

Touch Paper

The analysis we did of the stats behind figure 1 was sub-standard. The model fitted should be a two-factor linear mixed effect model (Genotype * treatment are the two factors, w/ interaction), with a random effect term for the tray (batch) effect. This removes the error in each of the two model factors caused by tray effect whilst fitting the model.

So, lets get the data:

```
fig1_data <- read.csv("TouchData.csv")
fig1_data$Genotype <- relevel(fig1_data$Genotype, ref="WT")
summary(fig1_data)
```

```
##      Genotype      Treatment      Tray      Rep
## WT      :252 NotTouched:215 Min.    :1.00 Min.    : 1.0
## ccr1.1:197 Touched   :234 1st Qu.:1.00 1st Qu.:10.0
##                                     Median :2.00 Median :19.0
##                                     Mean   :2.01 Mean   :19.6
##                                     3rd Qu.:3.00 3rd Qu.:29.0
##                                     Max.   :3.00 Max.   :42.0
##
##      Measurement      Value
## LeafLength   :150 Min.    : 4.05
## LeafWidth    :149 1st Qu.:10.35
## PetioleLength:150 Median :12.58
##                                     Mean   :14.10
##                                     3rd Qu.:18.33
##                                     Max.   :27.17
```

Now, for each phenotype, we want to fit the linear mixed effects model. We'll need the package nlme, which contains the lme function we use below.

```
library(nlme)

phenotypes <- unique(as.character(fig1_data$Measurement))
for (pheno in phenotypes) {
  pdata <- fig1_data[fig1_data$Measurement == pheno,]
  fit <- lme(Value ~ Genotype * Treatment, data = pdata, random =~ 1|Tray)
  print(pheno)
  print(summary(fit))
  # g.fit <- tapply(predict(fit),paste(gp$Genotype,gp$Treatment),mean)
  # oo <- barplot(g.fit,ylim = c(0,20),main = "PetLength")
  # segments(oo,g.fit,oo,g.fit+fit$sigma)
  # segments(oo,g.fit,oo,g.fit-fit$sigma)
}
```

```
## [1] "LeafLength"
## Linear mixed-effects model fit by REML
## Data: pdata
##      AIC BIC logLik
## 814.1 832 -401.1
##
## Random effects:
## Formula: ~1 | Tray
```

```

##          (Intercept) Residual
## StdDev:    0.000173    3.592
##
## Fixed effects: Value ~ Genotype * Treatment
##
##          Value Std.Error  DF t-value p-value
## (Intercept)    22.302    0.5542 144   40.24  0.0000
## Genotypeccr1.1    -6.132    0.8585 144   -7.14  0.0000
## TreatmentTouched    -1.301    0.7837 144   -1.66  0.0990
## Genotypeccr1.1:TreatmentTouched    3.438    1.1843 144    2.90  0.0043
## Correlation:
##
##          (Intr) Gnt1.1 TrtmnT
## Genotypeccr1.1    -0.645
## TreatmentTouched    -0.707  0.456
## Genotypeccr1.1:TreatmentTouched    0.468 -0.725 -0.662
##
## Standardized Within-Group Residuals:
##      Min      Q1      Med      Q3      Max
## -3.18435 -0.56608  0.08486  0.62879  2.01400
##
## Number of Observations: 150
## Number of Groups: 3
## [1] "PetioleLength"
## Linear mixed-effects model fit by REML
## Data: pdata
##      AIC      BIC logLik
##  661.6  679.5 -324.8
##
## Random effects:
## Formula: ~1 | Tray
##          (Intercept) Residual
## StdDev:    0.8381    2.098
##
## Fixed effects: Value ~ Genotype * Treatment
##
##          Value Std.Error  DF t-value p-value
## (Intercept)    15.098    0.5823 144   25.928  0.0000
## Genotypeccr1.1    -3.706    0.5030 144   -7.368  0.0000
## TreatmentTouched    -4.568    0.4580 144   -9.975  0.0000
## Genotypeccr1.1:TreatmentTouched    1.998    0.6937 144    2.880  0.0046
## Correlation:
##
##          (Intr) Gnt1.1 TrtmnT
## Genotypeccr1.1    -0.358
## TreatmentTouched    -0.393  0.456
## Genotypeccr1.1:TreatmentTouched    0.260 -0.726 -0.661
##
## Standardized Within-Group Residuals:
##      Min      Q1      Med      Q3      Max
## -3.05788 -0.50321  0.02794  0.67962  2.54746
##
## Number of Observations: 150
## Number of Groups: 3
## [1] "LeafWidth"
## Linear mixed-effects model fit by REML
## Data: pdata
##      AIC      BIC logLik

```

```
##    556.5 574.3 -272.2
##
## Random effects:
## Formula: ~1 | Tray
##      (Intercept) Residual
## StdDev:  4.082e-05    1.505
##
## Fixed effects: Value ~ Genotype * Treatment
##
##              Value Std.Error  DF t-value p-value
## (Intercept)    12.731    0.2322 143   54.82  0e+00
## Genotypeccr1.1    -3.383    0.3633 143   -9.31  0e+00
## TreatmentTouched    -1.366    0.3284 143   -4.16  1e-04
## Genotypeccr1.1:TreatmentTouched  1.801    0.4989 143    3.61  4e-04
## Correlation:
##              (Intr) Gnt1.1 TrtmnT
## Genotypeccr1.1    -0.639
## TreatmentTouched    -0.707  0.452
## Genotypeccr1.1:TreatmentTouched  0.465 -0.728 -0.658
##
## Standardized Within-Group Residuals:
##      Min      Q1      Med      Q3      Max
## -2.9119 -0.5388  0.1646  0.7409  2.0657
##
## Number of Observations: 149
## Number of Groups: 3
```

So, we have the following “ANOVA” table (though this is NOT an anova). The values are the size of the effect (in mm) and the p-value associated with the effect. This is all relative to non-touched wild-type plants.

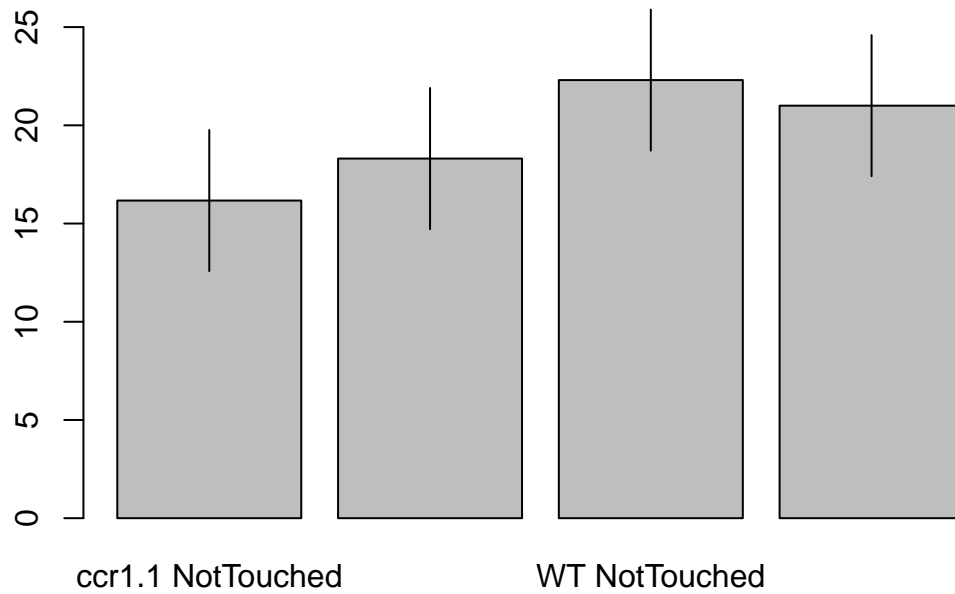
Phenotype	Genotype	Treatment	Genotype:Treatment
Leaf Len	-6.1 (0.00)	-1.3 (0.09)	3.4 (0.00)
Leaf Width	-3.4 (0.00)	-1.4 (0.00)	1.8 (0.00)
Petiole Len	-3.7 (0.00)	-4.6 (0.00)	2.0 (0.00)

Which is pretty sweet.

OK, now some plots.

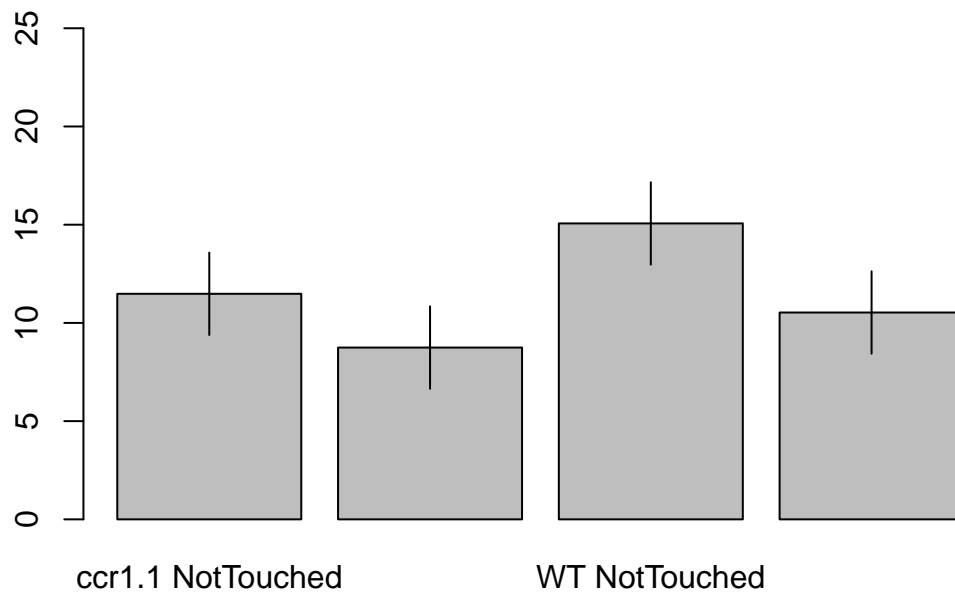
```
for (pheno in phenotypes) {
  pdata <- fig1_data[fig1_data$Measurement == pheno,]
  fit <- lme(Value ~ Genotype * Treatment, data = pdata, random =~ 1|Tray)
  fit.data <- tapply(predict(fit),paste(pdata$Genotype,pdata$Treatment),mean)
  xx <- barplot(fit.data,ylim = c(0,26),main = pheno)
  segments(xx,fit.data,xx,fit.data+fit$sigma)
  segments(xx,fit.data,xx,fit.data-fit$sigma)
  print(paste("Predicted values for", pheno))
}
```

LeafLength



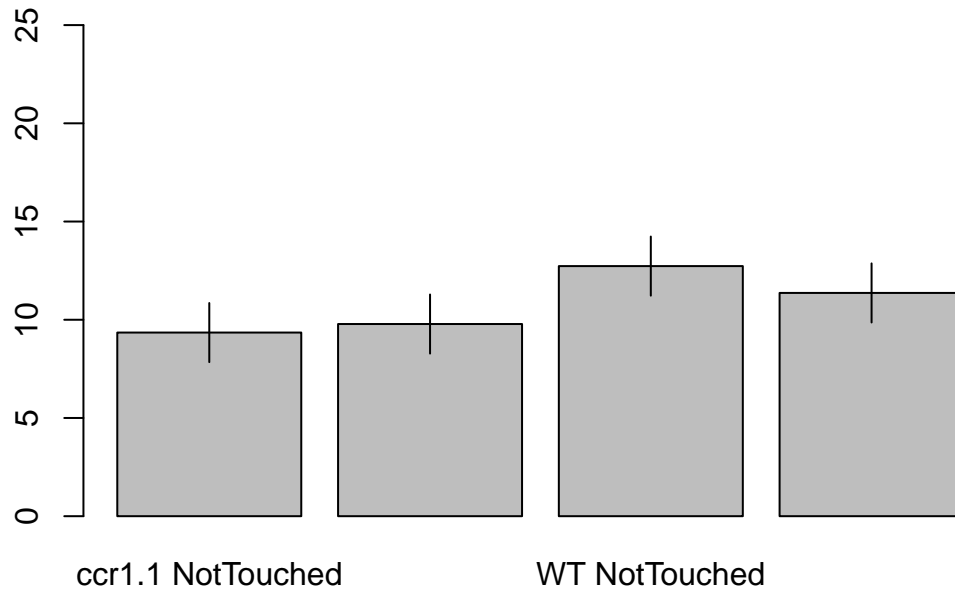
```
## [1] "Predicted values for LeafLength"
```

PetioleLength



```
## [1] "Predicted values for PetioleLength"
```

LeafWidth



[1] "Predicted values for LeafWidth"