Wheat

Table of Contents

## Load the working packages

library(AgroR)  
library(tidyverse)  
library(agricolae)  
library(lme4)  
library(emmeans)  
library(reshape)  
library(reshape2)  
library(car)

## Load the data

setwd("~/Code/R/SCASI")  
library(tidyverse)  
library(skimr)  
  
df <- readxl::read\_excel("./input/SCASI\_Field\_Trials\_Data\_2022ALL\_20230409\_Adama\_v2.xlsx",sheet = "MergePart2Part1")  
# First rows  
#knitr::kable(head(df))  
  
# Convert to factor  
df$Rep <- as.factor(df$Rep)  
df$Treatment <- as.factor(df$Treatment)  
df$Woreda <- as.factor(df$Woreda)

## Overview of the data

Number of observations and variables:

## [1] 709 153

## Extract data per crops

# crop\_data  
crop\_data <- df %>% filter(Plot\_crop==params$crop)

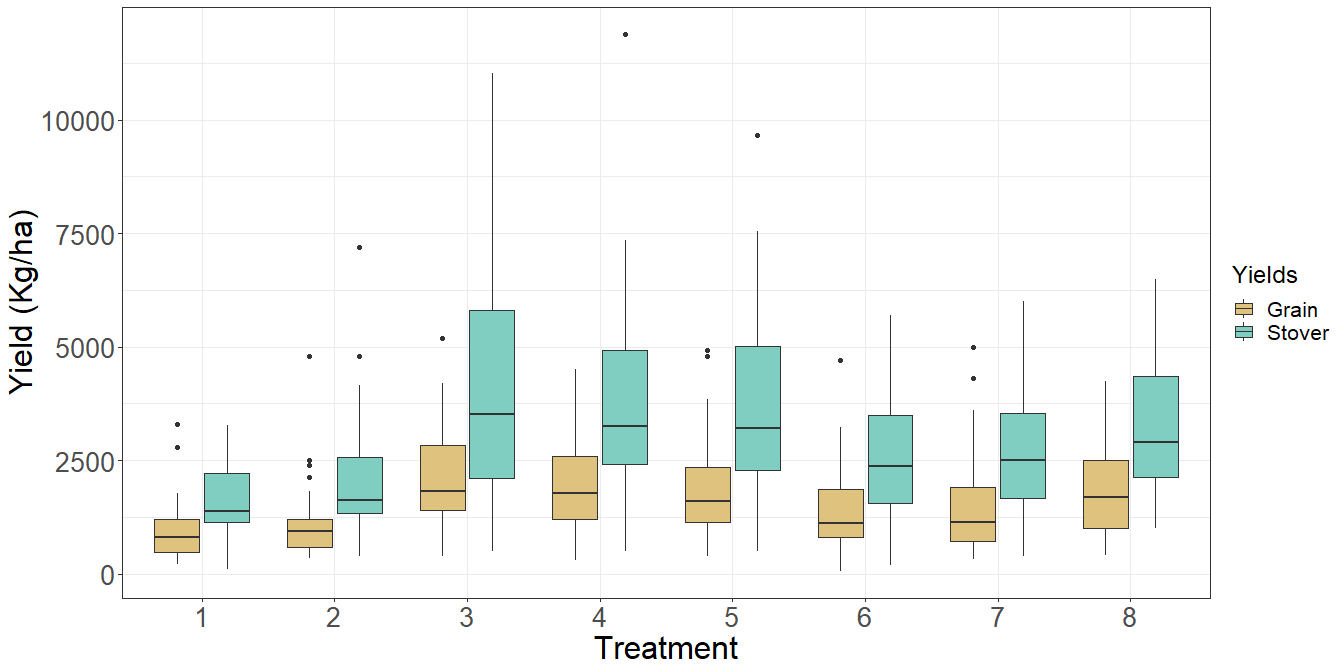
## Descriptive statistics

### Yield accorss treatments

yield\_trt\_summary<- crop\_data %>% group\_by(Treatment) %>% summarize(across(.cols = c(GY\_kg\_ha,SY\_kg\_ha),list(min=min,max=max,mean=mean,sd=sd),na.rm=T,.names = "{.fn}\_{.col}"))  
knitr::kable(yield\_trt\_summary)

|  |  |  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- | --- | --- |
| Treatment | min\_GY\_kg\_ha | max\_GY\_kg\_ha | mean\_GY\_kg\_ha | sd\_GY\_kg\_ha | min\_SY\_kg\_ha | max\_SY\_kg\_ha | mean\_SY\_kg\_ha | sd\_SY\_kg\_ha |
| 1 | 216.00000 | 3300 | 931.3243 | 642.2074 | 100.0000 | 3280.000 | 1649.441 | 830.5371 |
| 2 | 352.00000 | 4800 | 1108.5045 | 831.2530 | 400.0000 | 7200.000 | 2105.586 | 1321.9916 |
| 3 | 390.00000 | 5200 | 2159.2281 | 1113.8683 | 500.0000 | 11033.333 | 4151.263 | 2292.4439 |
| 4 | 300.00000 | 4500 | 2019.1930 | 1037.8021 | 500.0000 | 11893.333 | 3839.605 | 2108.1472 |
| 5 | 380.00000 | 4920 | 1933.2821 | 1080.5043 | 500.0000 | 9666.667 | 3623.889 | 1952.5966 |
| 6 | 58.41584 | 4700 | 1396.3530 | 893.8123 | 186.1386 | 5700.000 | 2571.372 | 1398.0494 |
| 7 | 320.00000 | 5000 | 1495.3761 | 1067.4977 | 400.0000 | 6000.000 | 2658.667 | 1523.8745 |
| 8 | 420.00000 | 4235 | 1842.2523 | 982.3119 | 1000.0000 | 6500.000 | 3338.081 | 1596.6123 |

#yield\_trt\_summary  
#with(crop\_data, desc(Treatment, GrainYield, ylab = "Grain Yield ", xlab = "Treatment", ylim = NA))  
  
#boxplot  
crop\_data %>% select(GY\_kg\_ha, SY\_kg\_ha, Treatment) %>% gather(key = "Yields",value= "Value", -Treatment) %>%   
ggplot(aes(x=Treatment,y=Value, fill=Yields)) +  
 geom\_boxplot() +  
 scale\_fill\_manual(values = c("GY\_kg\_ha" = "#dfc27d", "SY\_kg\_ha" = "#80cdc1"), labels=c("Grain", "Stover")) +  
 labs(y="Yield (Kg/ha)")+  
 theme\_bw()+  
 theme(legend.title = element\_text(size = 18),  
 legend.text = element\_text(size = 16),  
 axis.title = element\_text(size = 24),  
 axis.text = element\_text(size = 20))

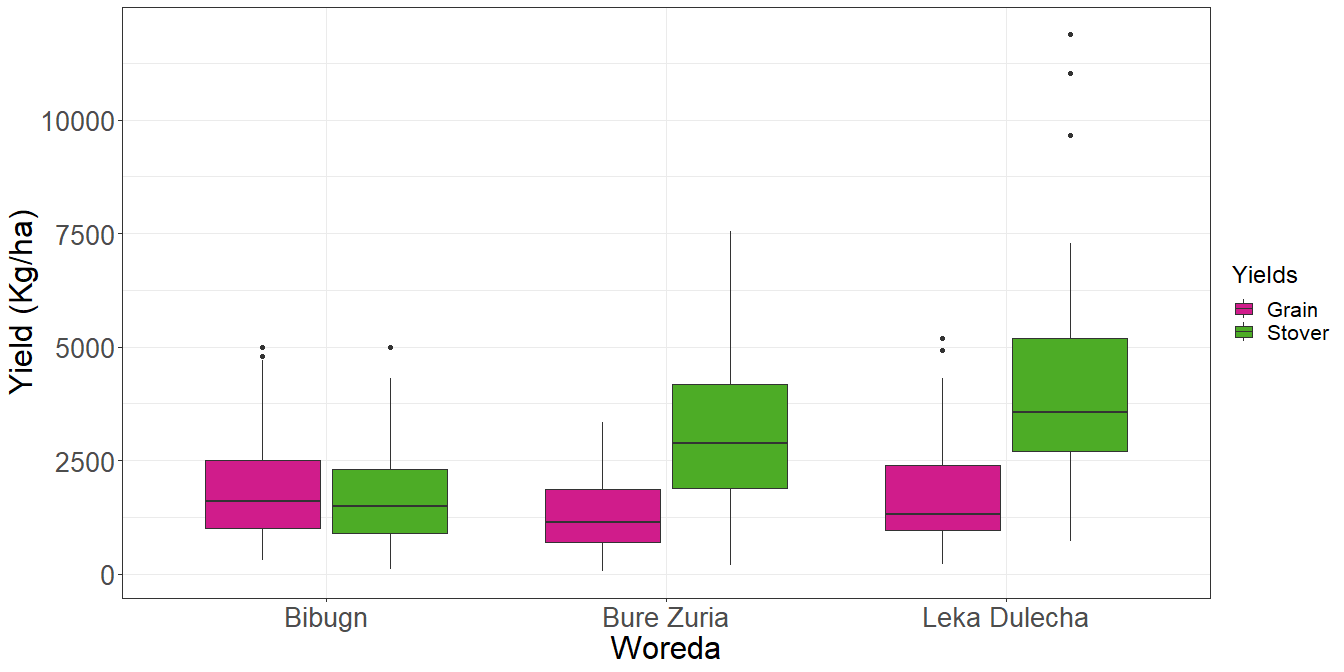


### Yield accorss locations

yield\_location<- crop\_data %>% group\_by(Woreda) %>% summarize(across(.cols = c(GY\_kg\_ha,SY\_kg\_ha),list(min=min,max=max,mean=mean,sd=sd),na.rm=T,.names = "{.fn}\_{.col}"))  
knitr::kable(yield\_location)

|  |  |  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- | --- | --- |
| Woreda | min\_GY\_kg\_ha | max\_GY\_kg\_ha | mean\_GY\_kg\_ha | sd\_GY\_kg\_ha | min\_SY\_kg\_ha | max\_SY\_kg\_ha | mean\_SY\_kg\_ha | sd\_SY\_kg\_ha |
| Bibugn | 300.00000 | 5000 | 1848.315 | 1113.8704 | 100.0000 | 5000.00 | 1680.899 | 997.8245 |
| Bure Zuria | 58.41584 | 3350 | 1326.316 | 786.2349 | 186.1386 | 7550.00 | 3219.814 | 1667.5542 |
| Leka Dulecha | 216.00000 | 5200 | 1801.484 | 1206.5370 | 712.0000 | 11893.33 | 4035.973 | 2043.3615 |

#yield\_trt\_summary  
#with(crop\_data, desc(Treatment, GrainYield, ylab = "Grain Yield ", xlab = "Treatment", ylim = NA))  
  
#boxplot  
crop\_data %>% select(GY\_kg\_ha, SY\_kg\_ha, Woreda) %>% gather(key = "Yields",value= "Value", -Woreda) %>%   
ggplot(aes(x=Woreda,y=Value, fill=Yields)) +  
 geom\_boxplot() +  
 scale\_fill\_manual(values = c("GY\_kg\_ha" = "#d01c8b", "SY\_kg\_ha" = "#4dac26"), labels=c("Grain", "Stover")) +  
 labs(y="Yield (Kg/ha)")+  
 theme\_bw()+  
 theme(legend.title = element\_text(size = 18),  
 legend.text = element\_text(size = 16),  
 axis.title = element\_text(size = 24),  
 axis.text = element\_text(size = 20))



### Yield accorss locations and treatments

* Summary table

yield\_location\_trt<- crop\_data %>% select(GY\_kg\_ha, SY\_kg\_ha, Woreda,Treatment) %>% group\_by(Woreda,Treatment) %>% summarize(across(.cols = c(GY\_kg\_ha, SY\_kg\_ha),list(min=min,max=max,mean=mean,sd=sd),na.rm=T,.names = "{.fn}\_{.col}"))

## `summarise()` has grouped output by 'Woreda'. You can override using the `.groups` argument.

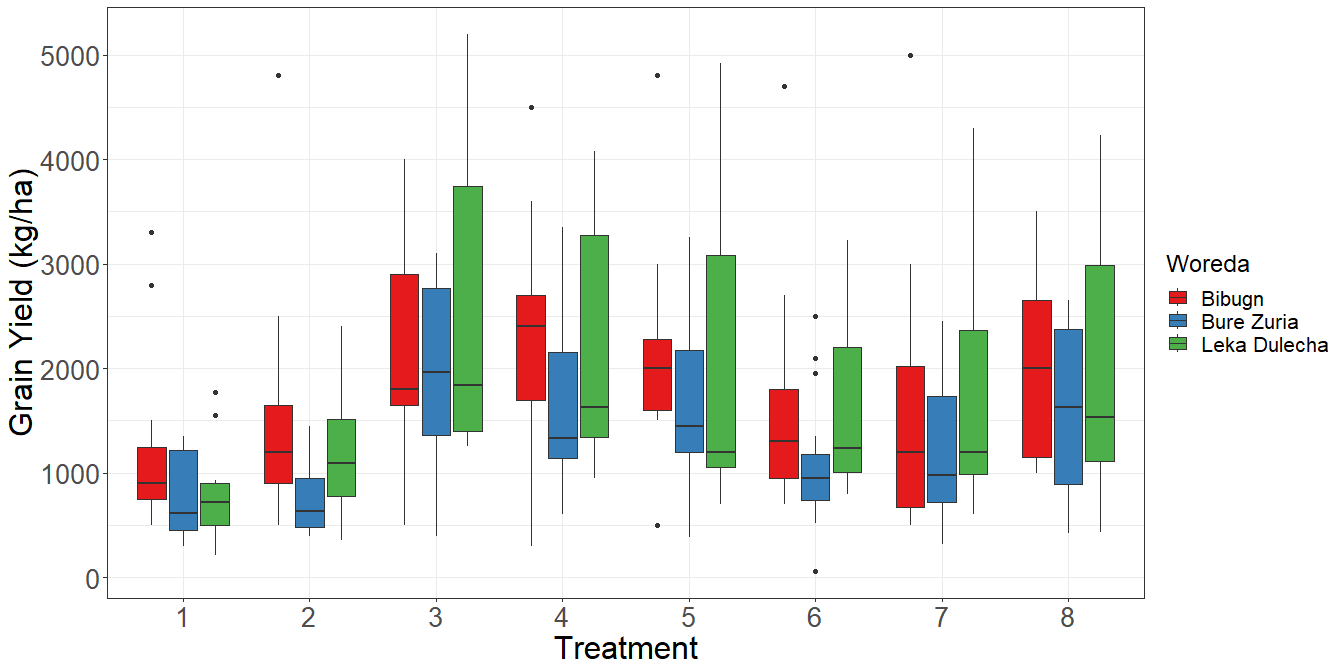
knitr::kable(yield\_location\_trt)

|  |  |  |  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- | --- | --- | --- |
| Woreda | Treatment | min\_GY\_kg\_ha | max\_GY\_kg\_ha | mean\_GY\_kg\_ha | sd\_GY\_kg\_ha | min\_SY\_kg\_ha | max\_SY\_kg\_ha | mean\_SY\_kg\_ha | sd\_SY\_kg\_ha |
| Bibugn | 1 | 500.00000 | 3300 | 1263.636 | 924.4163 | 100.0000 | 2200.000 | 954.5455 | 677.2941 |
| Bibugn | 2 | 500.00000 | 4800 | 1610.000 | 1256.4942 | 400.0000 | 3200.000 | 1280.0000 | 871.5248 |
| Bibugn | 3 | 500.00000 | 4000 | 2190.909 | 1021.2292 | 500.0000 | 5000.000 | 2309.0909 | 1176.8216 |
| Bibugn | 4 | 300.00000 | 4500 | 2309.091 | 1134.4202 | 500.0000 | 5000.000 | 2263.6364 | 1140.4146 |
| Bibugn | 5 | 500.00000 | 4800 | 2108.333 | 1043.1232 | 500.0000 | 3700.000 | 1983.3333 | 854.7550 |
| Bibugn | 6 | 700.00000 | 4700 | 1645.455 | 1175.0435 | 600.0000 | 4300.000 | 1463.6364 | 1085.6083 |
| Bibugn | 7 | 500.00000 | 5000 | 1616.667 | 1305.8145 | 400.0000 | 2500.000 | 1316.6667 | 784.8953 |
| Bibugn | 8 | 1000.00000 | 3500 | 2018.182 | 917.4074 | 1000.0000 | 2900.000 | 1845.4545 | 612.1497 |
| Bure Zuria | 1 | 300.00000 | 1350 | 775.000 | 396.2154 | 1000.0000 | 3000.000 | 1856.2500 | 679.7732 |
| Bure Zuria | 2 | 390.00000 | 1450 | 731.875 | 309.7142 | 1130.0000 | 4150.000 | 2083.1250 | 942.4558 |
| Bure Zuria | 3 | 390.00000 | 3100 | 1876.875 | 916.4258 | 1880.0000 | 6900.000 | 4502.5000 | 1878.0859 |
| Bure Zuria | 4 | 600.00000 | 3350 | 1652.500 | 788.5979 | 2130.0000 | 7350.000 | 4028.1250 | 1534.6410 |
| Bure Zuria | 5 | 380.00000 | 3250 | 1676.250 | 787.5521 | 1500.0000 | 7550.000 | 3962.5000 | 1637.2640 |
| Bure Zuria | 6 | 58.41584 | 2500 | 1080.526 | 632.6626 | 186.1386 | 5700.000 | 2625.3837 | 1362.4492 |
| Bure Zuria | 7 | 320.00000 | 2450 | 1210.625 | 672.2149 | 1400.0000 | 6000.000 | 2921.2500 | 1422.6495 |
| Bure Zuria | 8 | 420.00000 | 2650 | 1606.875 | 784.5739 | 1500.0000 | 6200.000 | 3779.3750 | 1594.4967 |
| Leka Dulecha | 1 | 216.00000 | 1775 | 815.900 | 495.3415 | 712.0000 | 3280.000 | 2082.9333 | 777.3679 |
| Leka Dulecha | 2 | 352.00000 | 2400 | 1200.424 | 659.5437 | 1344.0000 | 7200.000 | 2888.7879 | 1708.3628 |
| Leka Dulecha | 3 | 1253.33333 | 5200 | 2538.242 | 1414.6583 | 1384.0000 | 11033.333 | 5482.5455 | 2624.6900 |
| Leka Dulecha | 4 | 946.66667 | 4080 | 2262.667 | 1180.0381 | 2186.6667 | 11893.333 | 5141.3636 | 2637.7419 |
| Leka Dulecha | 5 | 696.00000 | 4920 | 2116.182 | 1459.4219 | 2520.0000 | 9666.667 | 4921.0606 | 2097.4534 |
| Leka Dulecha | 6 | 800.00000 | 3225 | 1606.636 | 835.6546 | 2200.0000 | 5600.000 | 3600.5455 | 865.6136 |
| Leka Dulecha | 7 | 600.00000 | 4300 | 1777.242 | 1241.7885 | 1800.0000 | 6000.000 | 3740.7273 | 1245.5335 |
| Leka Dulecha | 8 | 432.00000 | 4235 | 2025.333 | 1313.4315 | 2820.0000 | 6500.000 | 4273.9000 | 1225.0120 |

#yield\_trt\_summary  
#with(crop\_data, desc(Treatment, GrainYield, ylab = "Grain Yield ", xlab = "Treatment", ylim = NA))

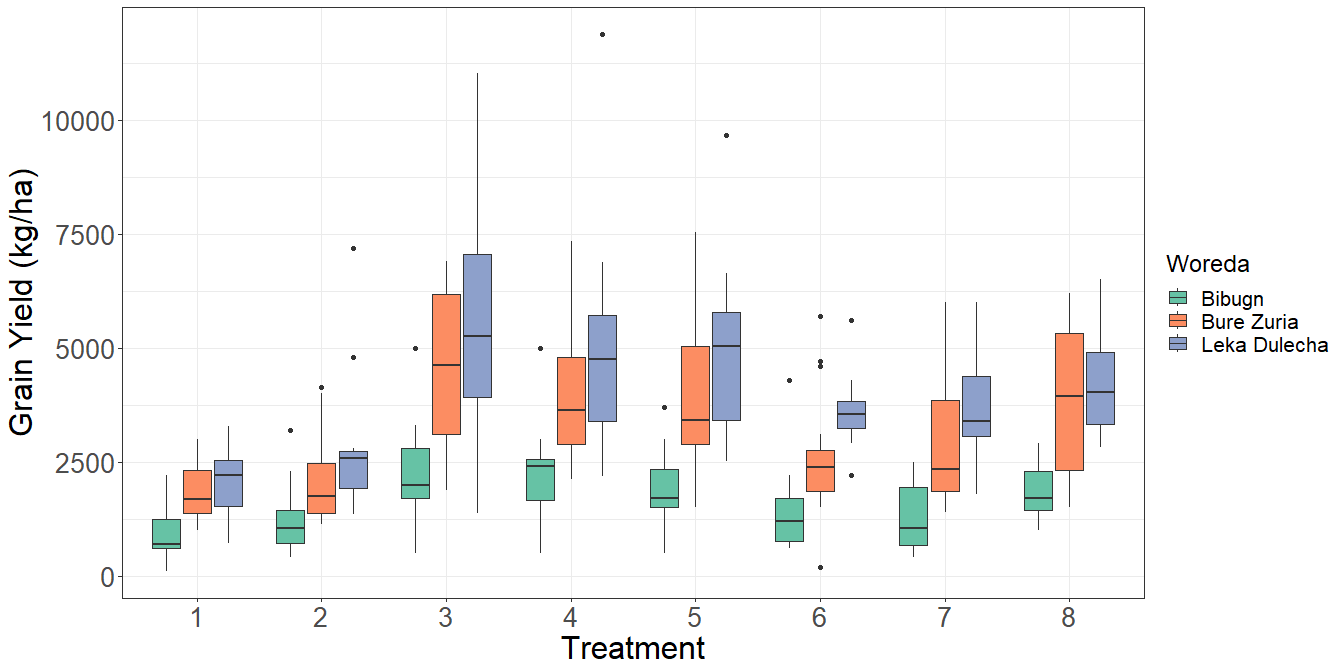
* Grain

#boxplot  
crop\_data %>% select(GY\_kg\_ha, Woreda,Treatment) %>%   
ggplot(aes(x=Treatment,y=GY\_kg\_ha, fill=Woreda)) +  
 geom\_boxplot() +  
 scale\_fill\_brewer(palette = "Set1") +  
 labs(y="Grain Yield (kg/ha)")+  
 theme\_bw()+  
 theme(legend.title = element\_text(size = 18),  
 legend.text = element\_text(size = 16),  
 axis.title = element\_text(size = 24),  
 axis.text = element\_text(size = 20))



* Stover

#boxplot  
crop\_data %>% select(SY\_kg\_ha, Woreda,Treatment) %>%   
ggplot(aes(x=Treatment,y=SY\_kg\_ha, fill=Woreda)) +  
 geom\_boxplot() +  
 scale\_fill\_brewer(palette = "Set2") +  
 labs(y="Grain Yield (kg/ha)")+  
 theme\_bw()+  
 theme(legend.title = element\_text(size = 18),  
 legend.text = element\_text(size = 16),  
 axis.title = element\_text(size = 24),  
 axis.text = element\_text(size = 20))



## Factorial analysis

# Analysis of variance  
  
#  
if (length(unique(crop\_data$Woreda))<=1) {  
 model <- lm(GY\_kg\_ha~ Treatment, data=crop\_data)  
 anova(model)  
} else {  
 model <- lm(GY\_kg\_ha~ Treatment + Woreda + (Treatment:Woreda) + (Woreda:Rep),data=crop\_data)  
 anova(model)  
}

## Analysis of Variance Table  
##   
## Response: GY\_kg\_ha  
## Df Sum Sq Mean Sq F value Pr(>F)   
## Treatment 7 52485595 7497942 9.5269 1.376e-10 \*\*\*  
## Woreda 2 18013080 9006540 11.4437 1.688e-05 \*\*\*  
## Treatment:Woreda 14 3665685 261835 0.3327 0.9893   
## Woreda:Rep 6 40905303 6817551 8.6624 1.226e-08 \*\*\*  
## Residuals 273 214859575 787031   
## ---  
## Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

## Posthoc test

# Compute the estimated marginal means  
emmeans\_result <- emmeans(model, ~ Treatment)

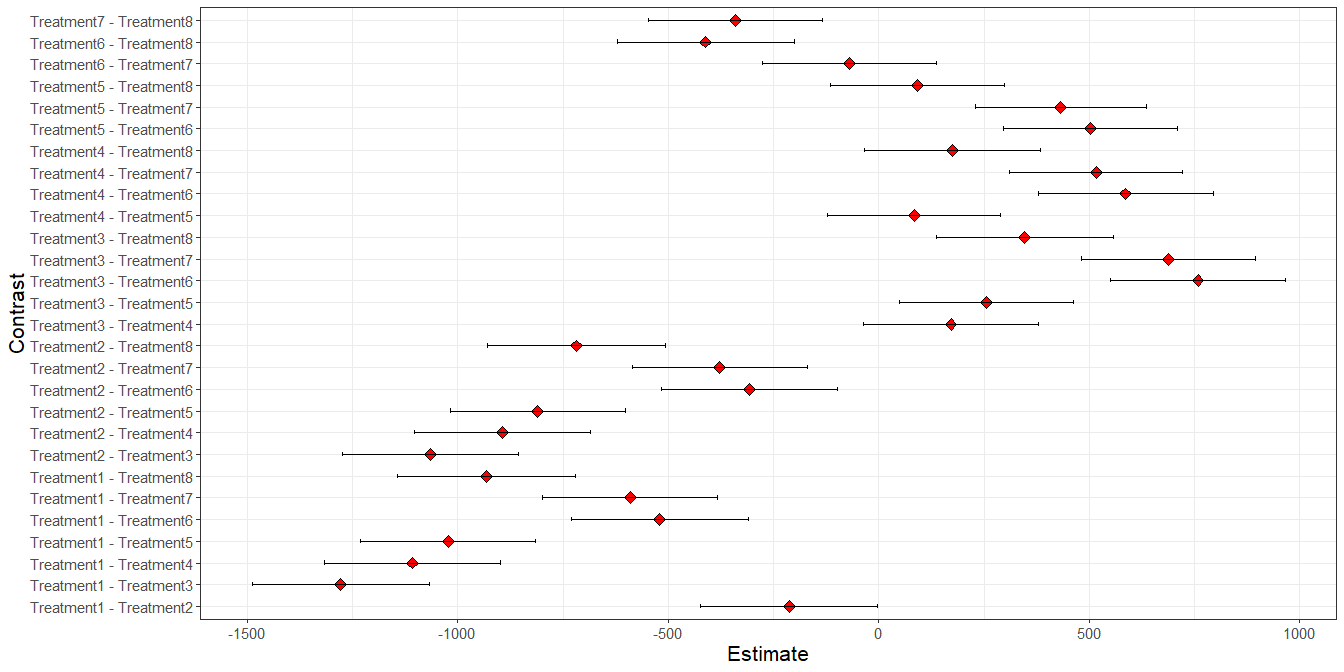
## NOTE: A nesting structure was detected in the fitted model:  
## Rep %in% Woreda

## NOTE: Results may be misleading due to involvement in interactions

# Perform pairwise comparisons using Tukey's adjustment  
pairwise\_result <- pairs(emmeans\_result, adjust = "tukey")  
  
# Display the pairwise comparisons  
knitr::kable(pairwise\_result)

|  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- |
| contrast | estimate | SE | df | t.ratio | p.value |
| Treatment1 - Treatment2 | -213.20143 | 210.9984 | 273 | -1.0104408 | 0.9726136 |
| Treatment1 - Treatment3 | -1278.38325 | 210.0544 | 273 | -6.0859625 | 0.0000001 |
| Treatment1 - Treatment4 | -1107.18754 | 209.1061 | 273 | -5.2948600 | 0.0000068 |
| Treatment1 - Treatment5 | -1022.99090 | 207.0419 | 273 | -4.9409858 | 0.0000369 |
| Treatment1 - Treatment6 | -520.58005 | 210.0544 | 273 | -2.4783105 | 0.2092149 |
| Treatment1 - Treatment7 | -590.91388 | 207.0419 | 273 | -2.8540792 | 0.0863763 |
| Treatment1 - Treatment8 | -931.95126 | 210.5236 | 273 | -4.4268248 | 0.0003656 |
| Treatment2 - Treatment3 | -1065.18182 | 209.5775 | 273 | -5.0825208 | 0.0000189 |
| Treatment2 - Treatment4 | -893.98611 | 208.6270 | 273 | -4.2850932 | 0.0006612 |
| Treatment2 - Treatment5 | -809.78947 | 207.4344 | 273 | -3.9038333 | 0.0029737 |
| Treatment2 - Treatment6 | -307.37861 | 209.5775 | 273 | -1.4666587 | 0.8244636 |
| Treatment2 - Treatment7 | -377.71245 | 207.4344 | 273 | -1.8208763 | 0.6064152 |
| Treatment2 - Treatment8 | -718.74983 | 210.9984 | 273 | -3.4064224 | 0.0170737 |
| Treatment3 - Treatment4 | 171.19571 | 207.6722 | 273 | 0.8243555 | 0.9916272 |
| Treatment3 - Treatment5 | 255.39235 | 206.4741 | 273 | 1.2369220 | 0.9201884 |
| Treatment3 - Treatment6 | 757.80321 | 206.7129 | 273 | 3.6659688 | 0.0070801 |
| Treatment3 - Treatment7 | 687.46937 | 206.4741 | 273 | 3.3295671 | 0.0218515 |
| Treatment3 - Treatment8 | 346.43199 | 210.0544 | 273 | 1.6492488 | 0.7197609 |
| Treatment4 - Treatment5 | 84.19664 | 205.5093 | 273 | 0.4096975 | 0.9999091 |
| Treatment4 - Treatment6 | 586.60750 | 207.6722 | 273 | 2.8246803 | 0.0931896 |
| Treatment4 - Treatment7 | 516.27367 | 205.5093 | 273 | 2.5121671 | 0.1947069 |
| Treatment4 - Treatment8 | 175.23628 | 209.1061 | 273 | 0.8380257 | 0.9907561 |
| Treatment5 - Treatment6 | 502.41085 | 206.4741 | 273 | 2.4332875 | 0.2296232 |
| Treatment5 - Treatment7 | 432.07702 | 203.4829 | 273 | 2.1234075 | 0.4025162 |
| Treatment5 - Treatment8 | 91.03964 | 207.0419 | 273 | 0.4397161 | 0.9998537 |
| Treatment6 - Treatment7 | -70.33383 | 206.4741 | 273 | -0.3406424 | 0.9999741 |
| Treatment6 - Treatment8 | -411.37122 | 210.0544 | 273 | -1.9584032 | 0.5120470 |
| Treatment7 - Treatment8 | -341.03738 | 207.0419 | 273 | -1.6471905 | 0.7210522 |

#plot(pairwise\_result, pch = "\*", cex = 1.5)  
as\_tibble(pairwise\_result) %>%   
ggplot(aes(x=estimate,y=contrast)) +  
 geom\_point(shape=23, size = 3, fill = "red") +  
 geom\_errorbar(aes(xmin = estimate - SE, xmax = estimate + SE),  
 width = 0.2) +  
 labs(x = "Estimate", y = "Contrast")+  
 theme\_bw()+  
 theme(axis.title = element\_text(size = 16),  
 axis.text = element\_text(size = 11))



## Best treatment per location

### For every woreda, what are the 3 best treatment?

* Grain

best\_comb<-yield\_location\_trt %>% select(Woreda, Treatment, mean\_GY\_kg\_ha) %>%   
 group\_by(Woreda) %>%  
 slice\_max(mean\_GY\_kg\_ha, n=3)  
knitr::kable(best\_comb)

|  |  |  |
| --- | --- | --- |
| Woreda | Treatment | mean\_GY\_kg\_ha |
| Bibugn | 4 | 2309.091 |
| Bibugn | 3 | 2190.909 |
| Bibugn | 5 | 2108.333 |
| Bure Zuria | 3 | 1876.875 |
| Bure Zuria | 5 | 1676.250 |
| Bure Zuria | 4 | 1652.500 |
| Leka Dulecha | 3 | 2538.242 |
| Leka Dulecha | 4 | 2262.667 |
| Leka Dulecha | 5 | 2116.182 |

* Stover

best\_comb<-yield\_location\_trt %>% select(Woreda, Treatment, mean\_SY\_kg\_ha) %>%   
 group\_by(Woreda) %>%  
 slice\_max(mean\_SY\_kg\_ha, n=3)  
knitr::kable(best\_comb)

|  |  |  |
| --- | --- | --- |
| Woreda | Treatment | mean\_SY\_kg\_ha |
| Bibugn | 3 | 2309.091 |
| Bibugn | 4 | 2263.636 |
| Bibugn | 5 | 1983.333 |
| Bure Zuria | 3 | 4502.500 |
| Bure Zuria | 4 | 4028.125 |
| Bure Zuria | 5 | 3962.500 |
| Leka Dulecha | 3 | 5482.545 |
| Leka Dulecha | 4 | 5141.364 |
| Leka Dulecha | 5 | 4921.061 |