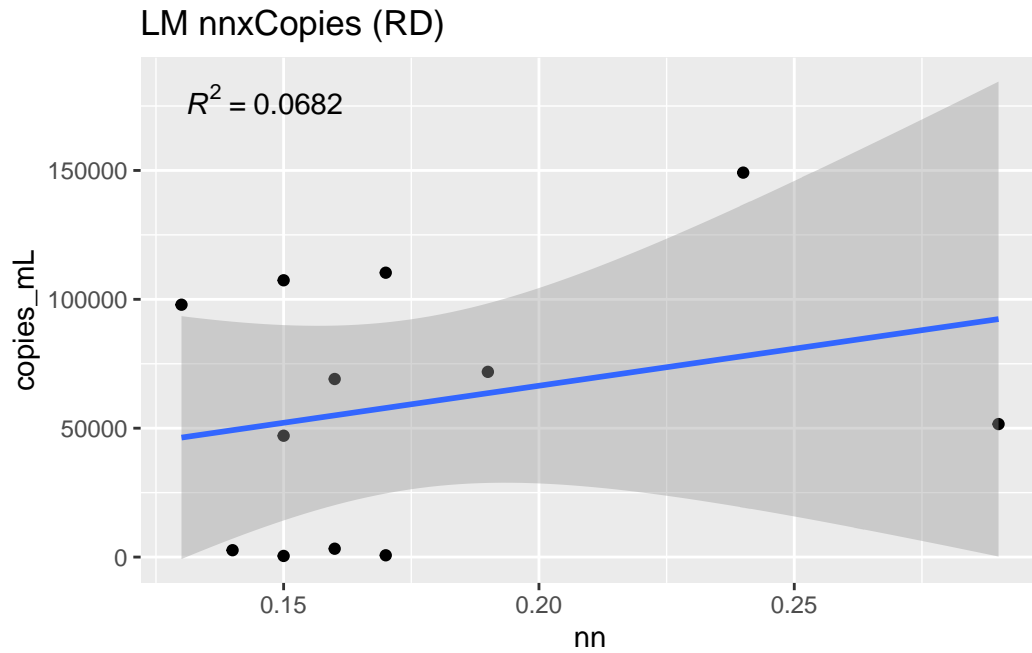
```
c %>% ggplot(aes(x = nn, y = copies_mL)) +
  geom_point() +
  stat_smooth(method = "lm")+
  stat_poly_eq(aes(x = nn, y = copies_mL),
               rr.digits = 4) +
  labs(title = "LM nnxCopies (C2)")
```

`geom\_smooth()` using formula = 'y ~ x'



```
r %>% ggplot(aes(x = nn, y = copies_mL)) +
  geom_point() +
  stat_smooth(method = "lm")+
  stat_poly_eq(aes(x = nn, y = copies_mL),
               rr.digits = 4) +
  labs(title = "LM nnxCopies (RD)")
```

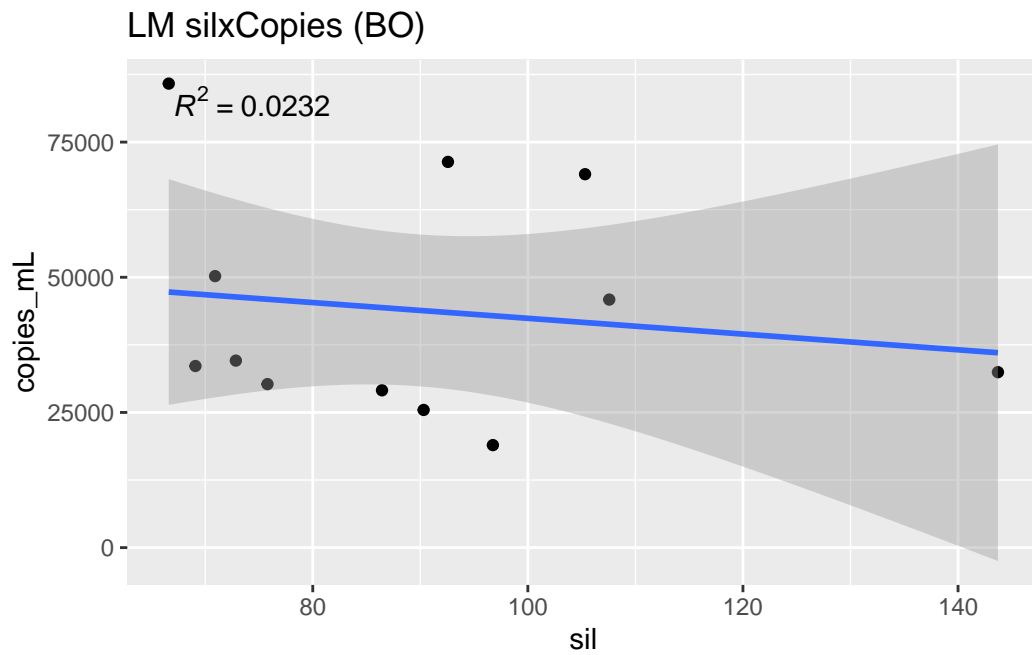
`geom\_smooth()` using formula = 'y ~ x'



### Silicate

```
b %>% ggplot(aes(x = sil, y = copies_mL)) +
  geom_point() +
  stat_smooth(method = "lm")+
  stat_poly_eq(aes(x = sil, y = copies_mL),
    rr.digits = 4) +
  labs(title = "LM silxCopies (B0)")
```

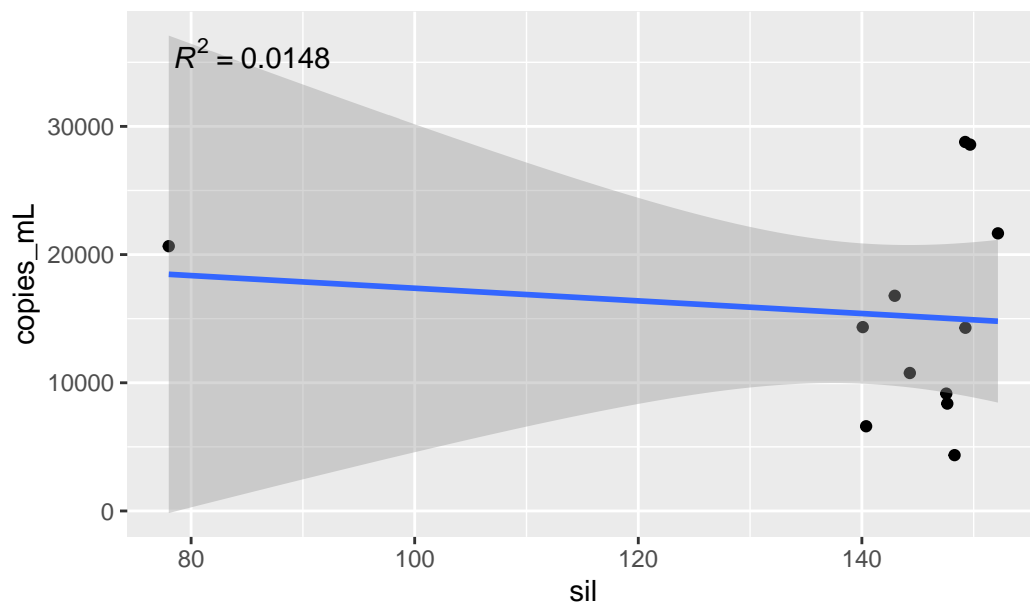
`geom\_smooth()` using formula = 'y ~ x'



```
c %>% ggplot(aes(x = sil, y = copies_mL)) +  
  geom_point() +  
  stat_smooth(method = "lm")+  
  stat_poly_eq(aes(x = sil, y = copies_mL),  
               rr.digits = 4) +  
  labs(title = "LM silxCopies (C2)")
```

`geom\_smooth()` using formula = 'y ~ x'

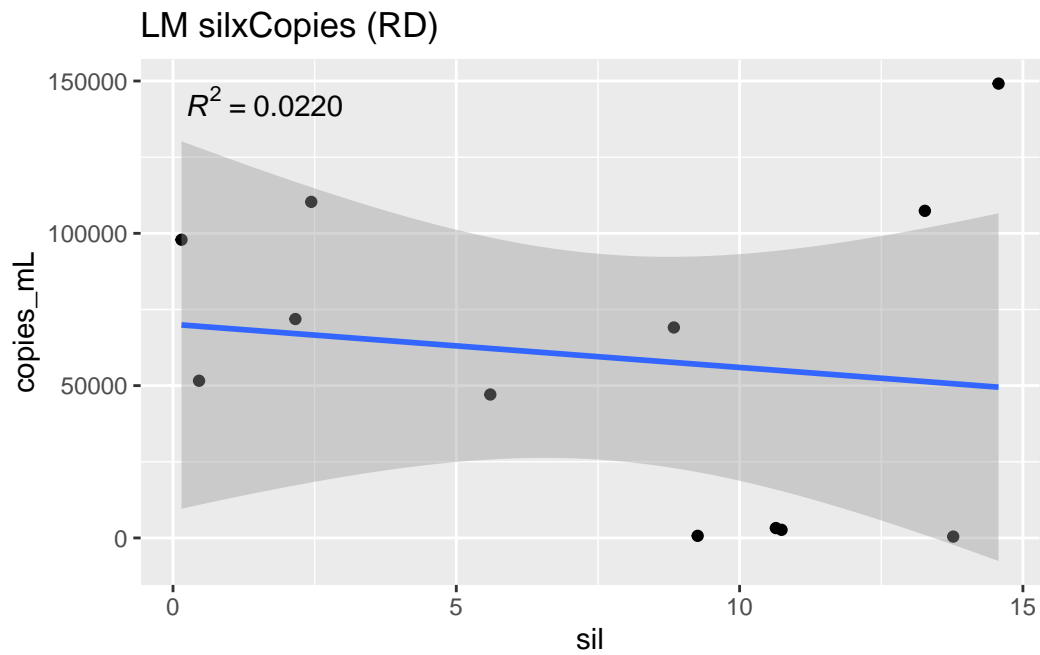
LM silxCopies (C2)



```
r %>% ggplot(aes(x = sil, y = copies_mL)) +
  geom_point() +
  stat_smooth(method = "lm")+
  stat_poly_eq(aes(x = sil, y = copies_mL),
               rr.digits = 4) +
  labs(title = "LM silxCopies (RD)")
```

`geom\_smooth()` using formula = 'y ~ x'

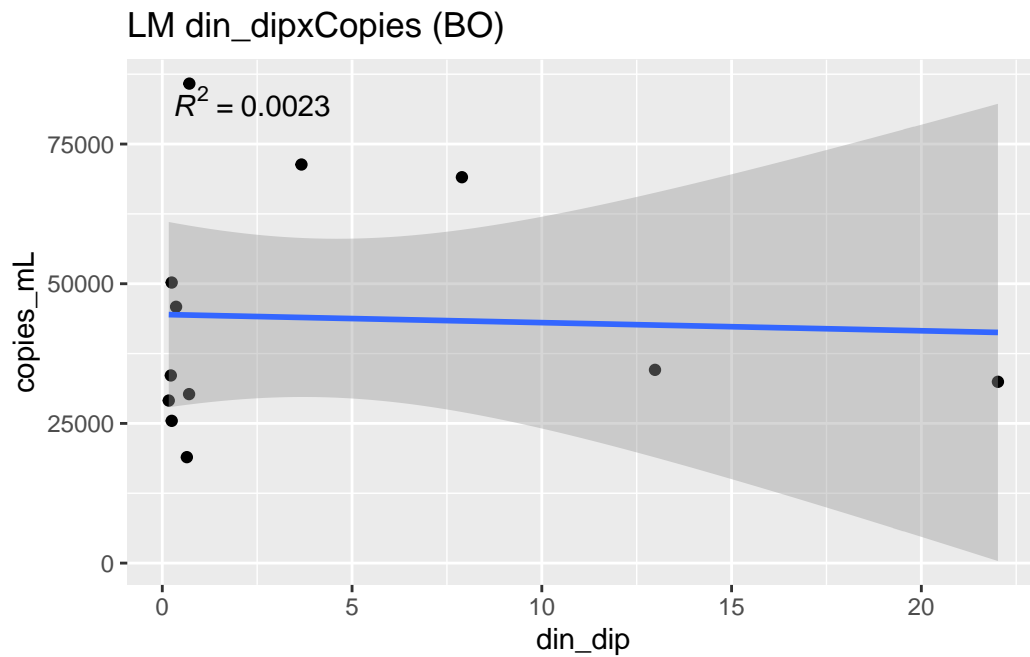




## DIN:DIP

```
b %>% ggplot(aes(x = din_dip, y = copies_mL)) +
  geom_point() +
  stat_smooth(method = "lm")+
  stat_poly_eq(aes(x = din_dip, y = copies_mL),
    rr.digits = 4) +
  labs(title = "LM din_dipxCopies (B0)")
```

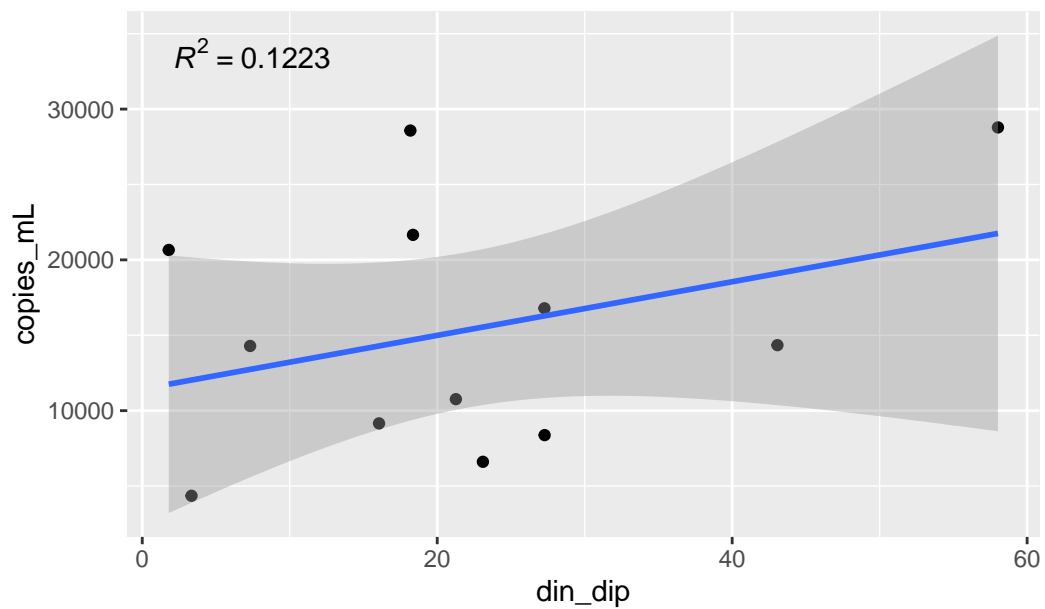
`geom\_smooth()` using formula = 'y ~ x'



```
c %>% ggplot(aes(x = din_dip, y = copies_mL)) +
  geom_point() +
  stat_smooth(method = "lm")+
  stat_poly_eq(aes(x = din_dip, y = copies_mL),
               rr.digits = 4) +
  labs(title = "LM din_dipxCopies (C2)")
```

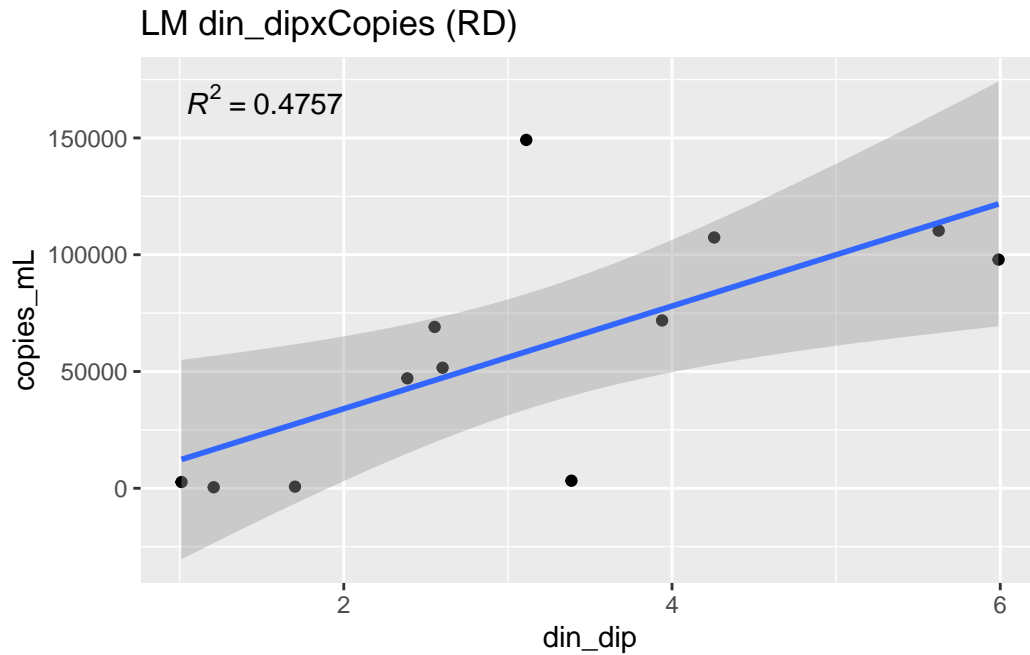
`geom\_smooth()` using formula = 'y ~ x'

LM din\_dipxCopies (C2)



```
r %>% ggplot(aes(x = din_dip, y = copies_mL)) +
  geom_point() +
  stat_smooth(method = "lm")+
  stat_poly_eq(aes(x = din_dip, y = copies_mL),
               rr.digits = 4) +
  labs(title = "LM din_dipxCopies (RD)")
```

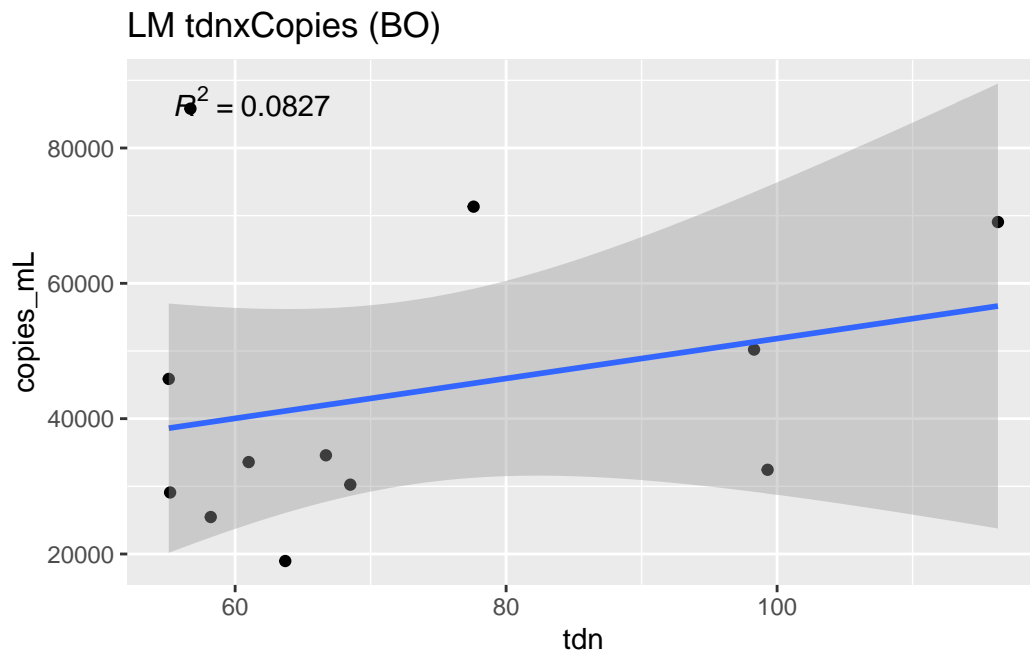
`geom\_smooth()` using formula = 'y ~ x'



## TDN

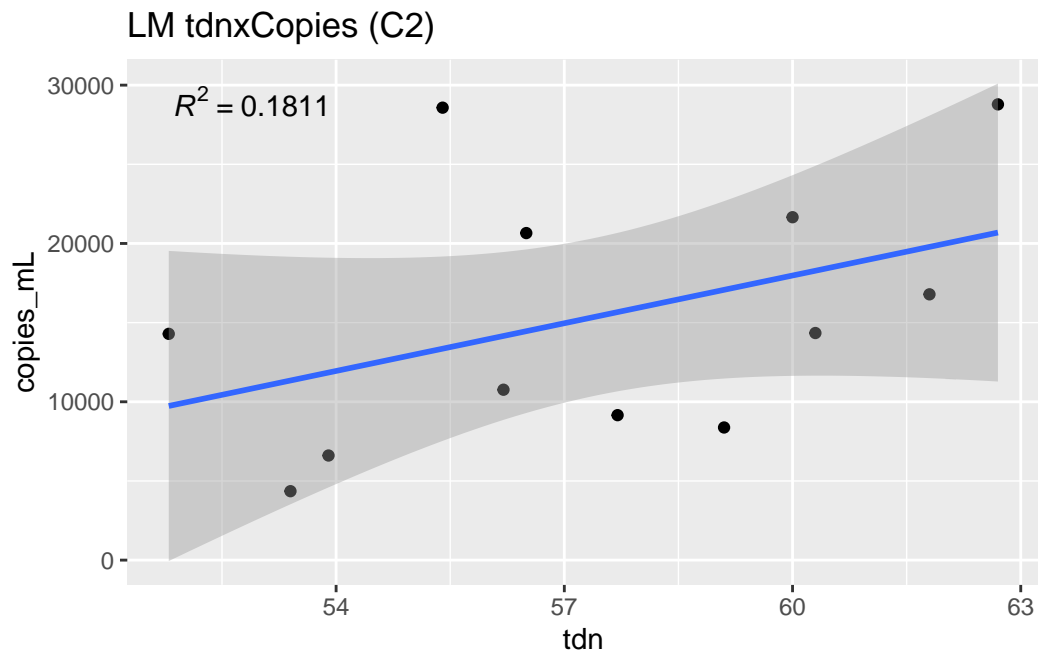
```
b %>% ggplot(aes(x = tdn, y = copies_mL)) +
  geom_point() +
  stat_smooth(method = "lm")+
  stat_poly_eq(aes(x = tdn, y = copies_mL),
    rr.digits = 4) +
  labs(title = "LM tdnxCopies (B0)")
```

`geom\_smooth()` using formula = 'y ~ x'



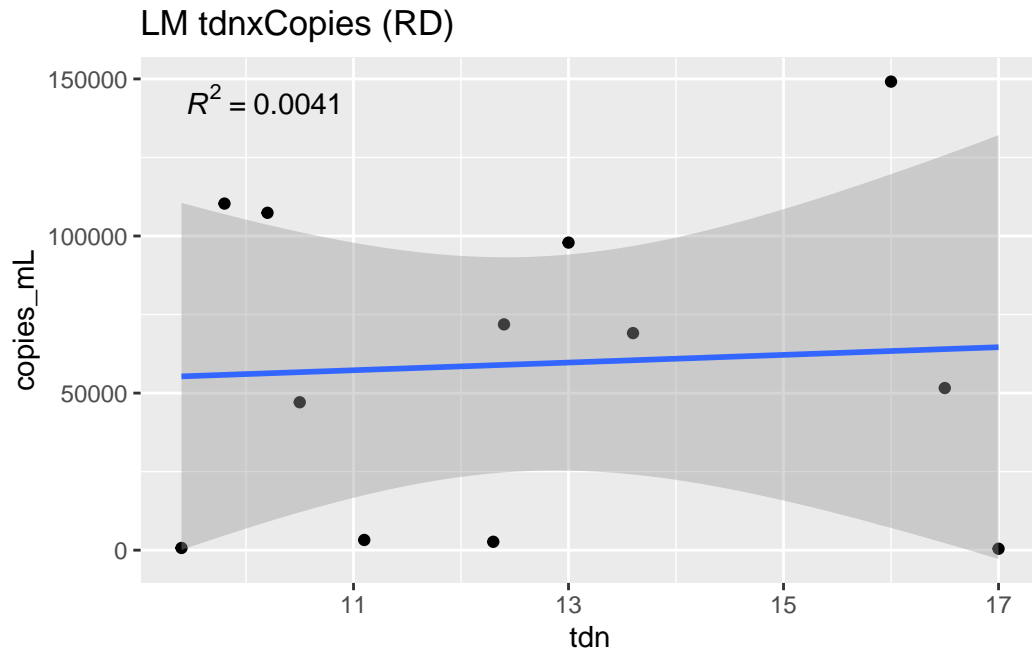
```
c %>% ggplot(aes(x = tdn, y = copies_mL)) +
  geom_point() +
  stat_smooth(method = "lm")+
  stat_poly_eq(aes(x = tdn, y = copies_mL),
               rr.digits = 4) +
  labs(title = "LM tdnxCopies (C2)")
```

`geom\_smooth()` using formula = 'y ~ x'



```
r %>% ggplot(aes(x = tdn, y = copies_mL)) +
  geom_point() +
  stat_smooth(method = "lm")+
  stat_poly_eq(aes(x = tdn, y = copies_mL),
               rr.digits = 4) +
  labs(title = "LM tdnxCopies (RD)")
```

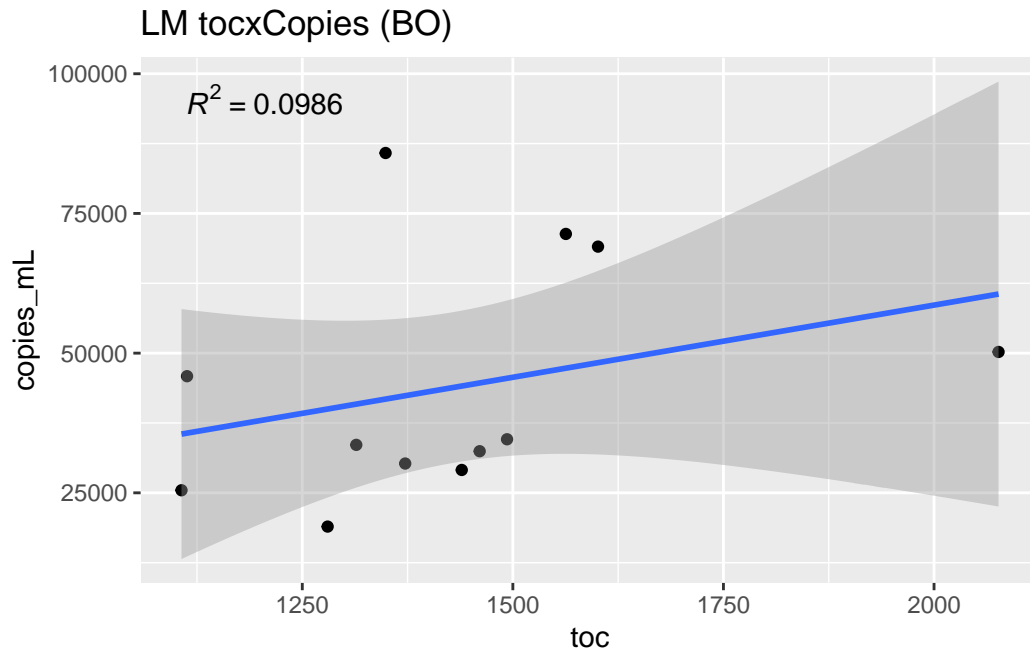
`geom\_smooth()` using formula = 'y ~ x'



## TOC

```
b %>% ggplot(aes(x = toc, y = copies_mL)) +
  geom_point() +
  stat_smooth(method = "lm")+
  stat_poly_eq(aes(x = toc, y = copies_mL),
    rr.digits = 4) +
  labs(title = "LM tocxCopies (B0)")
```

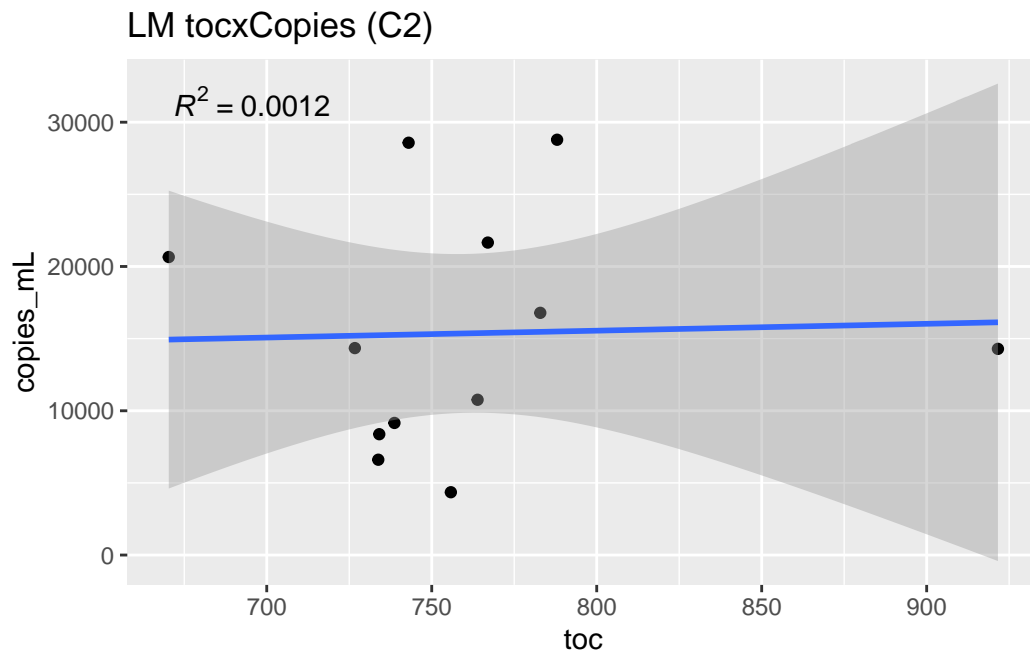
`geom\_smooth()` using formula = 'y ~ x'



```
c %>% ggplot(aes(x = toc, y = copies_mL)) +  
  geom_point() +  
  stat_smooth(method = "lm")+  
  stat_poly_eq(aes(x = toc, y = copies_mL),  
               rr.digits = 4) +  
  labs(title = "LM tocxCopies (C2)")
```

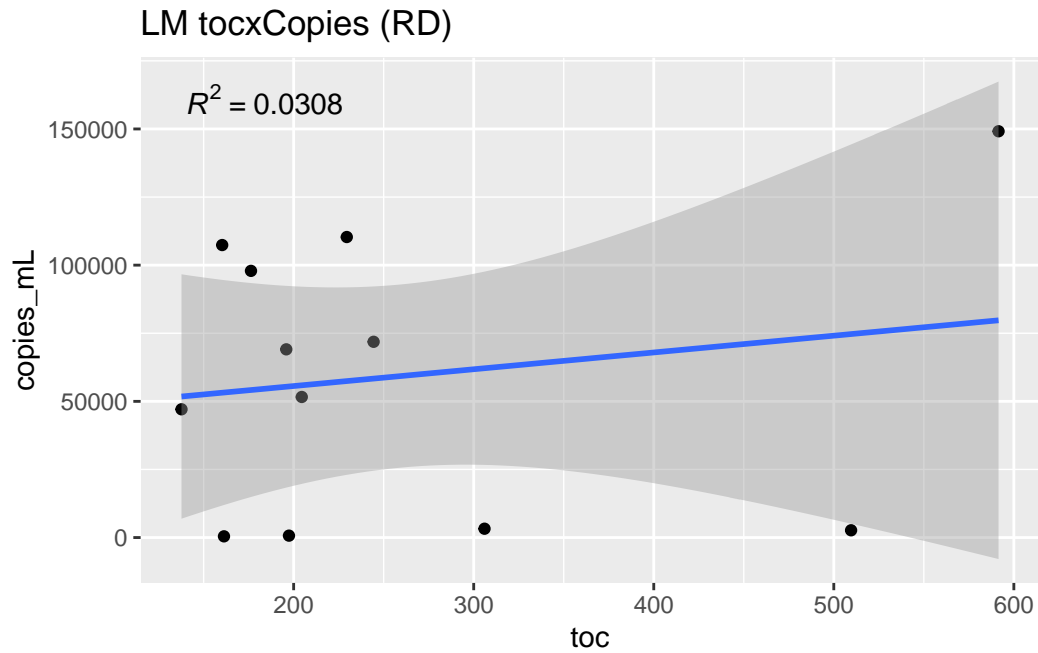
`geom\_smooth()` using formula = 'y ~ x'





```
r %>% ggplot(aes(x = toc, y = copies_mL)) +  
  geom_point() +  
  stat_smooth(method = "lm")+  
  stat_poly_eq(aes(x = toc, y = copies_mL),  
               rr.digits = 4) +  
  labs(title = "LM tocxCopies (RD)")
```

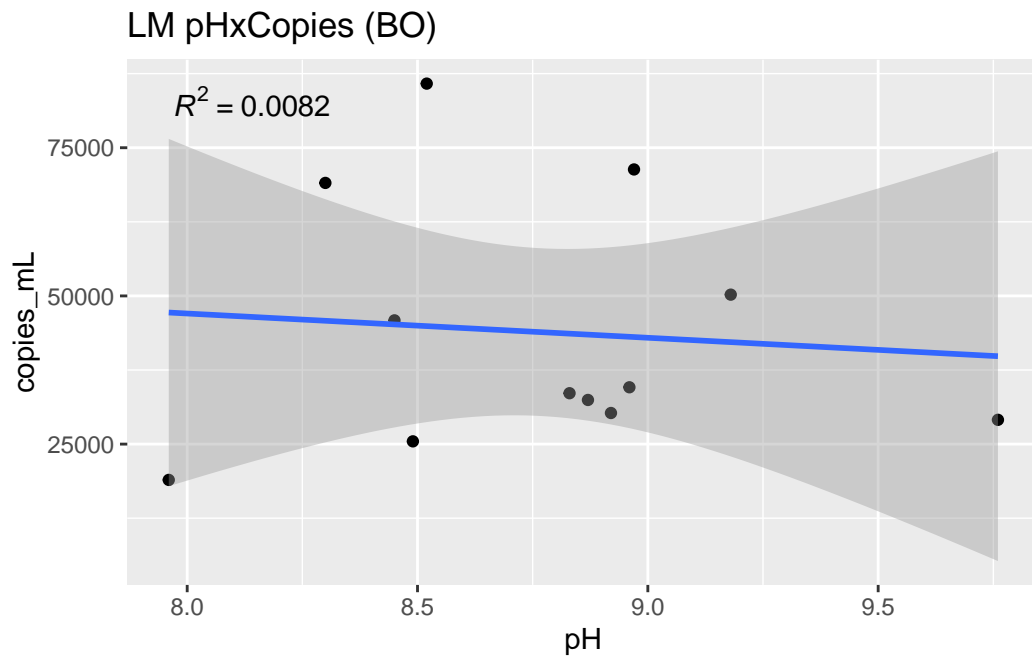
`geom\_smooth()` using formula = 'y ~ x'



pH

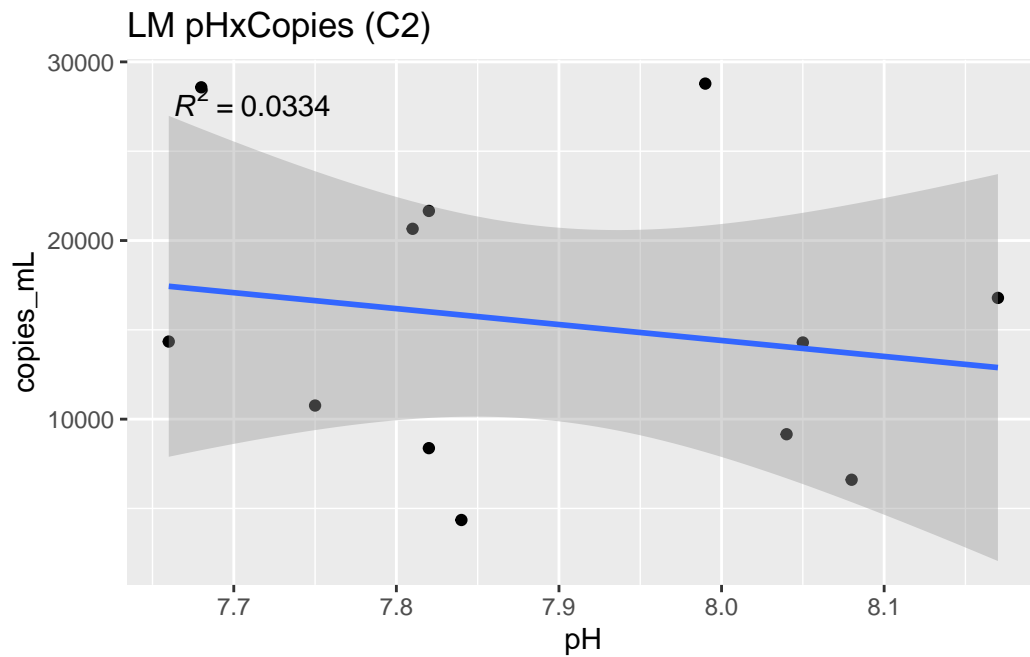
```
b %>% ggplot(aes(x = pH, y = copies_mL)) +
  geom_point() +
  stat_smooth(method = "lm")+
  stat_poly_eq(aes(x = pH, y = copies_mL),
    rr.digits = 4) +
  labs(title = "LM pHxCopies (B0)")
```

`geom\_smooth()` using formula = 'y ~ x'



```
c %>% ggplot(aes(x = pH, y = copies_mL)) +
  geom_point() +
  stat_smooth(method = "lm")+
  stat_poly_eq(aes(x = pH, y = copies_mL),
               rr.digits = 4) +
  labs(title = "LM pHxCopies (C2)")
```

`geom\_smooth()` using formula = 'y ~ x'



```
r %>% ggplot(aes(x = pH, y = copies_mL)) +
  geom_point() +
  stat_smooth(method = "lm")+
  stat_poly_eq(aes(x = pH, y = copies_mL),
               rr.digits = 4) +
  labs(title = "LM pHxCopies (RD)")
```

`geom\_smooth()` using formula = 'y ~ x'

