

# DAPS Air Pollution Monitor Differences

Melissa Lowe

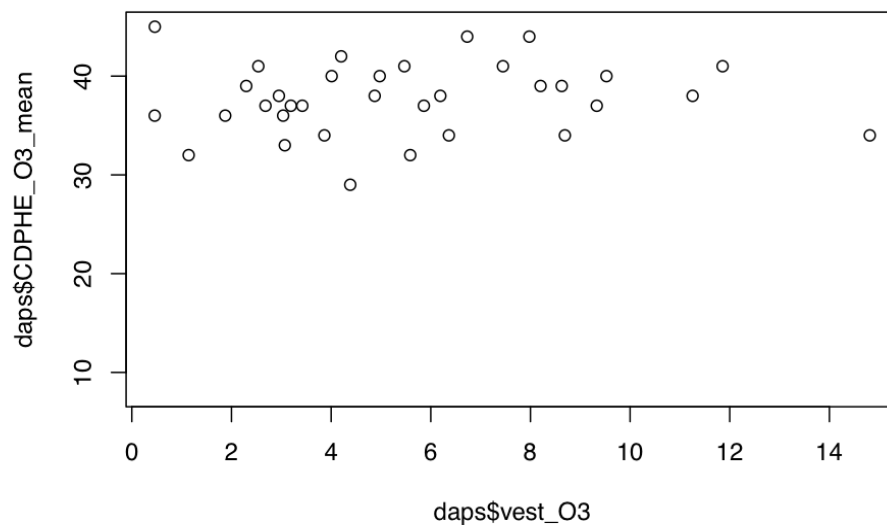
6/4/19

## Data Steps:

### Basic Correlations

```
#correlation of ozone
o3corr <- cor.test(daps$vest_O3, daps$CDPHE_O3_mean, method="pearson")
o3corr

##
## Pearson's product-moment correlation
##
## data: daps$vest_O3 and daps$CDPHE_O3_mean
## t = 0.42427, df = 32, p-value = 0.6742
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
## -0.2702091 0.4027696
## sample estimates:
## cor
## 0.0747906
plot(daps$vest_O3, daps$CDPHE_O3_mean)
```



```
#correlation of pm10
pm10corr <- rmcrr(id, log_vest_pm10, log_msp_pm10,daps)
pm10corr
```

```

##
## Repeated measures correlation
##
## r
## 0.5022392
##
## degrees of freedom
## 72
##
## p-value
## 5.131506e-06
##
## 95% confidence interval
## 0.3061977 0.6574155

pm10corr2 <- rmcrr(id, log_vest_pm10, log(daps$CDPHE_PM10_mean), daps)
pm10corr2

##
## Repeated measures correlation
##
## r
## -0.1460296
##
## degrees of freedom
## 74
##
## p-value
## 0.2081324
##
## 95% confidence interval
## -0.3624347 0.08531597

pm10corr3 <- rmcrr(id, log_msp_pm10, log(CDPHE_PM10_mean), daps)
pm10corr3

##
## Repeated measures correlation
##
## r
## -0.06262704
##
## degrees of freedom
## 73
##
## p-value
## 0.5934931
##
## 95% confidence interval
## -0.2885369 0.1698881

rmcrr(id, log(daps$CDPHE_PM10_mean), daps$log_FeNO_avg, daps)

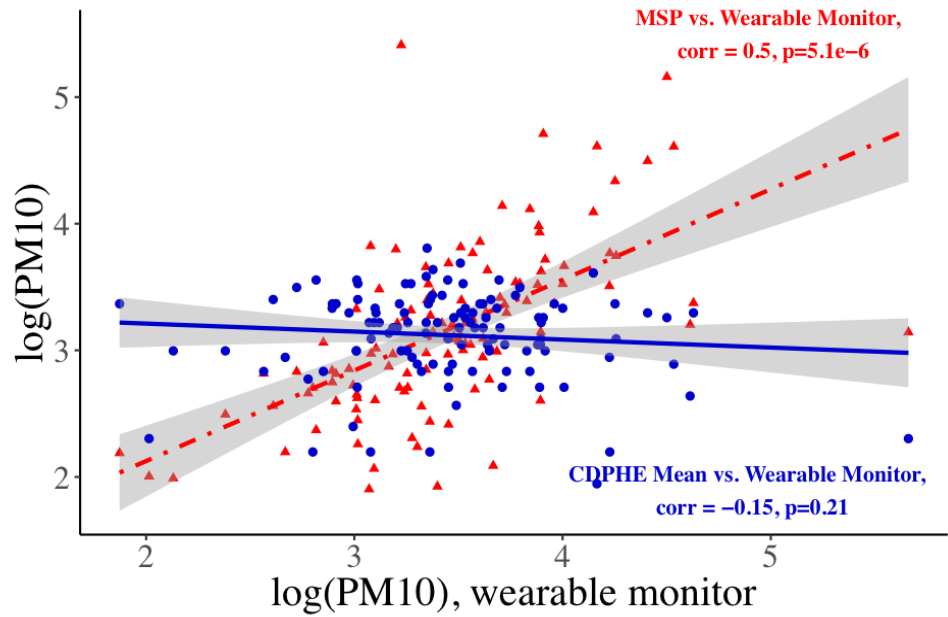
##
## Repeated measures correlation

```

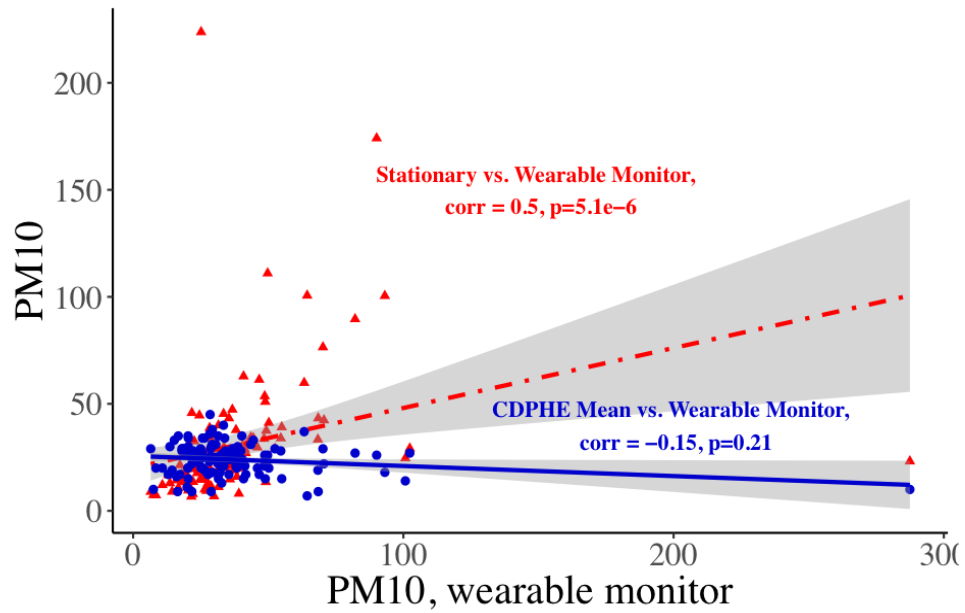
```
##
## r
## 0.3924604
##
## degrees of freedom
## 73
##
## p-value
## 0.0004966465
##
## 95% confidence interval
## 0.1785113 0.5709732
rmcorr(id, log_vest_pm10, daps$log_FeNO_avg, daps)

##
## Repeated measures correlation
##
## r
## 0.1095856
##
## degrees of freedom
## 71
##
## p-value
## 0.3560472
##
## 95% confidence interval
## -0.1269642 0.3343413
```

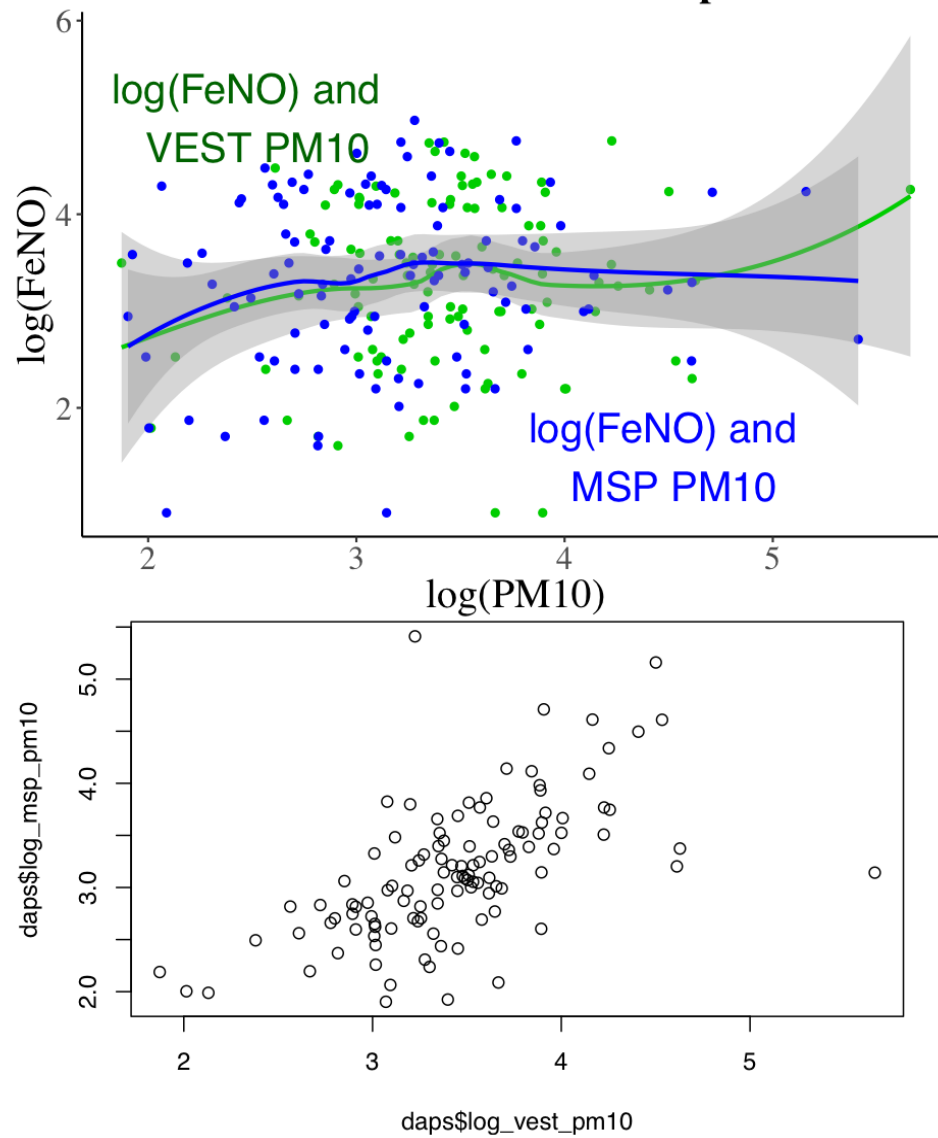
### Correlation with Wearable Monitor

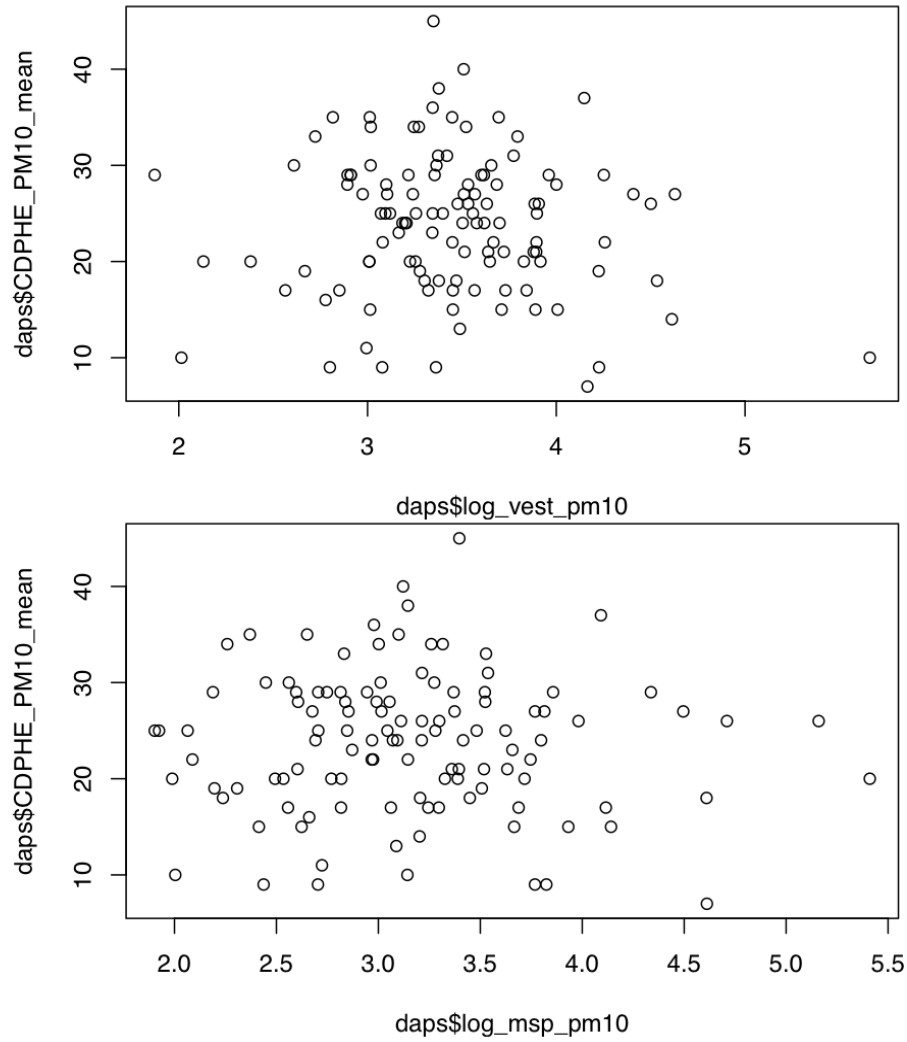


### Correlation with Wearable Monitor



## Exhaled Nitric Oxide and PM10 Exposure





```
#correlation of no2
no2corr <- rmcorr(id, log_vest_NO2, log_stat_NO2,daps)
no2corr
```

```
##
## Repeated measures correlation
##
## r
## 0.9508126
##
## degrees of freedom
```

```

## 68
##
## p-value
## 2.487073e-36
##
## 95% confidence interval
## 0.9212276 0.9694628

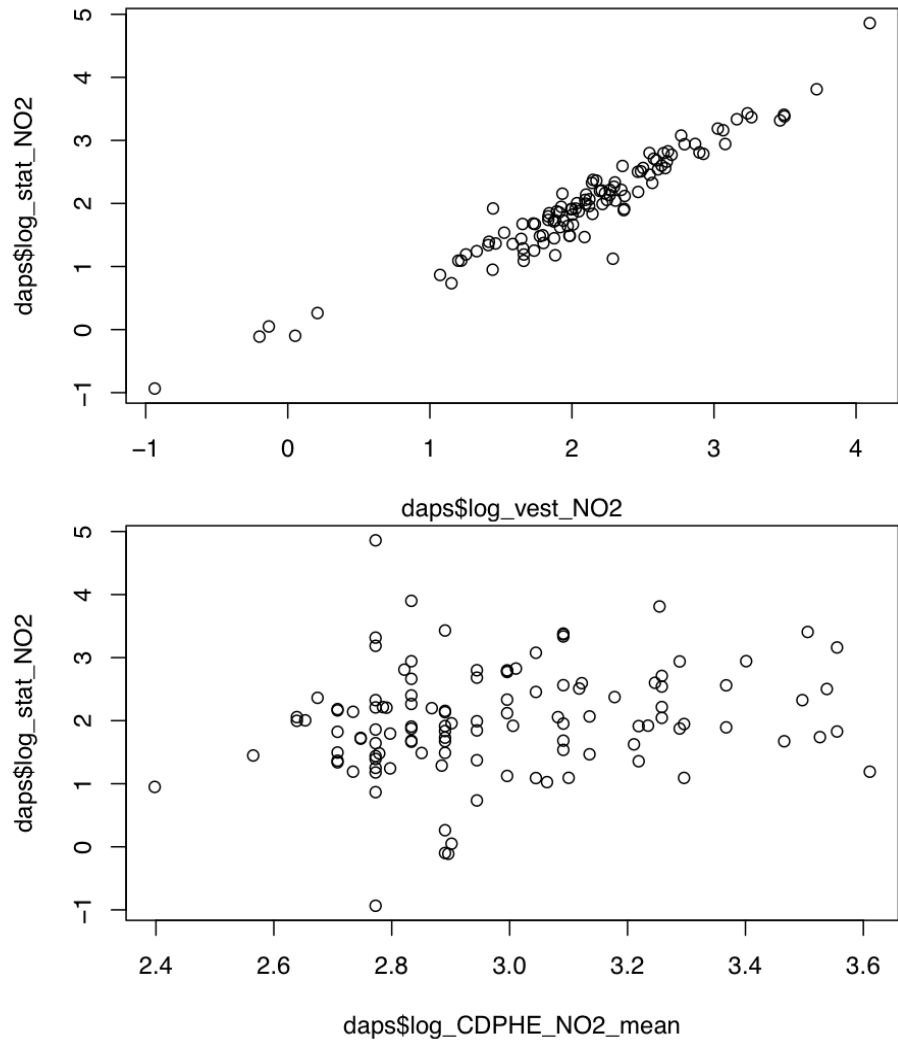
no2corr2 <- rmcorr(id, log_vest_NO2, log(log_CDPHE_NO2_mean), daps)
no2corr2

##
## Repeated measures correlation
##
## r
## 0.2338877
##
## degrees of freedom
## 68
##
## p-value
## 0.05132745
##
## 95% confidence interval
## -0.004805375 0.4473656

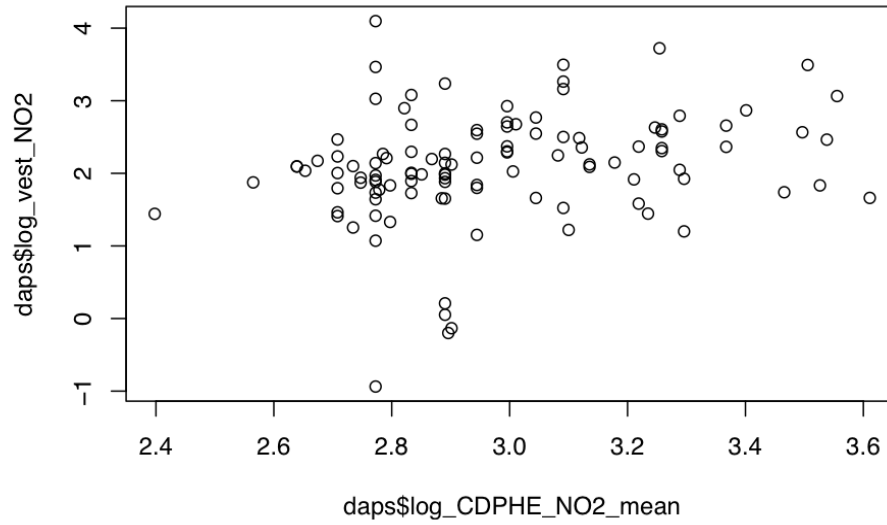
no2corr3 <- rmcorr(id, log_stat_NO2, log(log_CDPHE_NO2_mean), daps)
no2corr3

##
## Repeated measures correlation
##
## r
## 0.1900236
##
## degrees of freedom
## 75
##
## p-value
## 0.09786429
##
## 95% confidence interval
## -0.03860308 0.3997452

```

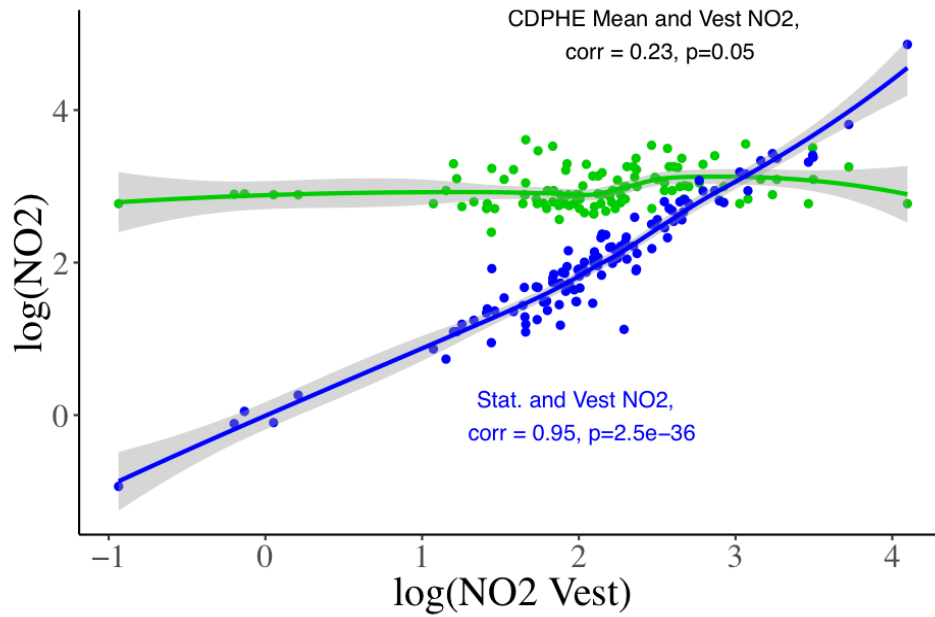




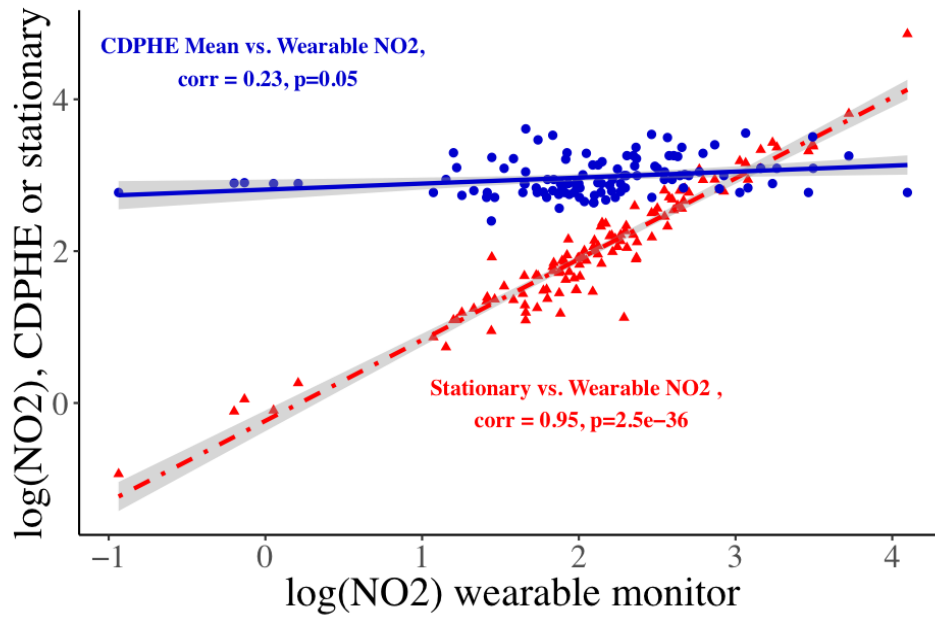


```
##
## Pearson's product-moment correlation
##
## data: daps$gluc_m3 and log(daps$endo_m3)
## t = 3.6936, df = 39, p-value = 0.0006762
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
##  0.2388288 0.7061327
## sample estimates:
##          cor
## 0.5090746
```

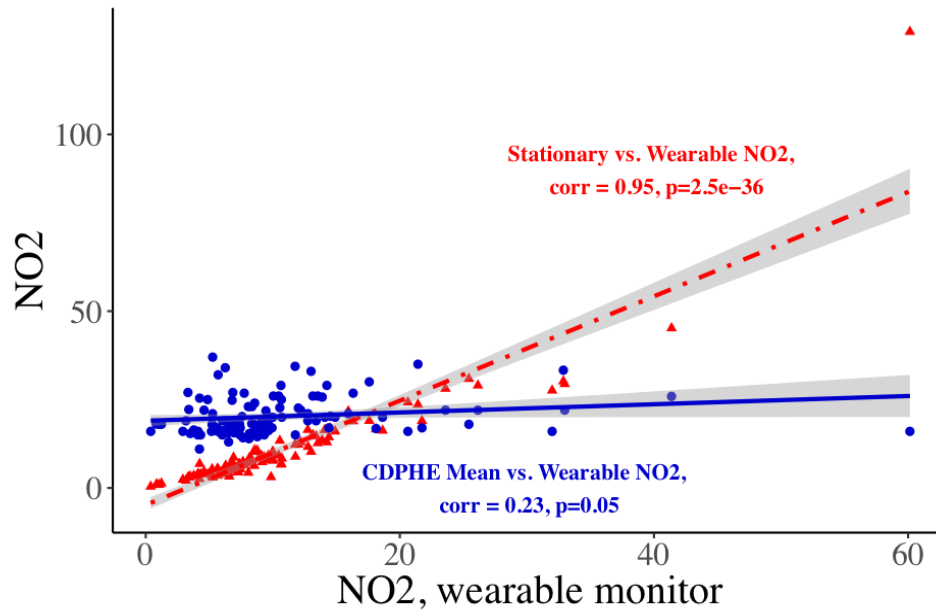
## Correlation with Vest Monitor



## Relative Correlation with Wearable Monitor



## Relative Correlation with Wearable Monitor



Ok. So, part of what we're interested in is estimating the mean bias and mean variance. If we assume that the “true” value is that from the wearable monitor, how do we estimate the mean bias? What about the mean variance?

Variance is defined as:

$$\text{Var}[\hat{x}] = E[(\hat{x} - E[\hat{x}])^2]$$

$$\text{Var}[\hat{x}] = E[\hat{x}^2] - E[\hat{x}]^2$$

Bias is defined as follows:

$$\text{Bias}[\hat{x}] = E[\hat{x} - x] = E[\hat{x}] - x$$

```
##### PM10 #####
wearable <- exp(daps$log_vest_pm10)
stationary <- exp(daps$log_msp_pm10)
cdphe_mean <- (daps$CDPHE_PM10_mean)
cdphe_max <- exp(daps$log_CDPHE_PM10_max)

PM10 <- as.data.frame(cbind(id, visit, season, wearable, stationary, cdphe_mean, cdphe_max))

pm10 <- melt(PM10, id.vars = c("id", "visit", "season"), measure.vars = c("wearable", "stationary", "cdphe_mean", "cdphe_max"),
  factorsAsStrings = TRUE)

names(pm10) <- c("id", "visit", "season", "type", "PM10")
```

```

#Mean Bias of the CDPHE MEAN
mean(mean(cdphe_mean, na.rm=TRUE) - wearable, na.rm=TRUE)

## [1] -13.23269
#Mean Variance of the CDPHE MEAN:

#Mean of the Sample Squared:
ss <- mean(cdphe_mean*cdphe_mean, na.rm=TRUE)

#Mean of the Squared Sample:
ess <- (mean(cdphe_mean, na.rm=TRUE)*mean(cdphe_mean, na.rm=TRUE))

ss-ess

## [1] 54.01609
var(cdphe_mean, na.rm=TRUE)

## [1] 54.47776
#Mean Bias of the Stationary Monitor
mean(mean(stationary, na.rm=TRUE)-wearable, na.rm=TRUE)

## [1] -6.665394
#Mean Variance of the Stationary Monitor
#Mean of the Sample Squared:
ss <- mean(stationary*stationary, na.rm=TRUE)

#Mean of the Squared Sample:
ess <- (mean(stationary, na.rm=TRUE)*mean(stationary, na.rm=TRUE))

ss-ess

## [1] 878.225
var(stationary, na.rm=TRUE)

## [1] 885.9288
mean_stat <- rep((mean(stationary, na.rm=TRUE)), length(wearable))
mean_cdphe <- rep((mean(cdphe_mean, na.rm=TRUE)), length(wearable))

dat2 <- as.data.frame(cbind(wearable, mean_stat, mean_cdphe))

#### N02 ####
wearable <- exp(daps$log_vest_N02)
stationary <- exp(daps$log_stat_N02)
cdphe_mean <- exp(daps$log_CDPHE_N02_mean)
cdphe_max <- exp(daps$log_CDPHE_N02_max)

N02 <- as.data.frame(cbind(id, visit, season, wearable, stationary, cdphe_mean, cdphe_max))

no2 <- melt(N02, id.vars = c("id", "visit", "season"), measure.vars = c("wearable", "stationary", "cdphe",
  variable.name = "type", value.name = "N02",
  factorsAsStrings = TRUE)

```

```

names(no2) <- c("id", "visit", "season", "type", "NO2")

#Mean Bias of the CDPHE MEAN
mean(mean(cdphe_mean, na.rm=TRUE) - wearable, na.rm=TRUE)

## [1] 9.806204

#Mean Variance of the CDPHE MEAN:

#Mean of the Sample Squared:
ss <- mean(cdphe_mean*cdphe_mean, na.rm=TRUE)

#Mean of the Squared Sample:
ess <- (mean(cdphe_mean, na.rm=TRUE)*mean(cdphe_mean, na.rm=TRUE))

ss-ess

## [1] 28.94186

var(cdphe_mean, na.rm=TRUE)

## [1] 29.18923

#Mean Bias of the Stationary Monitor
mean(mean(stationary, na.rm=TRUE)-wearable, na.rm=TRUE)

## [1] 0.3155416

#Mean Variance of the Stationary Monitor
#Mean of the Sample Squared:
ss <- mean(stationary*stationary, na.rm=TRUE)

#Mean of the Squared Sample:
ess <- (mean(stationary, na.rm=TRUE)*mean(stationary, na.rm=TRUE))

ss-ess

## [1] 190.236

var(stationary, na.rm=TRUE)

## [1] 191.876

mean_stat <- rep((mean(stationary, na.rm=TRUE)), length(wearable))
mean_cdphe <- rep((mean(cdphe_mean, na.rm=TRUE)), length(wearable))

dat2 <- as.data.frame(cbind(wearable, mean_stat, mean_cdphe))

#### 03 ####
wearable <- daps$vest_03
cdphe_mean <- daps$CDPHE_03_mean
cdphe_max <- daps$CDPHE_03_max

03 <- as.data.frame(cbind(id, visit, season, wearable, cdphe_mean, cdphe_max))

o3 <- melt(03, id.vars = c("id", "visit", "season"), measure.vars = c("wearable", "cdphe_mean", "cdphe_max"),
  variable.name = "type", value.name = "03",
  factorsAsStrings = TRUE)

```

```

names(o3) <- c("id", "visit", "season", "type", "03")

#Mean Bias of the CDPHE MEAN
mean(mean(cdphe_mean, na.rm=TRUE) - wearable, na.rm=TRUE)

## [1] 21.16721

#Mean Variance of the CDPHE MEAN:

#Mean of the Sample Squared:
ss <- mean(cdphe_mean*cdphe_mean, na.rm=TRUE)

#Mean of the Squared Sample:
ess <- (mean(cdphe_mean, na.rm=TRUE)*mean(cdphe_mean, na.rm=TRUE))

ss-ess

## [1] 93.15053

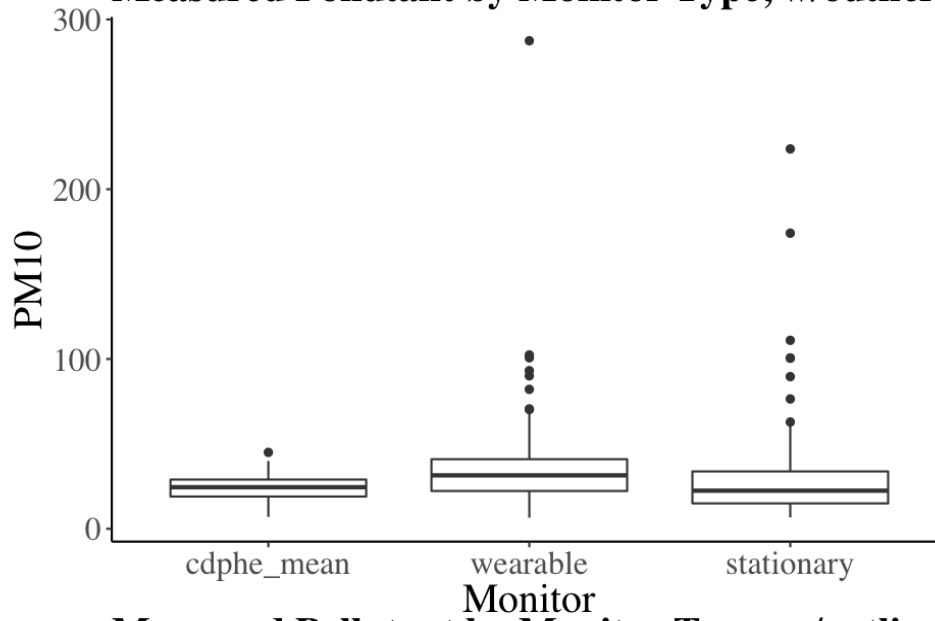
var(cdphe_mean, na.rm=TRUE)

## [1] 93.94669

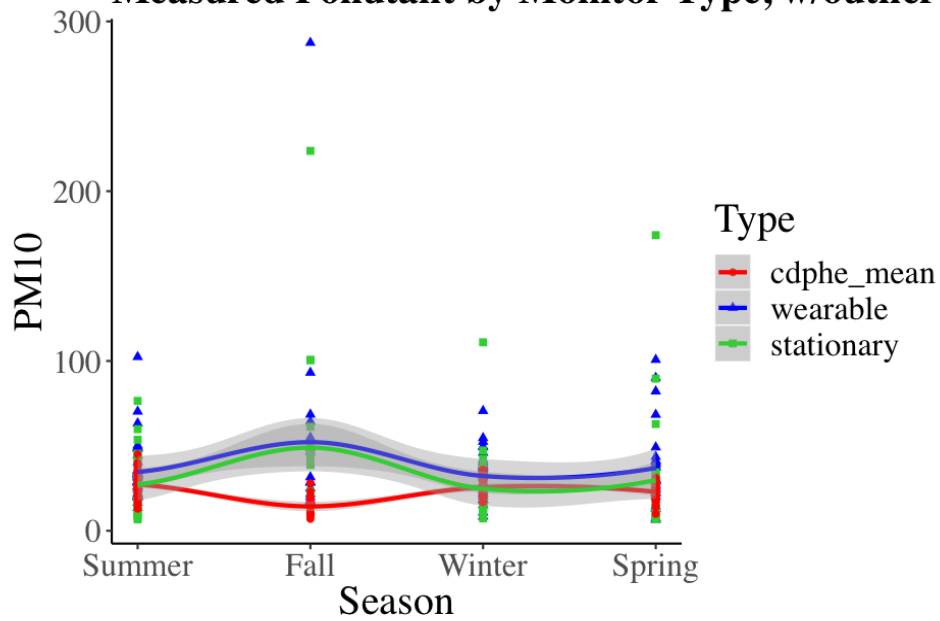
```

## PM10, MEASUREMENT DIFFERENCES

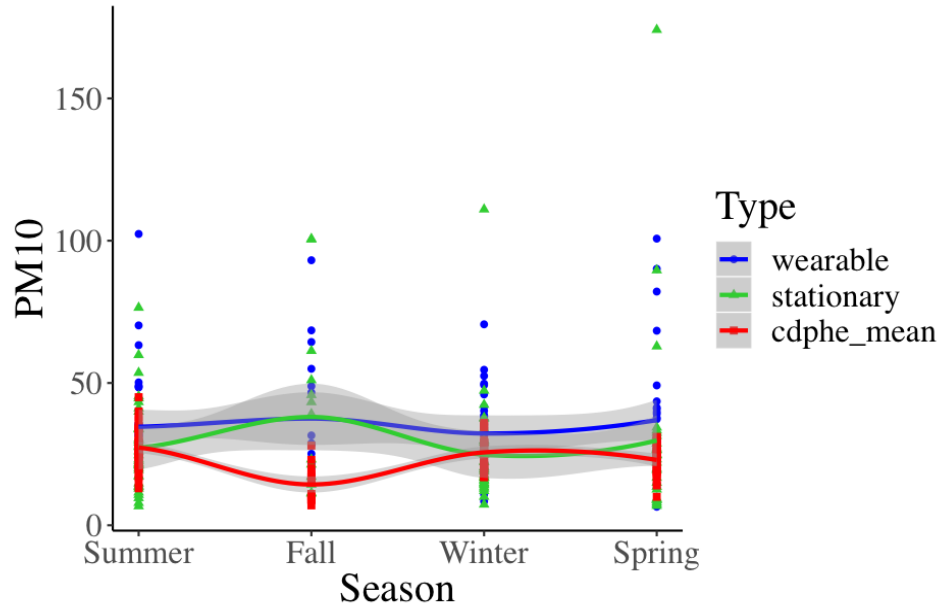
**Measured Pollutant by Monitor Type, w/outlier**



**Measured Pollutant by Monitor Type, w/outlier**



## Measured Pollutant by Monitor Type, w/o outlier



```
##
## PM10 Monitor Differences
## =====
##                               Dependent variable:
##                               -----
##                               PM10
##                               (1)      (2)
## -----
```

Vest	7.5589***	-8.8388***
	(3.9383, 11.1796)	(-12.4323, -5.2452)
	t = 4.0918	t = -4.8208
	p = 0.00005	p = 0.000002
MSP	1.2798	7.5589***
	(-2.3426, 4.9023)	(3.9383, 11.1796)
	t = 0.6925	t = 4.0918
	p = 0.4887	p = 0.00005
CDPHE Mean	-1.5252	-1.5252
	(-5.5129, 2.4626)	(-5.5129, 2.4626)
	t = -0.7496	t = -0.7496
	p = 0.4535	p = 0.4535
Fall	7.2817**	7.2817**
	(1.8978, 12.6656)	(1.8978, 12.6656)
	t = 2.6508	t = 2.6508
	p = 0.0081	p = 0.0081
Winter	-3.5210	-3.5210
	(-7.7069, 0.6650)	(-7.7069, 0.6650)
	t = -1.6486	t = -1.6486

```
##
```



```

##               p = 0.0993      p = 0.0993
## Spring        -2.6651      6.4351*
##               (-8.2757, 2.9455) (0.8419, 12.0283)
##               t = -0.9310      t = 2.2550
##               p = 0.3519      p = 0.0242
## Vest*Fall     -3.7700      -2.6651
##               (-9.4048, 1.8647) (-8.2757, 2.9455)
##               t = -1.3114      t = -0.9310
##               p = 0.1898      p = 0.3519
## MSP*Fall      6.0754      -15.2129***
##               (-1.2472, 13.3979) (-22.4476, -7.9782)
##               t = 1.6261      t = -4.1214
##               p = 0.1040      p = 0.00004
## CDPHE*Fall    9.1376*      6.0754
##               (1.8143, 16.4609) (-1.2472, 13.3979)
##               t = 2.4455      t = 1.6261
##               p = 0.0145      p = 0.1040
## Vest*Winter   -2.8868      6.8615*
##               (-8.6878, 2.9142) (1.0774, 12.6457)
##               t = -0.9754      t = 2.3250
##               p = 0.3294      p = 0.0201
## MSP*Winter    -3.9747      -2.8868
##               (-9.8047, 1.8553) (-8.6878, 2.9142)
##               t = -1.3363      t = -0.9754
##               p = 0.1815      p = 0.3294
## CDPHE*Winter  31.3133***    31.3133***
##               (28.1364, 34.4903) (28.1364, 34.4903)
##               t = 19.3181      t = 19.3181
##               p = 0.0000      p = 0.0000
## -----
## Observations      349      349
## Log Likelihood    -1,576.6580 -1,576.6580
## Akaike Inf. Crit.  3,181.3160  3,181.3160
## Bayesian Inf. Crit. 3,235.2870  3,235.2870
## =====
## Note:              *p<0.05; **p<0.01; ***p<0.001

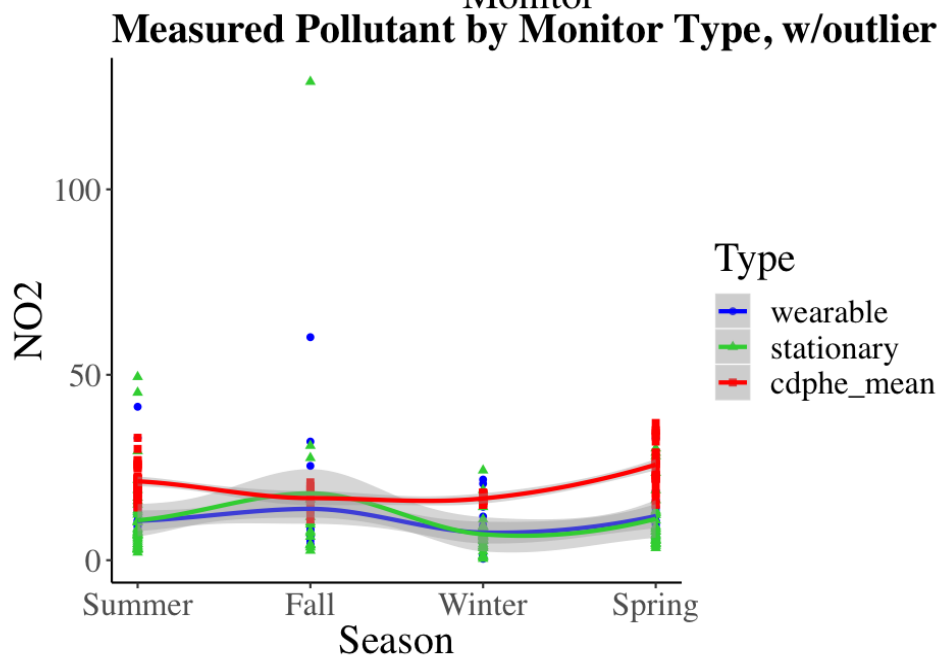
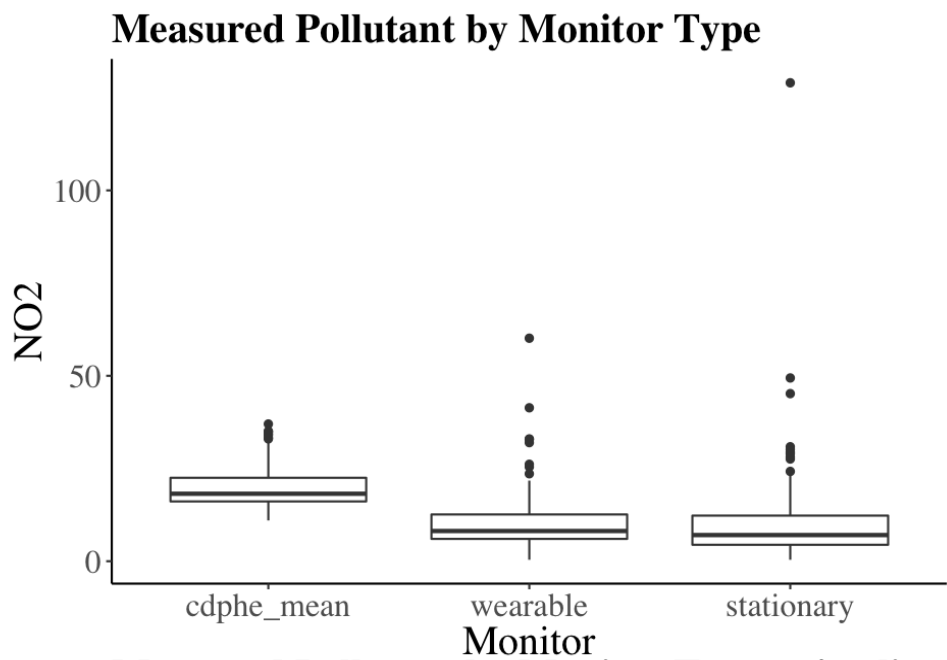
```

## NO2 MEASUREMENT DIFFERENCES

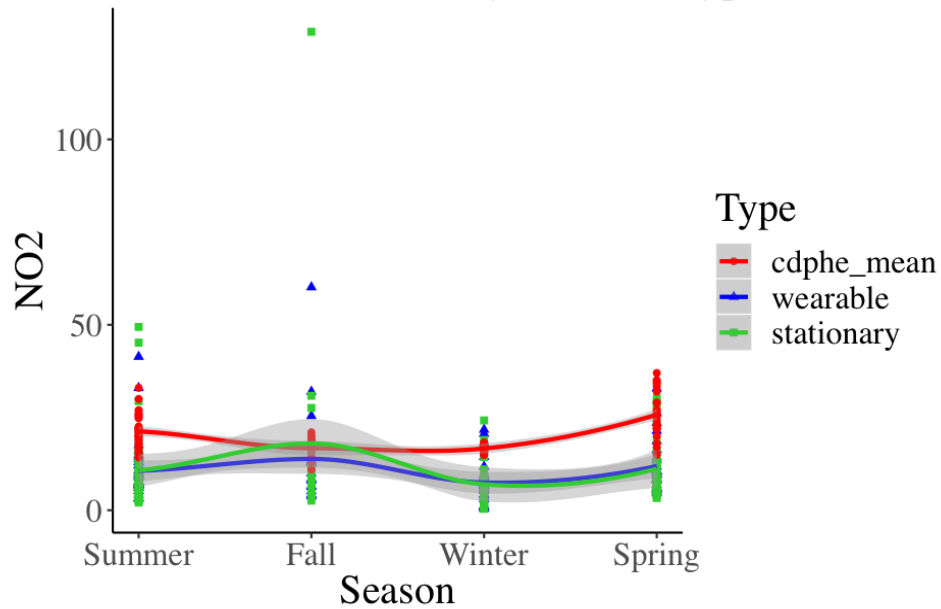
```

## [1] "cdphe_mean" "wearable" "stationary" "cdphe_max"
## [1] "wearable" "stationary" "cdphe_mean" "cdphe_max"

```



## Measured Pollutant by Monitor Type, w/o outlier



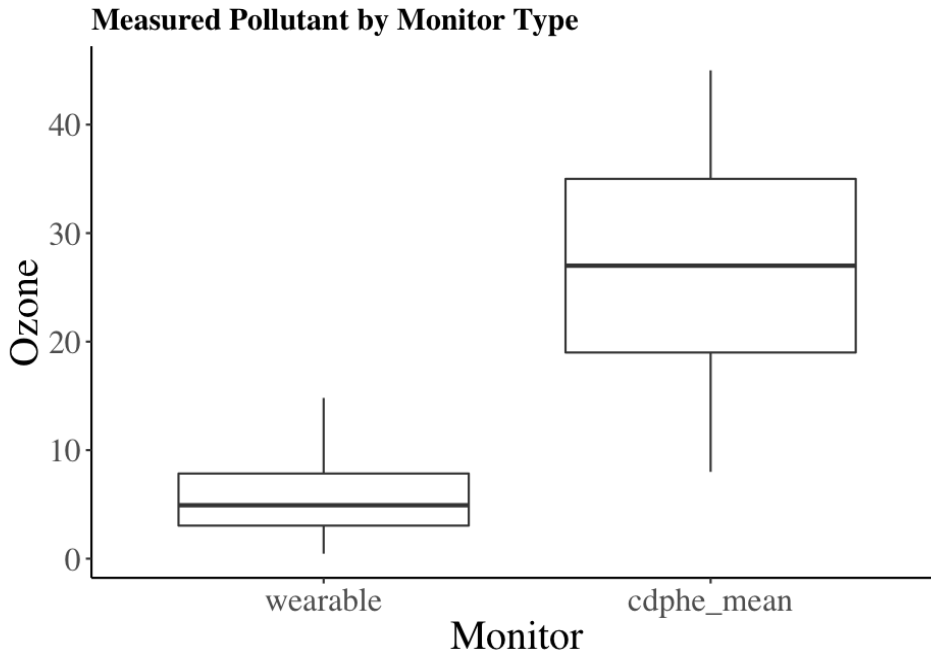
```
##
## NO2 Monitor Differences
## =====
##                               Dependent variable:
##                               -----
##                               N02
##                               (1)      (2)
## -----
## Wearable      -3.2763***      5.8460***
##                (-4.5232, -2.0294) (4.6212, 7.0709)
##                t = -5.1500      t = 9.3544
##                p = 0.000001      p = 0.0000
## MSP           -2.5697***      -3.2763***
##                (-3.8006, -1.3389) (-4.5232, -2.0294)
##                t = -4.0919      t = -5.1500
##                p = 0.00005      p = 0.000001
## CDPHE Mean     0.1091         0.1091
##                (-1.2668, 1.4849) (-1.2668, 1.4849)
##                t = 0.1554      t = 0.1554
##                p = 0.8766      p = 0.8766
## Fall          1.2246         1.2246
##                (-0.7069, 3.1562) (-0.7069, 3.1562)
##                t = 1.2426      t = 1.2426
##                p = 0.2141      p = 0.2141
## Winter        -3.3871***      -3.3871***
##                (-4.8577, -1.9165) (-4.8577, -1.9165)
##                t = -4.5142      t = -4.5142
##
```

```

##              p = 0.00001      p = 0.00001
## Spring      -0.0214          1.0482
##             (-1.9693, 1.9265) (-0.8606, 2.9570)
##             t = -0.0215      t = 1.0763
##             p = 0.9829      p = 0.2818
## Wearable*Fall -1.0269        -0.0214
##             (-2.9394, 0.8857) (-1.9693, 1.9265)
##             t = -1.0523      t = -0.0215
##             p = 0.2927      p = 0.9829
## MSP*Fall     0.8492          -5.1762***
##             (-1.6477, 3.3460) (-7.6385, -2.7139)
##             t = 0.6666      t = -4.1202
##             p = 0.5051      p = 0.00004
## CDPHE*Fall   4.3270***       0.8492
##             (1.8382, 6.8158) (-1.6477, 3.3460)
##             t = 3.4076      t = 0.6666
##             p = 0.0007      p = 0.5051
## Wearable*Winter 0.3695       0.4332
##             (-1.6349, 2.3739) (-1.5379, 2.4044)
##             t = 0.3613      t = 0.4308
##             p = 0.7179      p = 0.6667
## MSP*Winter   -0.8027        0.3695
##             (-2.7776, 1.1721) (-1.6349, 2.3739)
##             t = -0.7967      t = 0.3613
##             p = 0.4257      p = 0.7179
## CDPHE*Winter 14.1318***     14.1318***
##             (12.2828, 15.9808) (12.2828, 15.9808)
##             t = 14.9800      t = 14.9800
##             p = 0.0000      p = 0.0000
## -----
## Observations      345          345
## Log Likelihood    -1,220.5840    -1,220.5840
## Akaike Inf. Crit.  2,469.1690    2,469.1690
## Bayesian Inf. Crit. 2,522.9780    2,522.9780
## =====
## Note:              *p<0.05; **p<0.01; ***p<0.001

```

## OZONE MEASUREMENT DIFFERENCES



```
##
## Ozone Monitor Differences
## =====
##                               Dependent variable:
##                               -----
##                               O3
## -----
## CDPHE Mean                    -10.5836***
##                               (-12.2443, -8.9229)
##                               t = -12.4910
##                               p = 0.0000
## Wearable                      16.0944***
##                               (14.4337, 17.7550)
##                               t = 18.9949
##                               p = 0.0000
## -----
## Observations                  152
## R2                            0.5098
## Adjusted R2                   0.5066
## Residual Std. Error          8.7061 (df = 150)
## F Statistic                  156.0254*** (df = 1; 150) (p = 0.0000)
## =====
## Note:                        *p<0.05; **p<0.01; ***p<0.001
```