

viability_selection_mr_lifespan

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Libraries

```
library(dplyr) #data parsing library
```

```
##
```

```
## Attaching package: 'dplyr'
```

```

## The following objects are masked from 'package:stats':
##
##   filter, lag
## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union
library(data.table) #data parsing library mainly for reading and writing out

##
## Attaching package: 'data.table'
## The following objects are masked from 'package:dplyr':
##
##   between, first, last
library(ggplot2) #data visualization library
library(lubridate) #package to deal with time variables

##
## Attaching package: 'lubridate'
## The following objects are masked from 'package:data.table':
##
##   hour, isoweek, mday, minute, month, quarter, second, wday,
##   week, yday, year
## The following object is masked from 'package:base':
##
##   date
library(curl) #incorporate URLs
library(DiagrammeR) #figure aesthetics
library(MASS) #Functions and datasets to support

##
## Attaching package: 'MASS'
## The following object is masked from 'package:dplyr':
##
##   select
library(BBmisc) #Miscellaneous functions

##
## Attaching package: 'BBmisc'
## The following objects are masked from 'package:dplyr':
##
##   coalesce, collapse
## The following object is masked from 'package:base':
##
##   isFALSE
library(survival) #survival data analysis
library(survminer) #survival data analysis

## Loading required package: ggpubr
## Loading required package: magrittr

```

```
library(gridExtra) #figure aesthetics
```

```
##
## Attaching package: 'gridExtra'
## The following object is masked from 'package:dplyr':
##
##      combine
```

Download data and do preliminary parsing

Data are publicly available via github

```
data<- fread("https://raw.githubusercontent.com/adnguyen/Circadian_rhythm_runs_seasonal_timing/master/D
#look at summary of data
#glimpse(data)
```

Calculating metabolic rates

Metabolic Rate = CO2 production/ hours Mass Specific Metabolic Rate = CO2 production/hrs per mass

Problem with purge_time_1: only start and stop times except for apple cohort 1 **Solution** create time sequence for each cohort from start to end and merge into master data sheet

Day 11 Cohort

Creating the time sequence

```
#glimpse(data$purge_time_1)
#hm(data$purge_time_1)
data$day10purge <- lubridate::hour(hm(data$purge_time_1))+lubridate::minute(hm(data$purge_time_1))/60

## Warning in .parse_hms(..., order = "HM", quiet = quiet): Some strings
## failed to parse, or all strings are NAs

## Warning in .parse_hms(..., order = "HM", quiet = quiet): Some strings
## failed to parse, or all strings are NAs

#Obtaining start and end (min and max) of purges and sample size for each host, cohort day, and tape
param <- data%>%
  group_by(cohort_day, tape)%>%
  summarise(max=max(day10purge, na.rm=TRUE), min=min(day10purge, na.rm=TRUE), n=length(cohort_day))

#goal: for this section, we want a sequence of times for day 10 purge
data2 <- data%>%
  group_by(cohort_day, tape)%>%
  mutate(.,day10purge.trans=seq(from = min(day10purge, na.rm=TRUE), to = max(day10purge, na.rm=TRUE), 1
#glimpse(data2)

#cohort 2&3 for apple have the right sequence of purge times so they(day10purge) need to be replaced in
```

```

#extract number of rows we want to replace
data2[1:500,]%>%
  filter(cohort_day < 4)

## # A tibble: 211 x 46
## # Groups:   cohort_day, tape [4]
##   Ind_ID tape Site_name mass_day10 purge_time_1 purge1 collection_date
##   <int> <chr> <chr>          <dbl> <chr>          <dbl> <chr>
## 1     1  blue OG              6.94 13:38            NA 2017-08-21
## 2     2  blue Ferris          11.2 13:39            NA 2017-08-21
## 3     3  blue Ferris           6.72 13:39            NA 2017-08-21
## 4     4  blue Ferris          10.7 13:40            NA 2017-08-21
## 5     5  blue OG              3.85 13:41            NA 2017-08-21
## 6     6  blue OG              7.58 13:37            NA 2017-08-21
## 7     7  blue OG              6.41 13:35            NA 2017-08-21
## 8     8  blue OG              9.36 13:57            NA 2017-08-21
## 9     9  blue Ferris          7.98 13:41            NA 2017-08-21
## 10    10  blue OG              4.89 13:36            NA 2017-08-21
## # ... with 201 more rows, and 39 more variables: day10 <chr>,
## #   cohort_date <chr>, cohort_day <int>, Host <chr>, `Li-cor_1` <int>,
## #   resp_time_1 <chr>, resp_day11 <dbl>, mass_day14 <dbl>,
## #   purge_time_2 <chr>, resp_time_2 <chr>, resp_day15 <dbl>,
## #   Li_cor2 <int>, treatment_day15 <chr>, exit_fridge_date <chr>,
## #   Eclosion_reference_date <chr>, notes <chr>, Resp_code <int>,
## #   treatment <chr>, uniqueID <chr>, eclosion_date <chr>,
## #   eclosion_days <int>, well_id <chr>, organism <chr>,
## #   Trikinetics_position <int>, Trikinetic_monitor <int>,
## #   Trikinetics_entry_LD_time <chr>, Trikinetic_exit_date <chr>,
## #   Trikinetics_exit_LD_time <chr>, notes_2 <chr>,
## #   Free_run_trik_monitor <int>, Free_run_trik_position <int>,
## #   Free_run_entry_date <chr>, Free_run_entry_time <chr>,
## #   Free_run_exit_date <chr>, Free_run_exit_time <chr>, notes_3 <chr>,
## #   Adult_death_date <chr>, day10purge <dbl>, day10purge.trans <dbl>

data2[1:211,46]<- data2[1:211,45]
#data2[1:211,46]

```

Calculating start and end time for total amount of hours of CO2 production

```

#glimpse(data$resp_time_1)
#hms(data$resp_time_1)
data2$day10resp <- lubridate::hour(hms(data$resp_time_1))+lubridate::minute(hms(data$resp_time_1))/60

## Warning in .parse_hms(..., order = "HMS", quiet = quiet): Some strings
## failed to parse, or all strings are NAs

## Warning in .parse_hms(..., order = "HMS", quiet = quiet): Some strings
## failed to parse, or all strings are NAs

#reformatting to obtain positive values
data2$total_time_day10 <- (24 - data2$day10purge.trans) + data2$day10resp

```

Metabolic Rate Calculation

```
#getting denominator for mass-specific mr
data2$total_time_day10 * data$mass_day10
```

```
##      [1]      NA 229.83250 138.29942 220.81140  79.33293 156.94947
##      [7] 133.28352 191.35817 165.41053 102.06075 134.06937 198.97133
##     [13] 202.73053 174.27900 131.12400 153.31475 148.76400 137.45385
##     [19] 129.41402 124.37533 247.19317 137.31037 196.51778 186.87757
##     [25] 129.05108 191.59152 205.04187 159.63547 130.04567 174.42242
##     [31] 137.74995 171.24340 218.04005 204.34300 102.53520 167.59402
##     [37] 232.00425 206.99625 179.07500 104.12440 141.08700 147.13358
##     [43] 164.42125      NA 239.80840 207.43060 114.60095 168.56820
##     [49]      NA      NA 138.67740 196.90170 314.60017 142.91760
##     [55] 125.05880 113.25457 153.11313 140.15320 157.46640 229.13875
##     [61] 185.82117 133.44450 197.32320 145.49050 215.34893 196.72853
##     [67]  91.35700 165.46533 184.50458 151.48130 178.31283 209.33055
##     [73] 260.69035 164.04410 167.20608 179.64780 110.27360 261.05570
##     [79] 194.20050 214.30920 172.98820 225.70182 220.81920 176.24658
##     [85] 177.44417 146.92125 168.89110 229.35780      NA 186.92440
##     [91] 170.11611      NA 128.57202  72.88260 129.32833 128.61440
##     [97]  89.01920 159.33793  86.82412 112.69375 165.06190 115.35067
##    [103] 143.17950 115.47800 141.54450 117.52545 117.51428 114.32415
##    [109] 156.78033 118.28022 111.65880 107.33050 124.35120 142.86957
##    [115] 115.84625 150.19747 106.20518 117.50745 136.12157 102.05720
##    [121] 136.71595 123.43413 150.84300 120.24693 192.83600 148.08917
##    [127] 114.25620 155.41110 108.33290 122.71523 131.52720 124.16267
##    [133] 147.12793 103.73120 122.45975 189.97560 109.16353 185.47000
##    [139] 121.58408 114.76355 125.13542  85.46352 133.69167 176.58013
##    [145] 138.70560 127.59480 151.10165 189.68912 132.43253 121.99575
##    [151]      NA      NA 193.45755 125.64170 110.72000 104.15253
##    [157]  78.77340 156.23440 129.20732  91.36055 117.78230  83.02500
##    [163]  82.20875 165.95030 108.15933 147.41160 169.86240  78.14333
##    [169] 135.53155 105.65013 143.00550 146.07200 101.76903 163.52660
##    [175] 162.97340 113.43645  83.53340 103.07733 119.24325  92.65200
##    [181] 170.67317 116.43350 144.11613  99.21625 169.68645 122.85743
##    [187] 111.16300 143.25410  97.37500 146.13020 122.15800 112.01815
##    [193] 135.50570 116.38575 177.36157 110.93057 104.31200 118.35383
##    [199] 191.13900 147.22565 118.82573 140.15750 105.29482 137.67000
##    [205] 177.13395 157.43030 149.01320 112.64150 172.85580      NA
##    [211]      NA 160.98160 137.55440 184.14926 174.98031 191.56902
##    [217] 130.26728 221.11751 147.21585 185.66552  61.81770 183.90629
##    [223] 118.60771 145.96979 145.70552 128.39497 234.65451 130.91413
##    [229] 171.66903 172.36607 146.29393 227.10513 160.57917 125.69158
##    [235] 154.60008 181.70051 197.00156 184.72090 124.79015 191.41764
##    [241] 148.67394 100.12418 151.71858 162.62494  99.69210 182.48331
##    [247] 147.77721 131.76176 223.66187 220.85404 162.32503 164.41431
##    [253] 138.83681 144.21126 167.93866 117.94359 137.90120 208.16427
##    [259] 117.60693 142.84450 173.28354 125.29670 197.13775 249.00363
##    [265]  87.93122 129.72116 181.50986 221.52903 177.11113 199.71484
##    [271] 143.59812 137.42575 130.97763 100.38239 101.96694 194.29634
##    [277] 238.28225 110.63830 189.79042  83.82351 114.96796 182.28144
##    [283] 129.95645 151.67436 154.74247 186.75853 235.60695      NA
##    [289]      NA 104.58910 114.40759 174.24962 149.94311 166.28462
```

| | | | | | | | |
|----|-------|-----------|-----------|-----------|-----------|-----------|-----------|
| ## | [295] | 104.87402 | 151.05469 | 108.79407 | 171.44309 | 218.38588 | 119.94811 |
| ## | [301] | 158.76229 | 129.99138 | 180.47169 | 177.65541 | 104.29854 | 84.95010 |
| ## | [307] | 176.21938 | 153.05964 | 137.85460 | 177.45286 | 116.44867 | 118.43806 |
| ## | [313] | 134.50245 | 104.22427 | 152.81204 | 104.58031 | 188.10263 | 97.42360 |
| ## | [319] | 124.80878 | 104.22246 | 179.31408 | 92.35752 | 103.95750 | 226.67892 |
| ## | [325] | 143.04624 | 124.33558 | 220.15066 | 156.17497 | 185.22211 | 173.20529 |
| ## | [331] | 68.47136 | 130.34693 | 167.66992 | 179.35920 | 122.53376 | 169.45778 |
| ## | [337] | 156.86031 | 149.63238 | 223.31825 | 211.39354 | 120.14077 | 85.23373 |
| ## | [343] | 101.25868 | 206.39688 | 139.01981 | 131.60096 | 94.97724 | 90.01493 |
| ## | [349] | 148.91532 | 130.76857 | 169.55661 | 114.06377 | 125.39011 | 202.57852 |
| ## | [355] | 139.68017 | 115.63470 | 125.80445 | 162.43486 | 203.74528 | 156.70339 |
| ## | [361] | 122.19424 | 222.66181 | 147.45711 | 133.81049 | 125.63435 | NA |
| ## | [367] | NA | 114.17120 | 152.02211 | 104.25936 | 180.07640 | 132.07403 |
| ## | [373] | 145.75529 | 105.82967 | 180.42481 | 110.82069 | 139.22719 | 169.03201 |
| ## | [379] | 115.78122 | 102.40206 | 106.62088 | 116.57799 | 149.53794 | 169.14195 |
| ## | [385] | 141.70102 | 94.36315 | 111.54938 | 119.39799 | 164.12316 | 146.61286 |
| ## | [391] | 155.96672 | 102.18130 | 133.86163 | 144.72199 | 113.36554 | 101.47485 |
| ## | [397] | 118.34672 | 150.56059 | 101.83648 | 185.19999 | 122.23608 | 112.50406 |
| ## | [403] | 129.96509 | 148.46465 | 136.77738 | 143.86682 | 145.42596 | 141.68891 |
| ## | [409] | 115.95627 | 147.09597 | 121.32930 | 169.93489 | 147.72194 | 190.94451 |
| ## | [415] | 118.29552 | 143.88231 | 198.47243 | 160.72151 | 161.24418 | 135.63606 |
| ## | [421] | 221.81741 | 119.11098 | 114.86218 | 175.14436 | 99.21399 | 135.86251 |
| ## | [427] | 186.36390 | 109.93696 | 162.30509 | 117.68008 | 166.70285 | 182.91869 |
| ## | [433] | 164.85652 | 130.94903 | 131.43494 | 149.60601 | 127.62350 | 172.15866 |
| ## | [439] | 113.35145 | 179.60115 | NA | NA | 129.04167 | 94.89798 |
| ## | [445] | 135.26804 | 127.54211 | 141.51345 | 90.80238 | 173.95637 | 109.27991 |
| ## | [451] | 153.63981 | 123.73701 | 96.88159 | 111.12977 | 108.14407 | 110.95517 |
| ## | [457] | 125.67430 | 149.81600 | 102.16825 | 71.88306 | 126.76443 | 138.24708 |
| ## | [463] | 142.71816 | 213.94819 | 124.31272 | 154.43764 | 111.51000 | 162.63898 |
| ## | [469] | 140.57957 | 148.70665 | 113.75937 | 161.47634 | 102.55418 | 96.53827 |
| ## | [475] | 161.16447 | 105.12363 | 159.67721 | 112.51355 | 128.71857 | 173.07352 |
| ## | [481] | 143.56650 | 150.14673 | 147.38431 | 164.34063 | 143.20026 | 96.61558 |
| ## | [487] | 176.89015 | 100.25656 | 175.56435 | 97.15781 | 138.86573 | 217.30523 |
| ## | [493] | 119.30236 | 166.43456 | 131.89045 | 167.86442 | 175.93598 | 76.46047 |
| ## | [499] | 105.75255 | 162.04308 | 95.74179 | 115.32514 | 108.68151 | 163.93181 |
| ## | [505] | 97.50559 | 115.76844 | 218.95465 | 86.14281 | 189.65865 | 168.06256 |
| ## | [511] | 144.51311 | 115.73658 | 103.61306 | 107.44191 | 120.21442 | 66.77178 |
| ## | [517] | 112.55024 | NA | NA | 89.60792 | 135.99274 | 92.13925 |
| ## | [523] | 83.90501 | 157.98976 | 143.46973 | 171.59371 | 128.85333 | 103.59508 |
| ## | [529] | 89.02023 | 115.16380 | 84.68511 | 121.06293 | 101.38043 | 125.89552 |
| ## | [535] | 149.35228 | 173.32581 | 107.94427 | 126.61853 | 91.03290 | 170.07970 |
| ## | [541] | 106.17546 | 97.83041 | 76.80043 | 79.13677 | 76.23361 | 114.29979 |
| ## | [547] | 87.89238 | 115.97241 | 200.81198 | 68.92276 | 184.30115 | 152.07273 |
| ## | [553] | 122.83428 | 76.83550 | 54.82412 | 137.34469 | 158.10355 | 132.63194 |
| ## | [559] | 113.30461 | 103.90424 | 131.39523 | 147.56714 | 122.57605 | 88.43900 |
| ## | [565] | 186.60155 | 118.16147 | 110.56058 | 107.71312 | 154.63522 | 167.39581 |
| ## | [571] | 90.92231 | 120.61385 | 97.77491 | 133.09700 | 134.84764 | 103.58501 |
| ## | [577] | 210.43972 | 71.55708 | 138.55393 | 73.76099 | 115.49332 | 122.25481 |
| ## | [583] | 156.69138 | 147.05874 | 71.82052 | 155.03377 | 131.96618 | 86.10923 |
| ## | [589] | 190.51437 | 99.85836 | 111.52131 | 130.28281 | 201.18865 | 143.14706 |
| ## | [595] | 119.45067 | NA | NA | 119.63500 | 115.70564 | 109.97769 |
| ## | [601] | 84.11150 | 101.91854 | 105.80700 | 163.05083 | 132.14667 | 102.83111 |
| ## | [607] | 112.81900 | 116.16443 | 69.97429 | 108.60400 | 77.91131 | 141.63023 |
| ## | [613] | 153.76575 | 142.90833 | 158.54578 | 113.14422 | 129.22849 | 170.06278 |

| | | | | | | | |
|----|-------|-----------|-----------|-----------|-----------|-----------|-----------|
| ## | [619] | 103.38693 | 116.48976 | 120.22496 | 146.42750 | 98.04133 | 70.30732 |
| ## | [625] | 118.30453 | 101.47312 | 100.05735 | 166.76587 | 182.14980 | 183.72667 |
| ## | [631] | 102.60067 | 89.87078 | 125.54669 | 130.33670 | 107.27396 | 82.26349 |
| ## | [637] | 130.84605 | 131.12493 | 105.87344 | 104.90987 | 154.93172 | 89.75972 |
| ## | [643] | 54.86287 | 99.81554 | 83.38311 | 163.82750 | 168.12778 | 101.39176 |
| ## | [649] | 178.41390 | 109.99351 | 134.92269 | 89.47950 | 100.63481 | 95.05394 |
| ## | [655] | 83.93600 | 94.05100 | 122.40880 | 119.65520 | 95.13611 | 114.65067 |
| ## | [661] | 188.32515 | 112.86920 | 91.53478 | 100.66680 | 177.23608 | 89.49252 |
| ## | [667] | 139.70125 | 116.36451 | 103.77651 | 88.98170 | 112.40042 | 151.29811 |
| ## | [673] | 150.25650 | 88.40724 | NA | NA | 185.94333 | 139.95375 |
| ## | [679] | 131.52484 | 122.28105 | 147.64440 | 171.40194 | 167.60189 | 194.89025 |
| ## | [685] | 181.64644 | 222.23556 | 115.13557 | 86.41585 | 164.32688 | 97.10026 |
| ## | [691] | 91.82586 | 155.65060 | 93.46918 | 228.55036 | 173.21986 | 114.20893 |
| ## | [697] | 189.69403 | 138.66261 | 213.11254 | 94.14771 | 138.83520 | 230.59548 |
| ## | [703] | 152.68302 | 163.90062 | 155.20642 | 188.87217 | 120.67228 | NA |
| ## | [709] | NA | 194.23015 | 255.76490 | 150.58549 | 197.82436 | 202.23543 |
| ## | [715] | 138.55750 | 205.08268 | 260.08189 | 146.71509 | 169.98730 | 149.46466 |
| ## | [721] | 162.13095 | 206.64901 | 89.99596 | 153.69833 | 134.64714 | 126.56377 |
| ## | [727] | 178.32049 | 221.48969 | 153.32440 | 149.49620 | 139.70537 | 187.19701 |
| ## | [733] | 194.29798 | 162.31713 | 155.51597 | 159.06149 | 117.90465 | 298.18559 |
| ## | [739] | 156.94160 | 81.61460 | 124.55479 | 154.51687 | 203.50478 | 107.76720 |
| ## | [745] | 144.78826 | 239.55233 | 123.30920 | 105.00450 | 205.05270 | 198.74050 |
| ## | [751] | 245.93819 | 137.16900 | 100.04887 | 185.18833 | 158.71118 | 194.40355 |
| ## | [757] | 122.42304 | 131.01622 | 195.41531 | 170.31976 | 103.96842 | 195.76287 |
| ## | [763] | 72.91833 | 122.18592 | 149.02107 | 110.17400 | 149.69059 | 84.25531 |
| ## | [769] | 128.07301 | 97.24834 | 116.89326 | 133.64101 | 207.95864 | 198.31675 |
| ## | [775] | 134.35809 | 179.77615 | 158.39652 | 144.35674 | 146.46266 | 208.63462 |
| ## | [781] | 136.01120 | 125.82232 | 243.24489 | 133.93615 | 183.81389 | 155.97689 |
| ## | [787] | 146.62580 | 147.93977 | 155.32271 | 106.80806 | 125.03218 | 139.03290 |
| ## | [793] | 134.70249 | 112.61489 | 198.54930 | 192.39418 | 91.57236 | 151.66250 |
| ## | [799] | 135.29763 | 204.36869 | 131.55925 | 167.30835 | 133.11269 | 133.93418 |
| ## | [805] | 183.91530 | 159.39016 | 245.42675 | 125.04043 | 110.31178 | 64.38848 |
| ## | [811] | 166.83944 | 98.92447 | 113.26006 | 159.85266 | 141.63854 | NA |
| ## | [817] | NA | 242.00267 | 187.95620 | 184.30204 | 90.76780 | 239.19784 |
| ## | [823] | 151.60813 | 155.08402 | 160.51899 | 181.45933 | 125.11891 | 197.46458 |
| ## | [829] | 122.67061 | 182.10020 | 154.05487 | 159.32617 | 212.98948 | 205.17475 |
| ## | [835] | 201.18759 | 116.55287 | 103.23271 | 181.79849 | 74.35889 | 116.77060 |
| ## | [841] | 133.41942 | 209.18082 | 121.80952 | 197.57476 | 235.08032 | 179.93054 |
| ## | [847] | 216.40576 | 205.43619 | 112.58999 | 107.09379 | 90.20766 | 83.85344 |
| ## | [853] | 174.02456 | 164.27322 | 184.31481 | 114.06429 | 150.08956 | 144.48600 |
| ## | [859] | 116.34581 | 117.47332 | 183.57523 | 175.96135 | 174.25034 | 174.71819 |
| ## | [865] | 128.04229 | 184.17562 | 105.07557 | 149.20299 | 74.10898 | 180.54654 |
| ## | [871] | 144.42138 | 108.81461 | 122.08972 | 154.36449 | 143.73694 | 120.96534 |
| ## | [877] | 140.39208 | 151.42382 | 122.01917 | 84.88125 | 194.25631 | 197.88847 |
| ## | [883] | 85.62919 | 110.78646 | 145.61374 | 128.43631 | 157.28490 | 106.14017 |
| ## | [889] | 168.42726 | 164.13566 | 165.71969 | 114.48578 | 112.56015 | NA |
| ## | [895] | NA | 188.87733 | 142.77008 | 100.51128 | 156.74864 | 176.45575 |
| ## | [901] | 170.19481 | 139.29622 | 107.68737 | 55.87789 | 131.84202 | 74.13450 |
| ## | [907] | 122.45263 | 106.23768 | 92.04438 | 142.71279 | 118.74113 | 68.81785 |
| ## | [913] | 102.82762 | 136.37363 | 122.26594 | 104.88081 | 116.65928 | 58.99633 |
| ## | [919] | 133.89875 | 111.78581 | 135.08548 | 197.05147 | 135.45937 | 111.61904 |
| ## | [925] | 175.19706 | 92.00232 | 129.38189 | 207.19922 | 85.93670 | 110.57158 |
| ## | [931] | 97.82292 | 132.69178 | 91.53976 | 187.05555 | 97.78618 | 234.27203 |
| ## | [937] | 138.92104 | 124.97691 | 92.23114 | 154.30114 | 140.15043 | 165.45698 |

| | | | | | | | |
|----|--------|-----------|-----------|-----------|-----------|-----------|-----------|
| ## | [943] | 134.03450 | 120.69978 | 155.56993 | 167.73295 | 158.65154 | 185.67999 |
| ## | [949] | 134.32867 | 223.33826 | 127.07675 | 100.76959 | 127.16367 | 199.24956 |
| ## | [955] | 174.79408 | 168.92368 | 138.27653 | 137.85677 | 137.01765 | 125.33659 |
| ## | [961] | 132.06652 | 216.36658 | 163.19895 | 130.90068 | 126.21964 | 111.30702 |
| ## | [967] | 125.10399 | 164.65189 | 143.44467 | 161.58900 | NA | NA |
| ## | [973] | 161.40020 | 174.14401 | 147.90829 | 158.13235 | 205.00830 | 125.58432 |
| ## | [979] | 156.68621 | 118.60232 | 194.05737 | 175.16424 | 74.65430 | 125.86896 |
| ## | [985] | 103.24228 | 135.42213 | 107.47118 | 103.72129 | 177.16409 | 117.41503 |
| ## | [991] | 164.38699 | 144.11543 | 138.89903 | 181.89089 | 86.55914 | 183.17073 |
| ## | [997] | 229.89745 | 224.88473 | 201.08884 | 245.96768 | 197.49885 | 123.18913 |
| ## | [1003] | 166.93640 | 140.75071 | 206.67424 | 123.09466 | 115.00619 | 236.09784 |
| ## | [1009] | 190.81831 | 186.80620 | 183.37529 | 178.06876 | 123.25473 | 121.81579 |
| ## | [1015] | 113.14434 | 133.19119 | 192.72084 | 206.19750 | 169.09882 | 142.82576 |
| ## | [1021] | 189.17990 | 124.32965 | 112.68025 | 131.85116 | 69.82491 | 99.12279 |
| ## | [1027] | 134.56222 | 103.80503 | 90.67460 | 164.44986 | 90.41556 | 120.52433 |
| ## | [1033] | 202.70751 | 195.15780 | 107.35823 | 125.17495 | 145.96197 | 100.66644 |
| ## | [1039] | 195.18495 | NA | 125.96601 | 190.89313 | NA | 135.69220 |
| ## | [1045] | 161.44438 | NA | 141.29947 | 107.26013 | 179.81771 | 173.02445 |
| ## | [1051] | 139.08702 | 212.86140 | 149.18889 | 107.26954 | 128.97442 | 136.60285 |
| ## | [1057] | 174.45017 | 129.57646 | 102.74794 | 80.74769 | 128.87995 | 116.41035 |
| ## | [1063] | 170.49681 | 82.17580 | 190.45479 | 114.25144 | 131.26973 | 86.99943 |
| ## | [1069] | 140.32478 | 122.85424 | 64.62802 | 81.70921 | 157.14502 | 139.07834 |
| ## | [1075] | 211.88930 | 209.82638 | 173.82627 | 233.53173 | 202.16626 | 136.24919 |
| ## | [1081] | 139.99496 | 171.13063 | 92.07165 | 194.57546 | 140.88791 | 189.04264 |
| ## | [1087] | 94.90849 | 141.45570 | 176.52654 | 144.25555 | 139.56583 | 184.21939 |
| ## | [1093] | 182.51914 | 123.45368 | 108.81734 | 160.89824 | 168.30982 | 154.25158 |
| ## | [1099] | 172.50130 | 158.11884 | 122.98819 | 136.85618 | 86.48502 | 124.15686 |
| ## | [1105] | 153.19206 | 113.89664 | NA | 118.55089 | 146.11931 | NA |
| ## | [1111] | 165.01671 | 93.25969 | 192.78520 | 157.71832 | 192.02191 | 130.31452 |
| ## | [1117] | 142.52142 | 176.44757 | 192.40043 | 193.35538 | 106.51624 | NA |
| ## | [1123] | 209.23560 | 97.79022 | 94.10250 | 156.27903 | 167.25972 | 175.34618 |
| ## | [1129] | 178.20969 | 172.72205 | 153.63219 | 109.02387 | 246.01945 | 177.27207 |
| ## | [1135] | 156.58170 | 101.34316 | 103.81859 | 84.09385 | 108.11119 | 140.71988 |
| ## | [1141] | 139.48748 | 127.58015 | 121.92127 | 191.92305 | 69.28473 | 100.80842 |
| ## | [1147] | 110.33112 | 135.36432 | 144.10181 | NA | 131.96697 | 190.06135 |
| ## | [1153] | 137.17677 | 155.07845 | 62.36154 | 125.14438 | 155.71976 | 127.54945 |
| ## | [1159] | 105.14225 | 190.23968 | 119.31389 | 95.89143 | 103.61272 | 71.39818 |
| ## | [1165] | 185.35439 | 115.18859 | 113.22828 | 148.26812 | 114.90293 | 178.42950 |
| ## | [1171] | 106.15328 | 172.19544 | 183.32464 | 133.52552 | 115.99098 | 71.29650 |
| ## | [1177] | 81.24250 | 88.50603 | 143.59163 | 193.10940 | 94.25452 | 74.10967 |
| ## | [1183] | 170.72610 | 118.84786 | 142.82599 | 131.15406 | 96.60050 | 161.63473 |
| ## | [1189] | NA | NA | 125.24100 | 222.38906 | 102.70373 | 129.58987 |
| ## | [1195] | 125.17206 | 135.80160 | 166.46853 | 107.84126 | 115.13439 | 199.33492 |
| ## | [1201] | 109.69509 | 142.04792 | 79.38540 | 153.21543 | 204.46090 | 112.86950 |
| ## | [1207] | 160.72296 | 120.80817 | 185.95697 | 119.01319 | 137.01073 | 95.59241 |
| ## | [1213] | 163.11721 | 112.68315 | 145.52526 | NA | 128.37227 | 102.99358 |
| ## | [1219] | 130.54333 | 124.46763 | 146.46670 | 169.47449 | 174.35393 | 209.78459 |
| ## | [1225] | 115.45929 | 112.96324 | 178.08848 | 138.19074 | 208.48416 | 129.94223 |
| ## | [1231] | 76.56976 | 99.80877 | 164.27818 | 125.19515 | 165.56117 | 155.14622 |
| ## | [1237] | 119.98727 | 139.62845 | 130.90421 | 78.71567 | 154.87000 | 120.99552 |
| ## | [1243] | 172.24562 | 119.87583 | 185.34973 | 149.28639 | 186.82090 | 132.24814 |
| ## | [1249] | NA | 120.39153 | 62.70365 | 151.27550 | 181.48187 | 257.29440 |
| ## | [1255] | 230.48920 | 158.78582 | 165.71771 | 140.06250 | 110.67232 | 159.46962 |
| ## | [1261] | 129.03784 | 183.07514 | 58.86185 | 195.94564 | 181.33248 | 158.52721 |

| | | | | | | | |
|----|--------|-----------|-----------|-----------|-----------|-----------|-----------|
| ## | [1267] | 166.06998 | 103.13712 | 130.44214 | 162.12783 | 184.43924 | 183.60570 |
| ## | [1273] | 177.54498 | 155.28125 | 217.88771 | 185.58130 | 157.22397 | 151.64100 |
| ## | [1279] | 245.84774 | 149.92279 | 146.18557 | 148.78407 | 120.69537 | 147.61535 |
| ## | [1285] | 103.04634 | 163.31440 | 228.86843 | 181.84956 | 94.65262 | 147.40825 |
| ## | [1291] | NA | NA | 167.41060 | 139.91920 | 177.49981 | 224.01475 |
| ## | [1297] | 193.41696 | 138.99619 | 171.99369 | 237.28411 | 226.08175 | 234.10107 |
| ## | [1303] | 224.96252 | 76.53008 | 148.76576 | 129.32995 | 207.15345 | 179.11304 |
| ## | [1309] | 137.60457 | 121.85052 | 169.57946 | 160.39995 | 163.95529 | 183.09422 |
| ## | [1315] | 202.27354 | 152.36843 | 152.40566 | 136.38287 | 159.67222 | 129.36755 |
| ## | [1321] | 209.50420 | 125.34959 | 118.65316 | 128.99707 | 130.78535 | 156.07044 |
| ## | [1327] | 179.40675 | 132.51138 | NA | NA | 122.93398 | 176.52255 |
| ## | [1333] | 229.00253 | 255.54501 | 164.89954 | 166.59443 | 111.82722 | 229.30493 |
| ## | [1339] | 140.96983 | 264.11051 | 154.52414 | 201.45797 | 293.41517 | 109.00732 |
| ## | [1345] | 166.77159 | 144.01834 | 130.03841 | 161.62988 | 88.87576 | 199.54329 |
| ## | [1351] | 78.46794 | NA | 166.55057 | 132.61905 | 188.86211 | 227.53687 |
| ## | [1357] | 157.37313 | 188.97673 | 289.72767 | 115.52040 | 106.97229 | 258.29933 |
| ## | [1363] | NA | NA | 181.21833 | 269.44909 | 76.11165 | 182.23958 |
| ## | [1369] | 295.94153 | 228.57671 | 257.07097 | 162.91631 | 114.13193 | 184.46888 |
| ## | [1375] | 272.15066 | NA | 228.21406 | 117.72246 | 201.43576 | NA |
| ## | [1381] | 205.56945 | 258.47118 | 132.71567 | 190.22469 | NA | 119.85120 |
| ## | [1387] | 168.58311 | 112.66252 | 283.91932 | 138.41089 | 180.62476 | 191.69532 |
| ## | [1393] | 133.85439 | 178.69032 | 173.26437 | NA | NA | NA |
| ## | [1399] | 105.33883 | 255.49981 | 226.82146 | 168.16674 | 126.75349 | 113.50602 |
| ## | [1405] | 115.39740 | 148.18314 | 137.37477 | 188.60564 | 155.37624 | 159.50164 |
| ## | [1411] | 187.71644 | 223.54463 | 199.39100 | 77.97282 | 152.39923 | 248.14182 |
| ## | [1417] | 140.61874 | 220.11266 | 112.58019 | 143.82033 | 282.24373 | 191.37258 |
| ## | [1423] | 229.81161 | 170.70195 | 145.12953 | 190.37722 | 205.71058 | 157.07180 |
| ## | [1429] | 114.61053 | 189.86335 | NA | NA | 117.04825 | 175.52945 |
| ## | [1435] | 183.29962 | 92.76235 | 142.52697 | 282.45317 | 150.12132 | 226.61818 |
| ## | [1441] | 192.51811 | 197.23011 | 139.58596 | 143.47117 | 63.31617 | 295.51770 |
| ## | [1447] | 101.15477 | 175.30460 | 135.78100 | 130.45870 | 144.88385 | 169.90607 |
| ## | [1453] | 177.26170 | 96.37336 | 260.60725 | 156.16782 | 217.39098 | 95.39339 |
| ## | [1459] | 134.69075 | 59.63612 | 164.27327 | 97.74488 | 140.16818 | 155.90723 |
| ## | [1465] | NA | NA | 57.03733 | 117.98780 | 109.04279 | 96.53875 |
| ## | [1471] | 115.12486 | 125.88637 | 117.70489 | 169.94315 | 143.36794 | 151.76850 |
| ## | [1477] | 172.40367 | 131.71182 | 93.46892 | 159.63934 | 138.42591 | 155.70029 |
| ## | [1483] | 115.16696 | NA | NA | 142.46373 | 144.03979 | 93.25622 |
| ## | [1489] | 171.49212 | 122.52556 | 162.22194 | 144.81884 | 174.23853 | 121.74619 |
| ## | [1495] | 129.73450 | 125.72552 | 152.32078 | 195.37416 | 138.77211 | 45.84812 |
| ## | [1501] | 103.58647 | 118.07816 | NA | NA | 171.06479 | 231.40373 |
| ## | [1507] | 133.34522 | 166.00527 | 184.10000 | 164.72538 | 168.18218 | 117.46247 |
| ## | [1513] | 230.91246 | 148.86109 | 241.87710 | 124.80000 | 126.04622 | 226.17288 |
| ## | [1519] | 141.98321 | 211.59838 | 162.37694 | 182.63900 | 106.95641 | 216.76707 |
| ## | [1525] | 228.16920 | 222.12979 | 109.03537 | 226.84534 | 155.39157 | 184.90230 |
| ## | [1531] | 124.49999 | 148.65269 | 136.98108 | 176.25465 | 199.14195 | 154.42257 |
| ## | [1537] | 214.56905 | NA | NA | 191.59569 | 150.94633 | 165.19561 |
| ## | [1543] | 113.30868 | 167.63394 | 87.97359 | 179.01035 | 219.64297 | 183.29902 |
| ## | [1549] | 147.69612 | 170.90734 | 192.00438 | 114.68039 | 169.89503 | 173.17221 |
| ## | [1555] | 160.37919 | 215.30632 | 148.62550 | 202.97417 | 225.21471 | 124.47839 |
| ## | [1561] | 263.70393 | 214.53195 | 168.83965 | 111.12405 | 105.38232 | 167.77002 |
| ## | [1567] | 144.43641 | 176.89717 | 154.02407 | 153.83270 | 179.99599 | 171.70137 |
| ## | [1573] | NA | NA | 116.94893 | 110.09975 | 99.31155 | 130.78391 |
| ## | [1579] | 168.97578 | 155.24691 | 217.07984 | 206.01769 | 96.02860 | 94.98532 |
| ## | [1585] | 122.37158 | 171.32545 | 175.56504 | 121.48034 | 147.04657 | 157.14622 |

| | | | | | | | |
|----|--------|-----------|-----------|-----------|-----------|-----------|-----------|
| ## | [1591] | 83.97080 | 174.13751 | 205.55791 | 123.29382 | 138.50174 | 168.35341 |
| ## | [1597] | 124.97149 | 139.02230 | 123.38955 | 141.90060 | NA | NA |
| ## | [1603] | 155.99566 | 106.87204 | 155.74629 | 147.40216 | 144.09796 | 214.62090 |
| ## | [1609] | 127.98428 | 136.21218 | 200.43442 | 126.27453 | 160.20536 | 197.56400 |
| ## | [1615] | 193.03993 | 127.27776 | 177.49962 | 200.55889 | 178.80624 | 135.99812 |
| ## | [1621] | 187.58995 | 101.66556 | 121.15564 | 200.30916 | 177.85682 | 222.77266 |
| ## | [1627] | 75.35852 | 118.80559 | NA | NA | 284.63086 | 248.41157 |
| ## | [1633] | 342.87828 | 268.42123 | 168.89510 | 167.94445 | NA | 135.95820 |
| ## | [1639] | 348.35425 | 287.32466 | 184.03375 | 271.58357 | 295.98007 | 239.65876 |
| ## | [1645] | 357.39165 | 257.55359 | 191.05995 | 285.32195 | 333.82929 | 166.41071 |
| ## | [1651] | 391.70711 | 173.07923 | 141.63764 | 242.47639 | NA | NA |
| ## | [1657] | 193.67465 | 393.91076 | 244.67326 | 160.90620 | 429.54313 | 258.88923 |
| ## | [1663] | NA | 261.02222 | 290.47172 | 215.23385 | 298.27315 | 339.98629 |
| ## | [1669] | 303.45588 | 421.63708 | 230.37945 | 237.98462 | 271.31641 | 344.85642 |
| ## | [1675] | 240.89971 | 372.17733 | 293.00589 | 396.55512 | 181.10943 | 252.37705 |
| ## | [1681] | 341.43129 | NA | NA | 208.67490 | 339.26514 | 233.13299 |
| ## | [1687] | 272.27025 | 298.62639 | 245.87831 | 244.55076 | 299.53460 | 166.48503 |
| ## | [1693] | 122.30472 | 126.88488 | 195.46997 | 297.62357 | 199.33570 | 246.35248 |
| ## | [1699] | 213.09656 | 341.29702 | 279.99078 | 96.63607 | 129.15913 | 236.71715 |
| ## | [1705] | 139.71983 | NA | NA | 300.60385 | 218.35745 | 231.03637 |
| ## | [1711] | 136.90436 | 207.58799 | 174.17073 | 153.68186 | 282.55217 | 223.63561 |
| ## | [1717] | 189.79175 | 196.04115 | 215.88723 | 161.86514 | 220.65230 | 250.48661 |
| ## | [1723] | 185.54992 | 309.50275 | 292.05523 | 221.33783 | 326.81631 | 249.02293 |
| ## | [1729] | 121.85137 | NA | NA | 300.49000 | 224.06535 | 282.25470 |
| ## | [1735] | 299.08908 | 257.06478 | 269.13952 | 205.40337 | 204.37415 | 167.69048 |
| ## | [1741] | 292.37558 | 212.54228 | 106.41820 | 196.93909 | 203.08274 | 181.90241 |
| ## | [1747] | 203.87173 | 279.37121 | 303.91372 | 134.95179 | 222.12850 | 106.63448 |
| ## | [1753] | 157.52470 | 152.53430 | 77.87794 | 258.42837 | 205.42185 | 256.04759 |
| ## | [1759] | 214.76853 | 173.26972 | 162.83829 | NA | NA | 268.25590 |
| ## | [1765] | 165.04748 | 194.73156 | 190.72314 | 186.96013 | 259.66761 | 188.49696 |
| ## | [1771] | 103.46579 | 98.40217 | 175.29615 | 206.41318 | 261.27665 | 242.12741 |
| ## | [1777] | 258.93460 | 196.56143 | 298.77908 | 228.20470 | 292.11188 | 182.28691 |
| ## | [1783] | 281.48766 | 224.67127 | 180.61264 | 132.63237 | 194.59127 | 289.26213 |
| ## | [1789] | 303.84891 | 220.00214 | 282.26321 | 319.69799 | 135.20901 | NA |
| ## | [1795] | NA | 273.47017 | 204.65839 | 182.75302 | 197.10565 | 97.16776 |
| ## | [1801] | 127.81085 | 206.63180 | 175.16403 | 359.52211 | 251.50889 | 246.87747 |
| ## | [1807] | 250.52297 | 225.46683 | 143.75743 | 229.28302 | 265.92420 | 241.07951 |
| ## | [1813] | NA | NA | 170.01225 | 192.49412 | 240.64383 | 271.07313 |
| ## | [1819] | 232.27686 | 203.80063 | 170.66534 | 190.24117 | 119.26710 | 182.65867 |
| ## | [1825] | 282.34700 | 108.89333 | 196.79987 | 192.14163 | 287.35014 | 228.00400 |
| ## | [1831] | NA | NA | 194.32812 | 192.30220 | 266.61094 | 106.91639 |
| ## | [1837] | 109.80647 | 255.94035 | 173.66084 | 285.16990 | 161.12841 | 129.67188 |
| ## | [1843] | 203.56829 | 114.60427 | 254.57737 | 215.57163 | 286.00971 | 176.85772 |
| ## | [1849] | 272.02362 | 184.39696 | NA | NA | 163.99443 | 157.42036 |
| ## | [1855] | 212.05384 | 114.49340 | 198.74667 | 230.93381 | 241.30453 | 186.10239 |
| ## | [1861] | 249.04186 | 222.06497 | 231.07758 | 198.79689 | 165.42985 | 245.85451 |
| ## | [1867] | 248.39027 | 196.69650 | 193.23200 | NA | NA | 190.80040 |
| ## | [1873] | 160.06129 | 157.68684 | 187.36272 | NA | NA | 222.82840 |
| ## | [1879] | 153.41239 | 225.50421 | 151.69722 | NA | NA | 258.86433 |
| ## | [1885] | 189.11456 | 163.90162 | 276.17524 | 277.53257 | 225.61704 | NA |
| ## | [1891] | NA | 272.08317 | 223.48276 | 163.59400 | 210.05217 | 178.80597 |
| ## | [1897] | NA | NA | 241.96603 | 239.51537 | 137.91522 | 199.50657 |
| ## | [1903] | NA | NA | 212.69424 | 196.33720 | 133.97971 | NA |
| ## | [1909] | NA | | | | | |

```
#Metabolic Rate for day 11
data2$MR11<- data$resp_day11/(data2$total_time_day10)

#Mass specific Metabolic Rate for day 11
data2$msMR11<- data$resp_day11/(data2$total_time_day10 * data$mass_day10)
```

Controlling for blank controls

```
#Assign the blanks to an object
data3 <- data2%>%
  group_by(cohort_day, tape)%>%
  filter(Site_name=="Blank")%>%
  summarise(mean.blank=mean(MR11,na.rm=TRUE))

#check columns
#glimpse(data3)
data3$mean.blank

## [1] 0.0011257856 0.0006226170 0.0003137692 0.0038187342 0.0006020381
## [6] 0.0005606454 0.0067232397 0.0010034398 0.0004597089 0.0003777458
## [11] 0.0004217324 0.0008299999 0.0010143259 0.0006680801 0.0012768032
## [16] 0.0005854209 0.0010247606 0.0006378185 0.0011384133 0.0005885973
## [21] 0.0004956928 0.0143301474 0.0009422219 0.0008234242 0.0010247213
## [26] 0.0009524855 0.0006771652 0.0006364143 0.0007423115 0.0004865312
## [31] 0.0051276472 0.0004456934 0.0003163145 0.0009844019 0.0006993465
## [36] 0.0004824640 0.0009931376 0.0002493671

#merge data3 and data2 by cohort day and tape and create a mean blanks column
data4 <- inner_join(data2, data3, by=c("cohort_day", "tape"))
data4$mean.blank

## [1] 0.0011257856 0.0011257856 0.0011257856 0.0011257856 0.0011257856
## [6] 0.0011257856 0.0011257856 0.0011257856 0.0011257856 0.0011257856
## [11] 0.0011257856 0.0011257856 0.0011257856 0.0011257856 0.0011257856
## [16] 0.0011257856 0.0011257856 0.0011257856 0.0011257856 0.0011257856
## [21] 0.0011257856 0.0011257856 0.0011257856 0.0011257856 0.0011257856
## [26] 0.0011257856 0.0011257856 0.0011257856 0.0011257856 0.0011257856
## [31] 0.0011257856 0.0011257856 0.0011257856 0.0011257856 0.0011257856
## [36] 0.0011257856 0.0011257856 0.0011257856 0.0011257856 0.0011257856
## [41] 0.0011257856 0.0011257856 0.0011257856 0.0011257856 0.0011257856
## [46] 0.0011257856 0.0011257856 0.0011257856 0.0011257856 0.0006226170
## [51] 0.0006226170 0.0006226170 0.0006226170 0.0006226170 0.0006226170
## [56] 0.0006226170 0.0006226170 0.0006226170 0.0006226170 0.0006226170
## [61] 0.0006226170 0.0006226170 0.0006226170 0.0006226170 0.0006226170
## [66] 0.0006226170 0.0006226170 0.0006226170 0.0006226170 0.0006226170
## [71] 0.0006226170 0.0006226170 0.0006226170 0.0006226170 0.0006226170
## [76] 0.0006226170 0.0006226170 0.0006226170 0.0006226170 0.0006226170
## [81] 0.0006226170 0.0006226170 0.0006226170 0.0006226170 0.0006226170
## [86] 0.0006226170 0.0006226170 0.0006226170 0.0006226170 0.0006226170
## [91] 0.0006226170 0.0006226170 0.0003137692 0.0003137692 0.0003137692
## [96] 0.0003137692 0.0003137692 0.0003137692 0.0003137692 0.0003137692
## [101] 0.0003137692 0.0003137692 0.0003137692 0.0003137692 0.0003137692
## [106] 0.0003137692 0.0003137692 0.0003137692 0.0003137692 0.0003137692
```

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

```
## [1731] 0.0004597089 0.0008299999 0.0008299999 0.0008299999 0.0008299999
## [1736] 0.0008299999 0.0008299999 0.0008299999 0.0008299999 0.0008299999
## [1741] 0.0008299999 0.0008299999 0.0008299999 0.0008299999 0.0008299999
## [1746] 0.0008299999 0.0008299999 0.0008299999 0.0008299999 0.0008299999
## [1751] 0.0008299999 0.0008299999 0.0008299999 0.0008299999 0.0008299999
## [1756] 0.0008299999 0.0008299999 0.0008299999 0.0008299999 0.0008299999
## [1761] 0.0008299999 0.0008299999 0.0008299999 0.0004217324 0.0004217324
## [1766] 0.0004217324 0.0004217324 0.0004217324 0.0004217324 0.0004217324
## [1771] 0.0004217324 0.0004217324 0.0004217324 0.0004217324 0.0004217324
## [1776] 0.0004217324 0.0004217324 0.0004217324 0.0004217324 0.0004217324
## [1781] 0.0004217324 0.0004217324 0.0004217324 0.0004217324 0.0004217324
## [1786] 0.0004217324 0.0004217324 0.0004217324 0.0004217324 0.0004217324
## [1791] 0.0004217324 0.0004217324 0.0004217324 0.0004217324 0.0004217324
## [1796] 0.0006680801 0.0006680801 0.0006680801 0.0006680801 0.0006680801
## [1801] 0.0006680801 0.0006680801 0.0006680801 0.0006680801 0.0006680801
## [1806] 0.0006680801 0.0006680801 0.0006680801 0.0006680801 0.0006680801
## [1811] 0.0006680801 0.0006680801 0.0006680801 0.0006680801 0.0010143259
## [1816] 0.0010143259 0.0010143259 0.0010143259 0.0010143259 0.0010143259
## [1821] 0.0010143259 0.0010143259 0.0010143259 0.0010143259 0.0010143259
## [1826] 0.0010143259 0.0010143259 0.0010143259 0.0010143259 0.0010143259
## [1831] 0.0010143259 0.0010143259 0.0005854209 0.0005854209 0.0005854209
## [1836] 0.0005854209 0.0005854209 0.0005854209 0.0005854209 0.0005854209
## [1841] 0.0005854209 0.0005854209 0.0005854209 0.0005854209 0.0005854209
## [1846] 0.0005854209 0.0005854209 0.0005854209 0.0005854209 0.0005854209
## [1851] 0.0005854209 0.0005854209 0.0012768032 0.0012768032 0.0012768032
## [1856] 0.0012768032 0.0012768032 0.0012768032 0.0012768032 0.0012768032
## [1861] 0.0012768032 0.0012768032 0.0012768032 0.0012768032 0.0012768032
## [1866] 0.0012768032 0.0012768032 0.0012768032 0.0012768032 0.0012768032
## [1871] 0.0012768032 0.0006378185 0.0006378185 0.0006378185 0.0006378185
## [1876] 0.0006378185 0.0006378185 0.0010247606 0.0010247606 0.0010247606
## [1881] 0.0010247606 0.0010247606 0.0010247606 0.0011384133 0.0011384133
## [1886] 0.0011384133 0.0011384133 0.0011384133 0.0011384133 0.0011384133
## [1891] 0.0011384133 0.0004956928 0.0004956928 0.0004956928 0.0004956928
## [1896] 0.0004956928 0.0004956928 0.0004956928 0.0008234242 0.0008234242
## [1901] 0.0008234242 0.0008234242 0.0008234242 0.0008234242 0.0009524855
## [1906] 0.0009524855 0.0009524855 0.0009524855 0.0009524855 0.0009524855
```

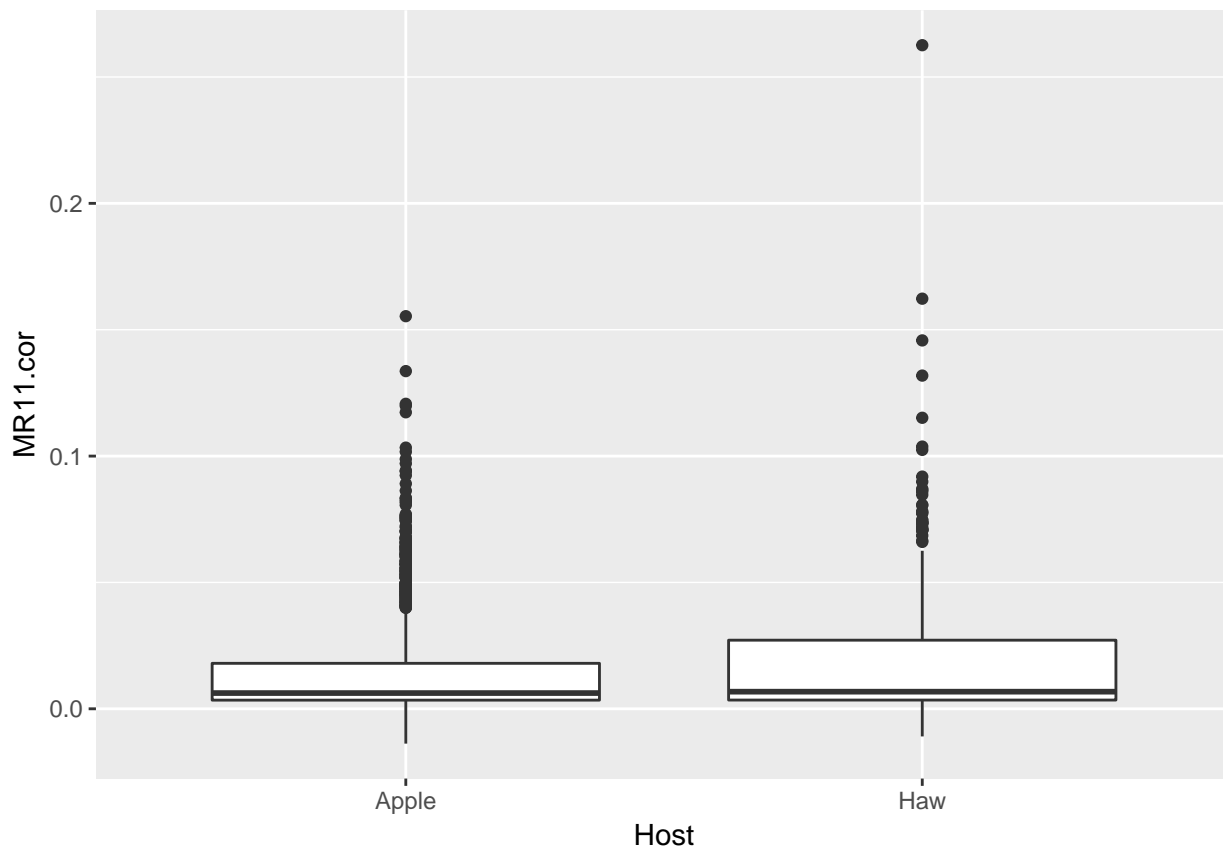
```
#do some corrections
data5 <- data4%>%
  mutate(MR11.cor = MR11 - mean.blank, msMR11.cor = msMR11 - mean.blank)
#glimpse(data5)

data5.neg <- data5%>%
  filter(MR11.cor<0)
```

Figure of Metabolic Rate between Hosts

```
#Boxplot
ggplot(data5, aes(x=Host, y=MR11.cor))+geom_boxplot()
```

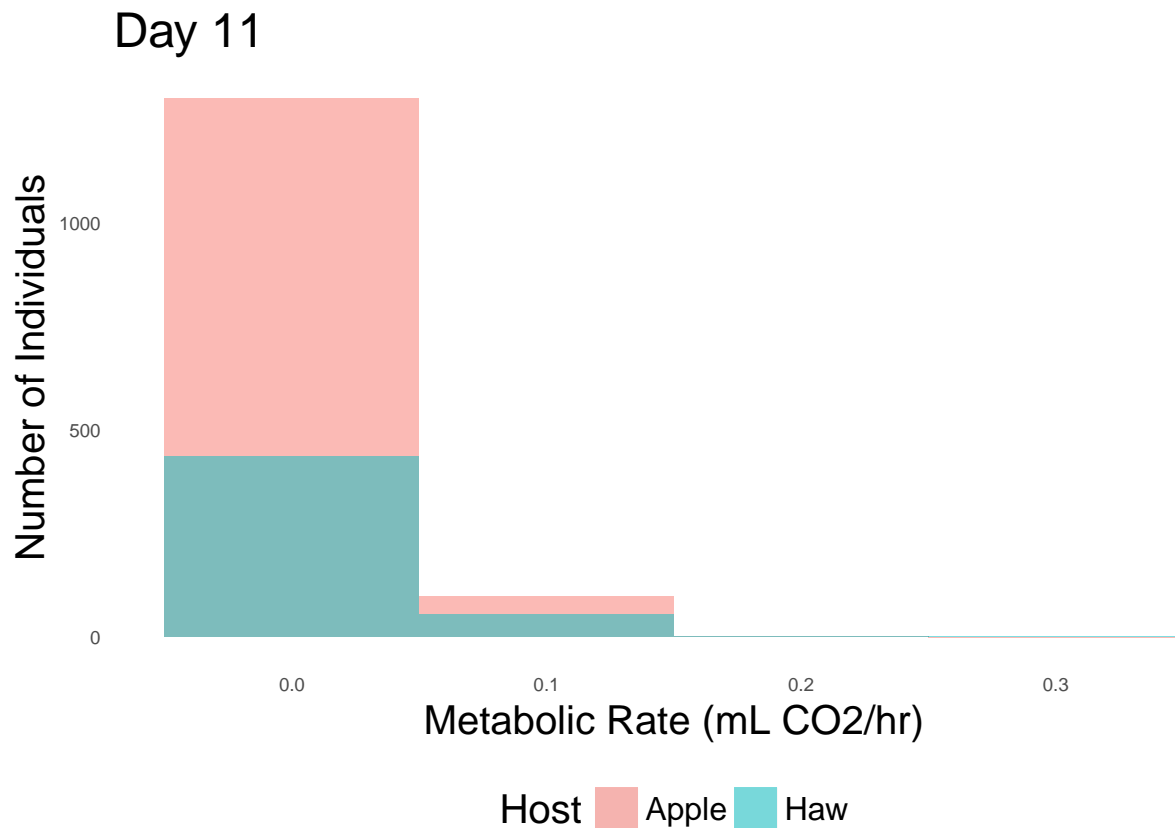
```
## Warning: Removed 13 rows containing non-finite values (stat_boxplot).
```



```
#Histogram
setup<-theme_bw()+theme(axis.text.x=element_blank(),
  axis.ticks.x=element_blank(),legend.position="bottom",
  axis.text.y=element_blank(),
  axis.ticks.y=element_blank(),panel.background = element_blank(),
  panel.grid.major = element_blank(), panel.grid.minor = element_blank(),axis.text=element_text(s

ggplot(data5, aes(x=MR11.cor, fill=Host))+geom_histogram(position = "identity", alpha=.5, binwidth = .1,
  axis.ticks.x=element_blank(),legend.position="bottom",
  axis.ticks.y=element_blank(),panel.background = element_blank(),
  panel.grid.major = element_blank(), panel.grid.minor = element_blank(),axis.text=element_text(s

## Warning: Removed 13 rows containing non-finite values (stat_bin).
```



Calculate eclosion days

```
data5$neweclosions<-difftime(as.Date(data5$eclosion_date), as.Date(data5$Eclosion_reference_date),units="days")
data5$neweclosions <- as.numeric(data5$neweclosions)
```

Figures of eclosion and metabolic rate data

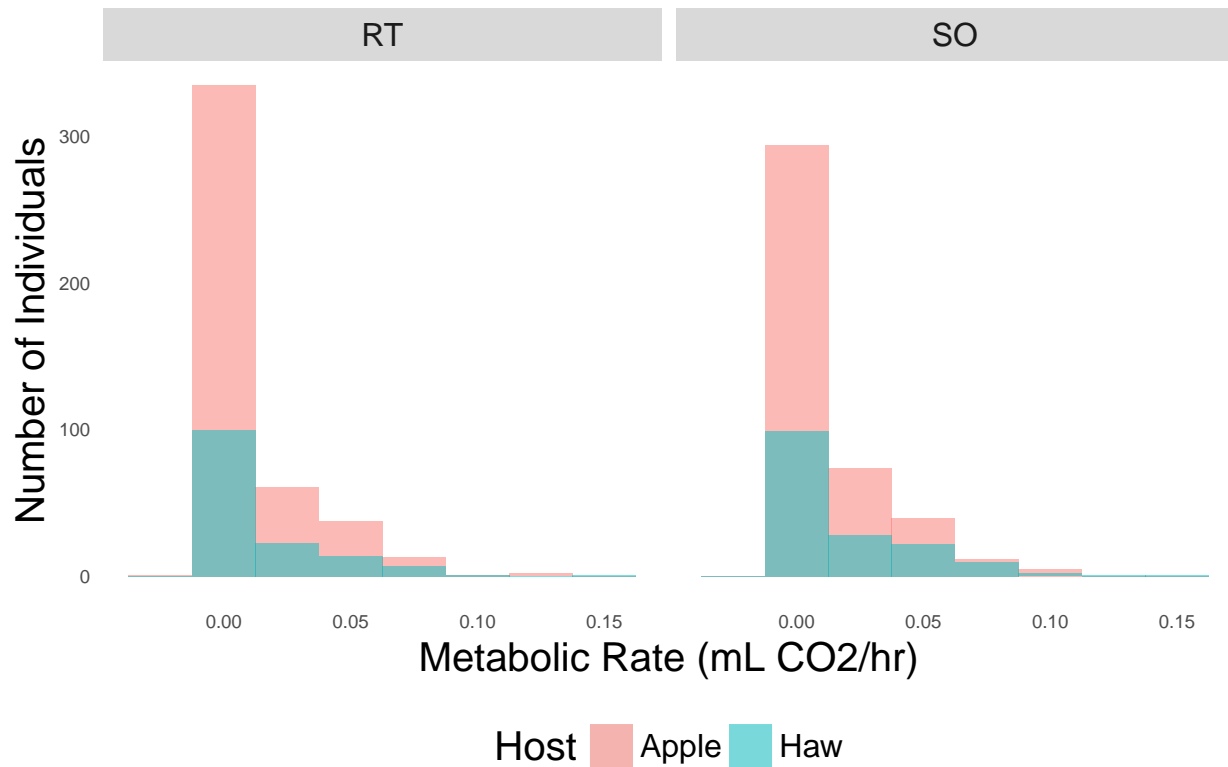
```
data5.treatsub<-data5%>%
  filter(treatment!="GC"&treatment!="")

data5.treatsub$neweclosions<-as.numeric(data5.treatsub$neweclosions)

#Histogram comparison between MR of RT & SO
ggplot(data5.treatsub, aes(x=MR11.cor, fill=Host))+geom_histogram(position = "identity", alpha=.5, binwidth=.05,
  axis.ticks.x=element_blank(),legend.position="bottom",
  axis.ticks.y=element_blank(),panel.background = element_blank(),
  panel.grid.major = element_blank(), panel.grid.minor = element_blank(),axis.text=element_text(size=12))

## Warning: Removed 4 rows containing non-finite values (stat_bin).
```

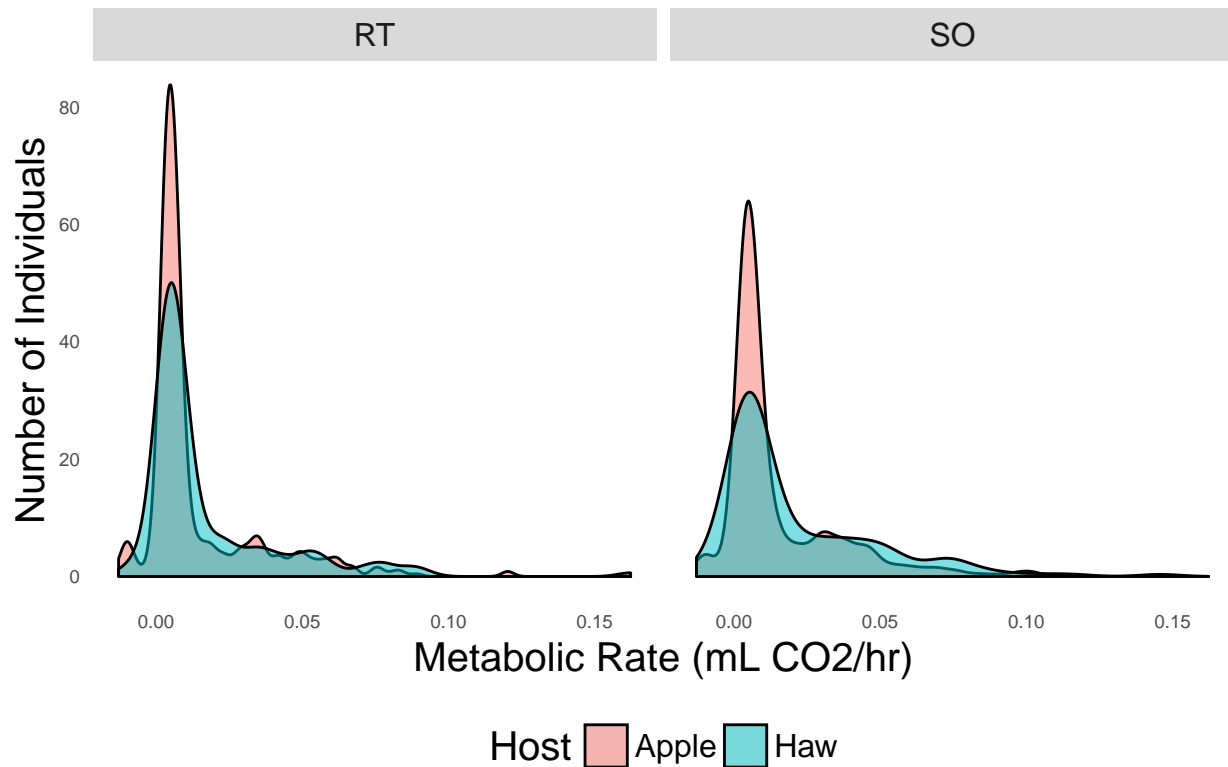
Day 11



```
#Density comparison between MR of RT & SO
ggplot(data5.treatsub, aes(x=MR11.cor, fill=Host))+geom_density(position = "identity", alpha=.5)+facet_
axis.ticks.x=element_blank(),legend.position="bottom",
axis.ticks.y=element_blank(),panel.background = element_blank(),
panel.grid.major = element_blank(), panel.grid.minor = element_blank(),axis.text=element_text(s
```

```
## Warning: Removed 4 rows containing non-finite values (stat_density).
```

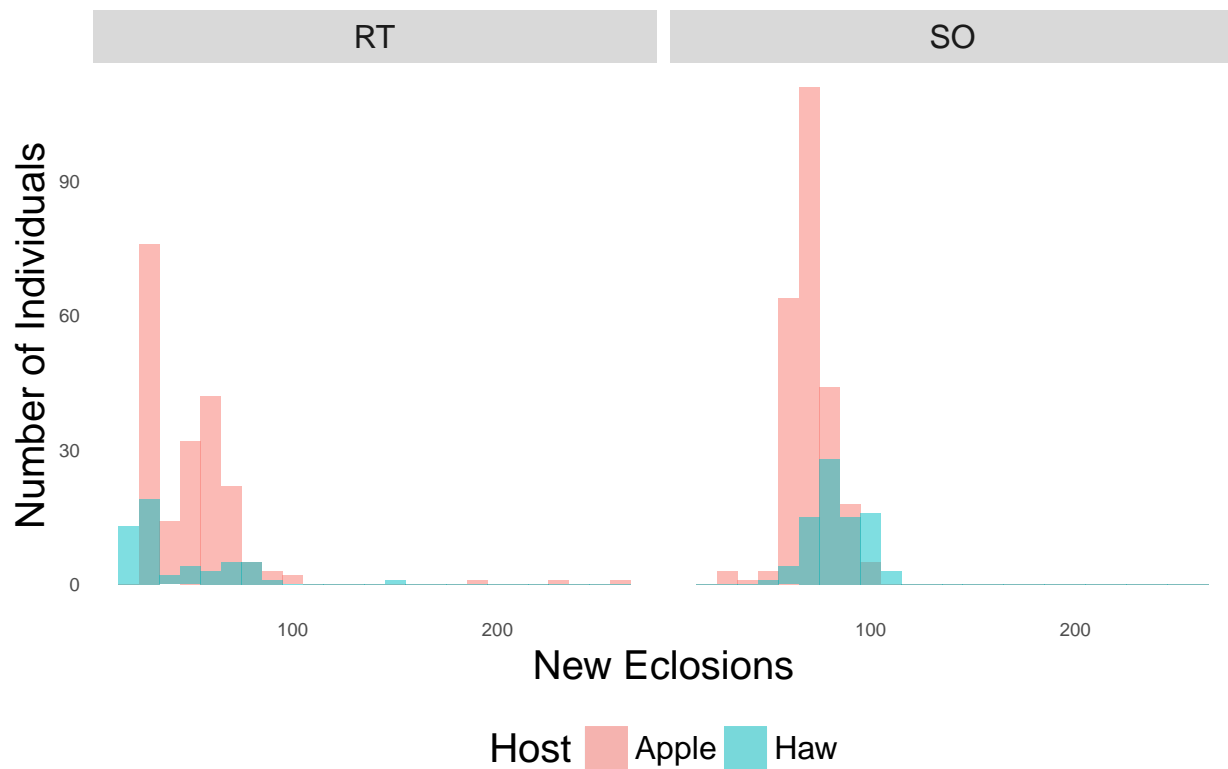
Day 11



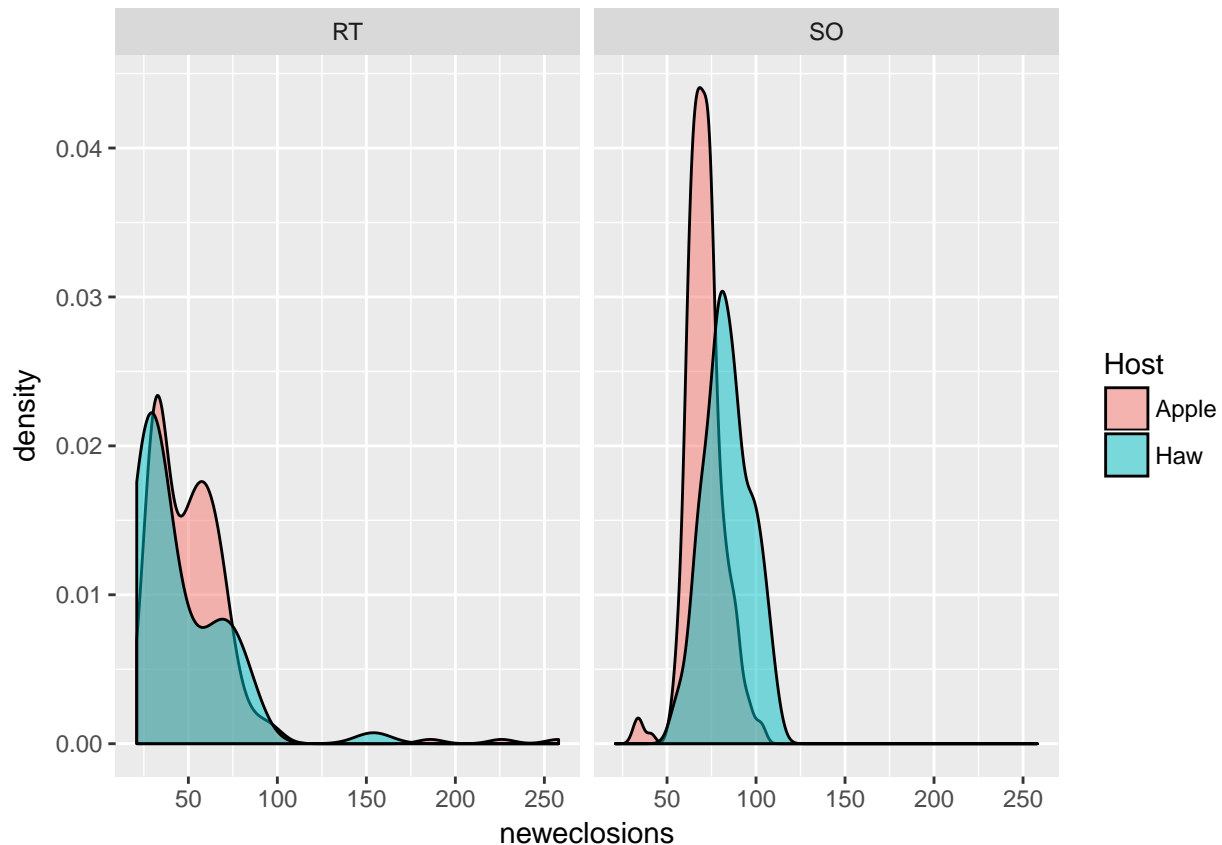
```
#Histogram comparison between eclosion date of RT & SO
ggplot(data5.treatsub, aes(x=neweclosures, fill=Host))+geom_histogram(position = "identity", alpha=.5,
axis.ticks.x=element_blank(),legend.position="bottom",
axis.ticks.y=element_blank(),panel.background = element_blank(),
panel.grid.major = element_blank(), panel.grid.minor = element_blank(),axis.text=element_text(s
```

```
## Warning: Removed 606 rows containing non-finite values (stat_bin).
```

Day 11



```
#Density comparison between eclosion date of RT & SO
ggplot(data5.treatsub, aes(x=neweclosions, fill=Host))+geom_density(position = "identity", alpha=.5)+fa
## Warning: Removed 606 rows containing non-finite values (stat_density).
```



Test differences in eclosion timing between host and experiment (ANOVA)

Using ANOVA which tests differences between two or more means * The null hypothesis would be that the two means are equal * Significance would indicate two means are not equal One Way ANOVA compares two means from two independent groups, in this case eclosion date and host (specific to SO and RT)

Two Way ANOVA compares means of two independent variables affecting one dependent variable, in this case the effect of Host and treatment (interacting) on eclosion

```
#Focus only on RT and SO treatments
data5.treatsub<-data5[%>%
  filter(treatment!="GC"&treatment!="")

#Two Way ANOVA
mod1<- aov(neweclosions ~ Host*treatment, data=data5.treatsub)
summary(mod1)
```

```
##           Df Sum Sq Mean Sq F value    Pr(>F)
## Host       1   3685    3685    9.705 0.00193 **
## treatment  1  82798   82798  218.056 < 2e-16 ***
## Host:treatment  1   9903    9903   26.080 4.45e-07 ***
## Residuals 579 219851     380
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## 606 observations deleted due to missingness
```



```

#narrow down to SO & RT
data5.SO<-data5.treatsub%>%
  filter(treatment=="SO")

data5.RT<-data5.treatsub%>%
  filter(treatment=="RT")

#One Way ANOVA for RT & SO
mod2RT<-aov(neweclosions~Host, data=data5.RT)
summary(mod2RT)

##              Df Sum Sq Mean Sq F value Pr(>F)
## Host          1   2143   2142.9    2.97  0.086 .
## Residuals    250 180353    721.4
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## 346 observations deleted due to missingness

mod2SO<-aov(neweclosions~Host, data=data5.SO)
summary(mod2SO)

##              Df Sum Sq Mean Sq F value Pr(>F)
## Host          1  10064   10064   83.83 <2e-16 ***
## Residuals    329  39499    120
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## 260 observations deleted due to missingness

#how balanced it is
data5.RT%>%
  group_by(Host)%>%
  summarise(n=length(Host))

```

```

## # A tibble: 2 x 2
##   Host      n
##   <chr> <int>
## 1 Apple  452
## 2 Haw   146

```

Both the one way and two way ANOVA tests showed significance between the two means (meaning they are not equal). The two way ANOVA suggests that there is a difference in number of days to eclosion based on treatment type - favorable conditions vs. simulated overwintering. Therefore, a one way ANOVA was done for each treatment to test whether there is any significance between host type and days to eclosion. The one way ANOVA showed significance for both treatments; however, the simulated overwintering samples had a lower p value, which indicates more significance.

Explore relationship between eclosion days and MR

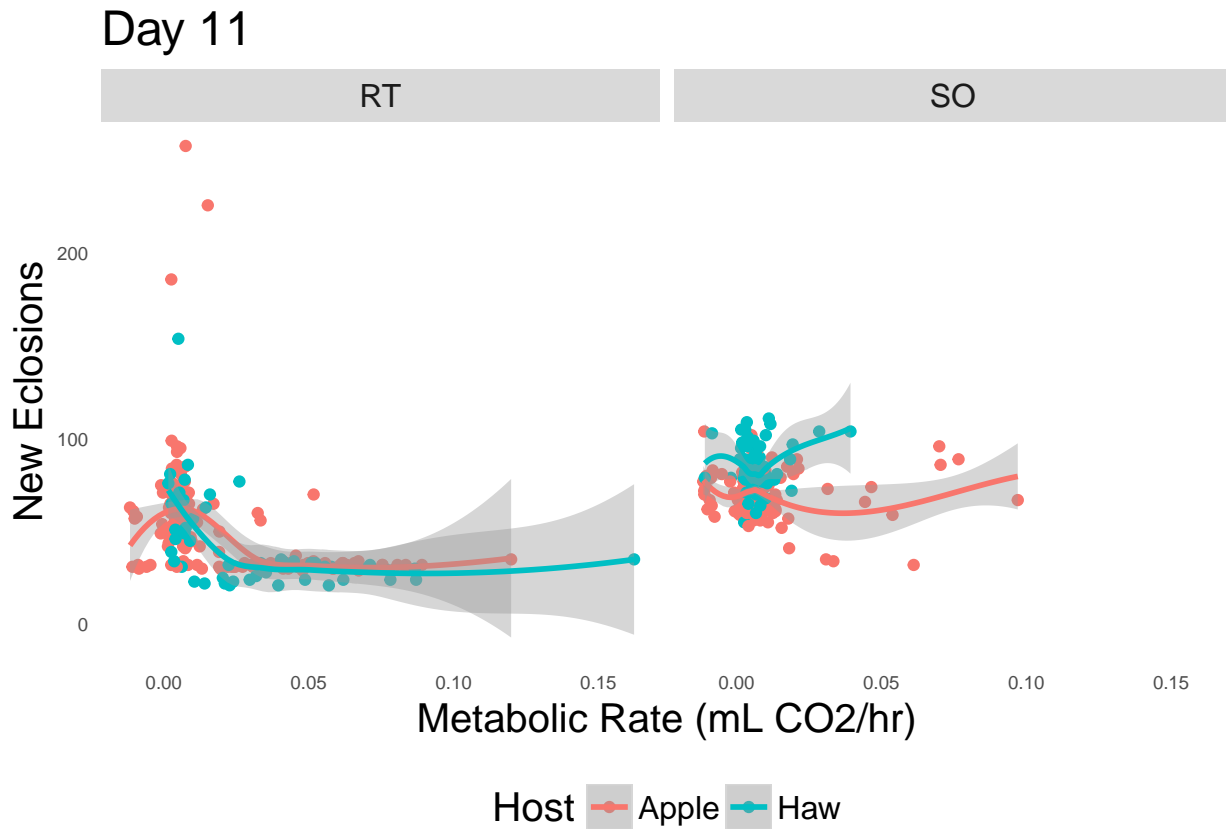
scatter plots

```

ggplot(data5.treatsub,aes(x=MR11.cor, y=neweclosions, colour=Host))+geom_point()+stat_smooth(method="lo
  axis.ticks.x=element_blank(),legend.position="bottom",
  axis.ticks.y=element_blank(),panel.background = element_blank(),
  panel.grid.major = element_blank(), panel.grid.minor = element_blank(),axis.text=element_text(s
## Warning: Removed 609 rows containing non-finite values (stat_smooth).

```

Warning: Removed 609 rows containing missing values (geom_point).

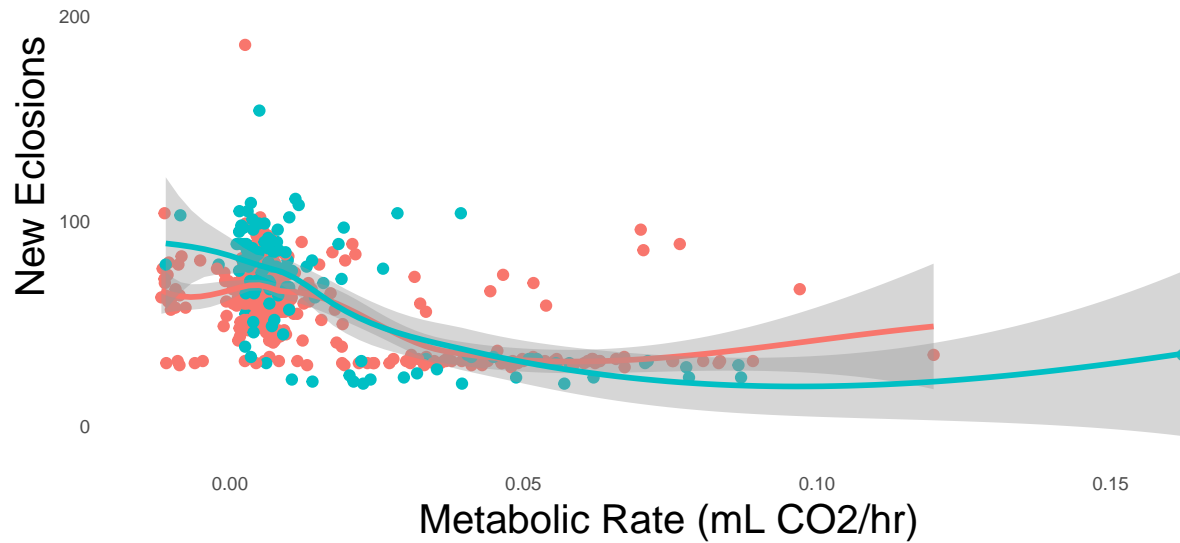


```
#plotting mr11 on the x axis and new eclosions on the y axis; color points by host, fit curve to data
ggplot(data5.treatsub,aes(x=MR11.cor, y=neweclosions, colour=Host))+geom_point()+stat_smooth(method="loess",
axis.ticks.x=element_blank(),legend.position="bottom",
axis.ticks.y=element_blank(),panel.background = element_blank(),
panel.grid.major = element_blank(), panel.grid.minor = element_blank(),axis.text=element_text(s
```

Warning: Removed 609 rows containing non-finite values (stat_smooth).

Warning: Removed 609 rows containing missing values (geom_point).

Day 11



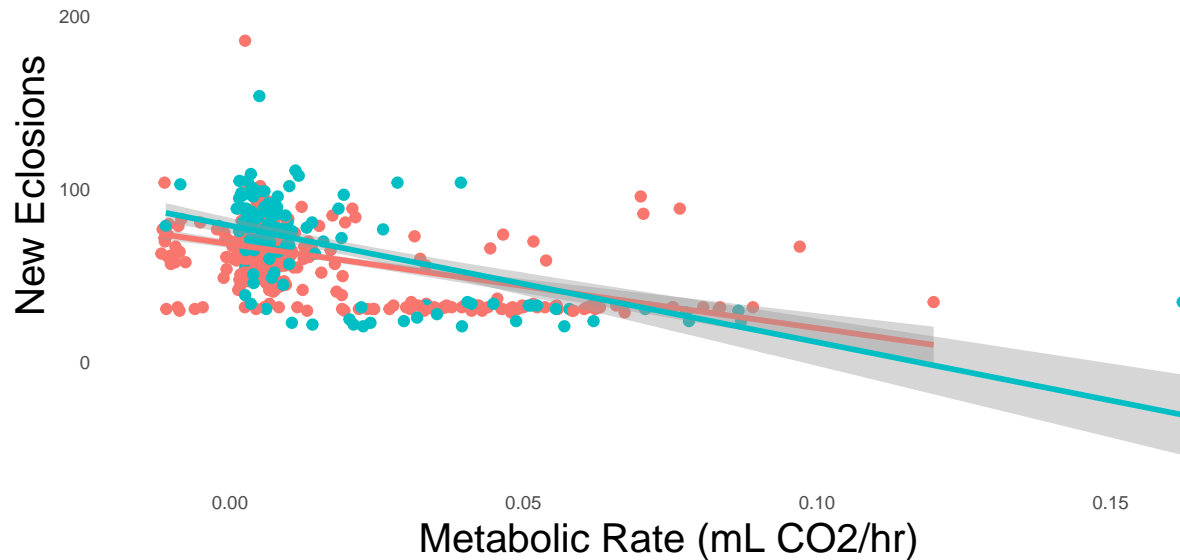
Host —●— Apple —●— Haw

```
#same - fit linear curve to data
ggplot(data5.treatsub,aes(x=MR11.cor, y=neweclosures, colour=Host))+geom_point()+stat_smooth(method="lm",
axis.ticks.x=element_blank(),legend.position="bottom",
axis.ticks.y=element_blank(),panel.background = element_blank(),
panel.grid.major = element_blank(), panel.grid.minor = element_blank(),axis.text=element_text(s
```

```
## Warning: Removed 609 rows containing non-finite values (stat_smooth).
```

```
## Warning: Removed 609 rows containing missing values (geom_point).
```

Day 11



Host —●— Apple —●— Haw

```
#could log transform mr to make it more linear
ggplot(data5.treatsub,aes(x=log10(MR11.cor), y=neweclosions, colour=Host))+geom_point()+stat_smooth(met
axis.ticks.x=element_blank(),legend.position="bottom",
axis.ticks.y=element_blank(),panel.background = element_blank(),
panel.grid.major = element_blank(), panel.grid.minor = element_blank(),axis.text=element_text(s
```

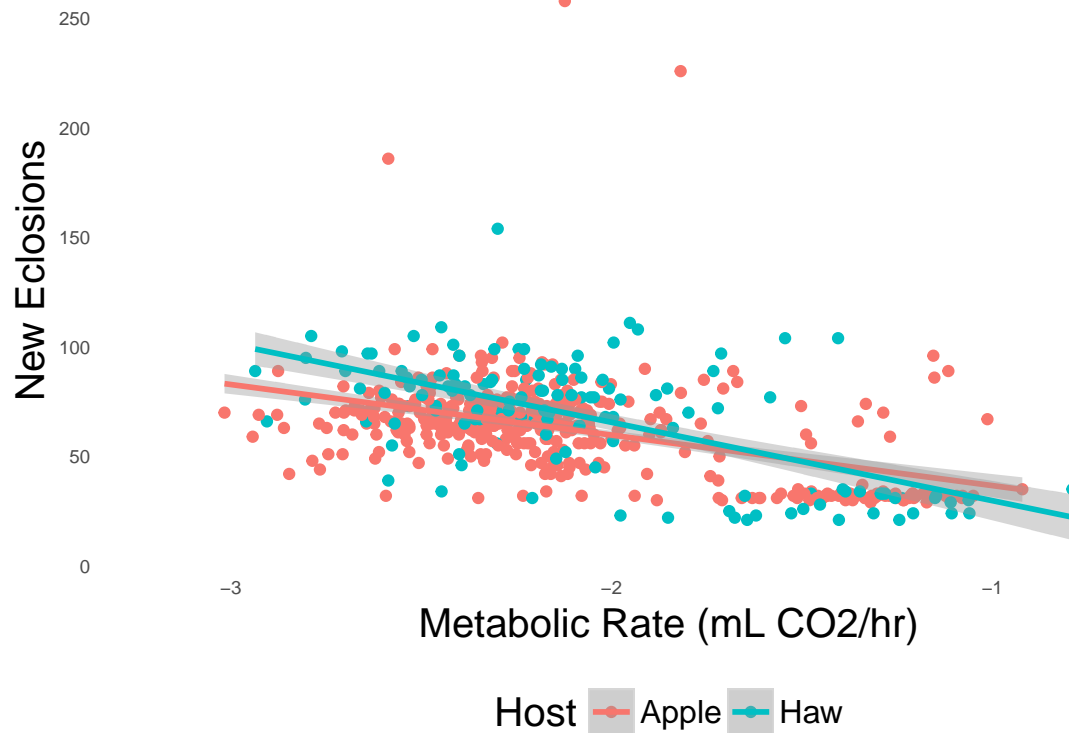
```
## Warning in FUN(X[[i]], ...): NaNs produced
```

```
## Warning in FUN(X[[i]], ...): NaNs produced
```

```
## Warning: Removed 644 rows containing non-finite values (stat_smooth).
```

```
## Warning: Removed 644 rows containing missing values (geom_point).
```

Day 11



Calculating lifespan

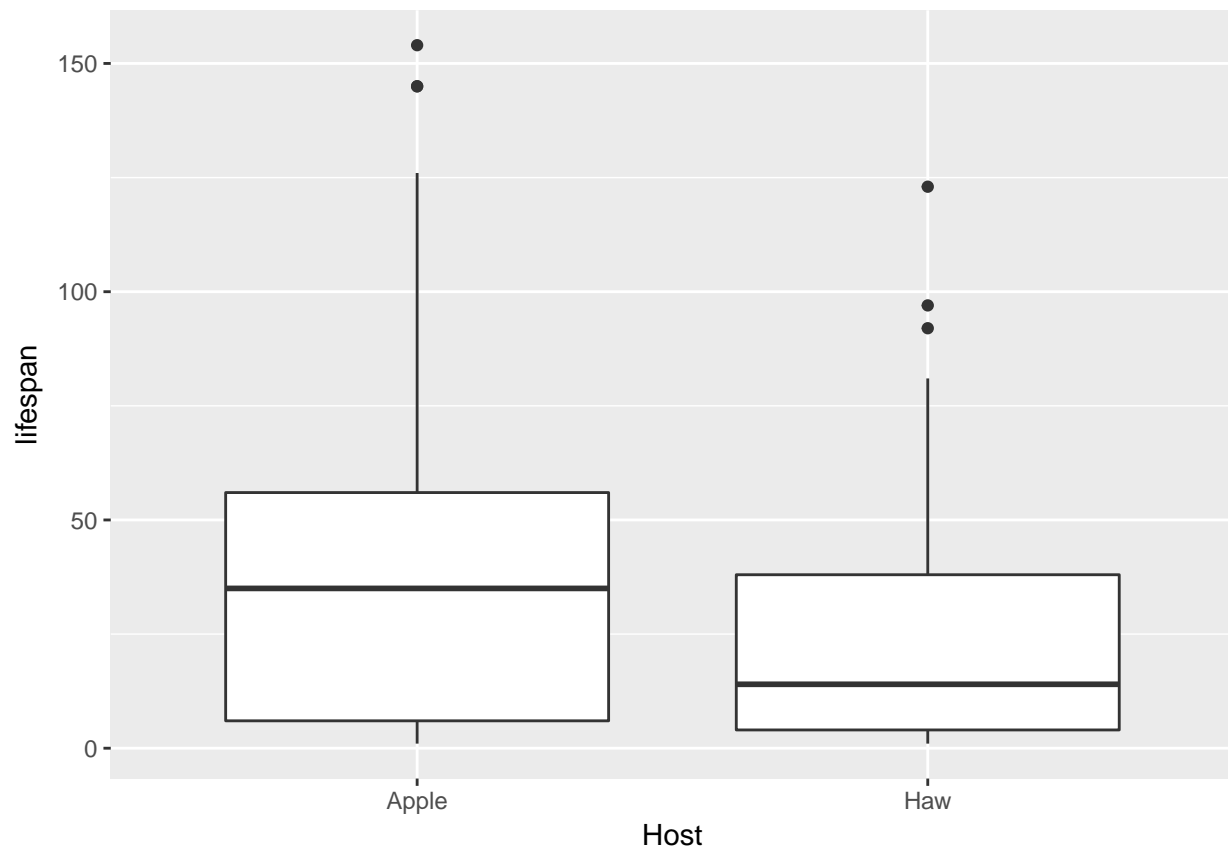
```
data5.treatsub$lifespan<-difftime(as.Date(data5.treatsub$Adult_death_date, na.rm=TRUE), as.Date(data5.t
```

Figures of lifespan between hosts

```
ggplot(data5.treatsub, aes(x=Host, y=lifespan))+geom_boxplot()
```

```
## Don't know how to automatically pick scale for object of type difftime. Defaulting to continuous.
```

```
## Warning: Removed 927 rows containing non-finite values (stat_boxplot).
```

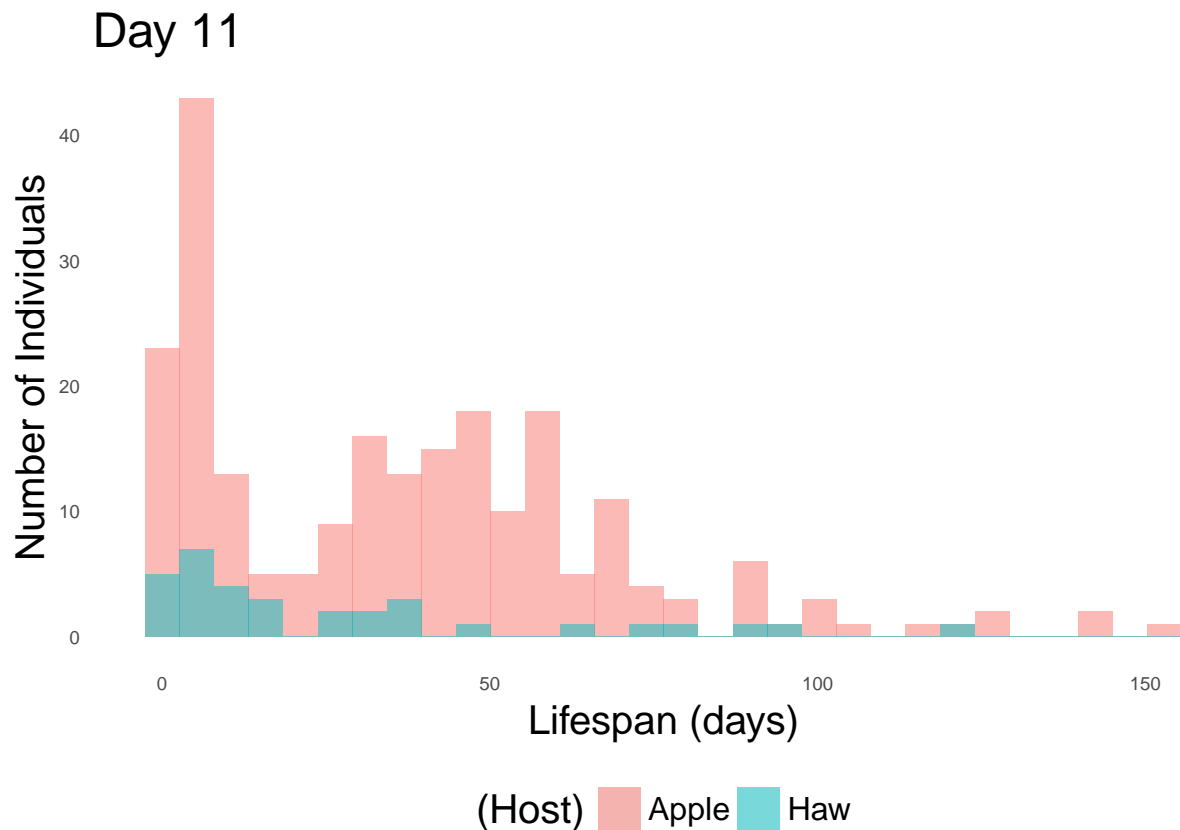


```
ggplot(data5.treatsub, aes(x=lifespan, fill=(Host)))+geom_histogram(position = "identity", alpha=.5) +
  axis.ticks.x=element_blank(),legend.position="bottom",
  axis.ticks.y=element_blank(),panel.background = element_blank(),
  panel.grid.major = element_blank(), panel.grid.minor = element_blank(),axis.text=element_text(s
```

Don't know how to automatically pick scale for object of type difftime. Defaulting to continuous.

`stat_bin()` using `bins = 30`. Pick better value with `binwidth`.

Warning: Removed 927 rows containing non-finite values (stat_bin).

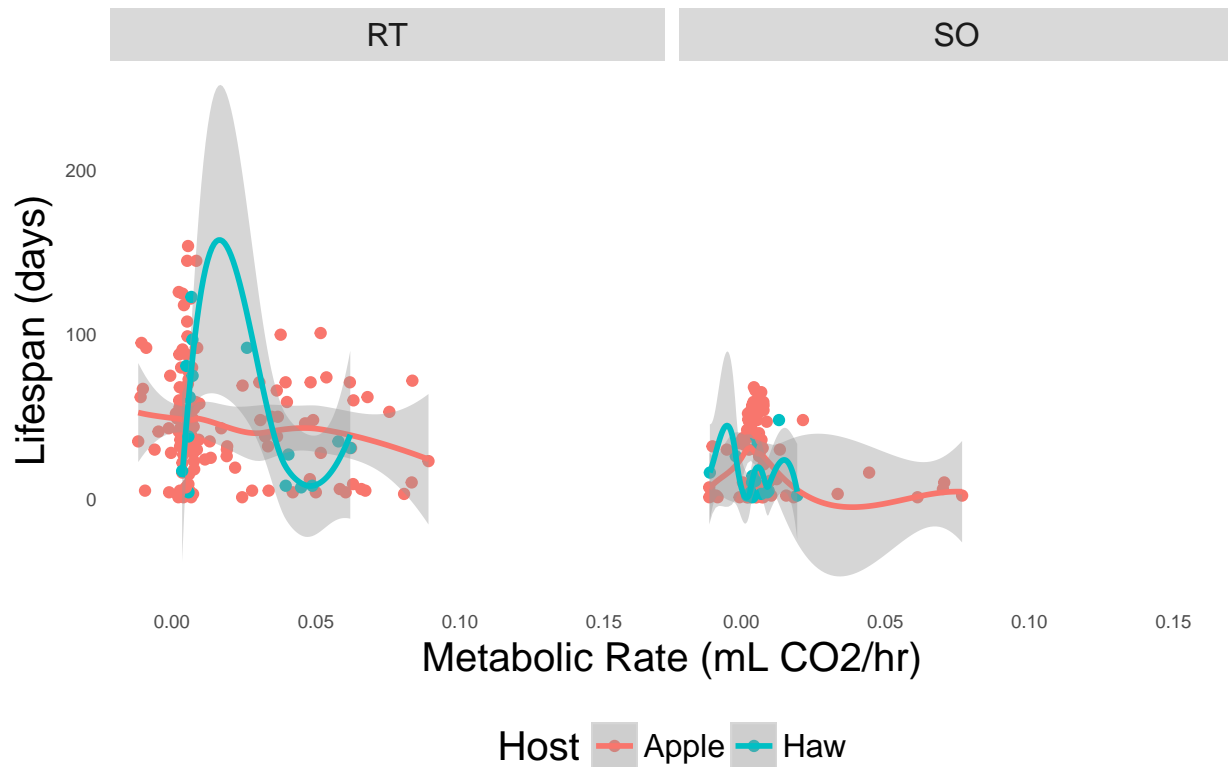


Figures Associating MR with lifespan

```
#Loess
ggplot(data5.treatsub,aes(x= MR11.cor, y=lifespan, colour=Host))+geom_point()+stat_smooth(method="loess",
axis.ticks.x=element_blank(),legend.position="bottom",
axis.ticks.y=element_blank(),panel.background = element_blank(),
panel.grid.major = element_blank(), panel.grid.minor = element_blank(),axis.text=element_text(s
```

```
## Don't know how to automatically pick scale for object of type difftime. Defaulting to continuous.
## Warning: Removed 928 rows containing non-finite values (stat_smooth).
## Warning: Removed 928 rows containing missing values (geom_point).
```

Day 11

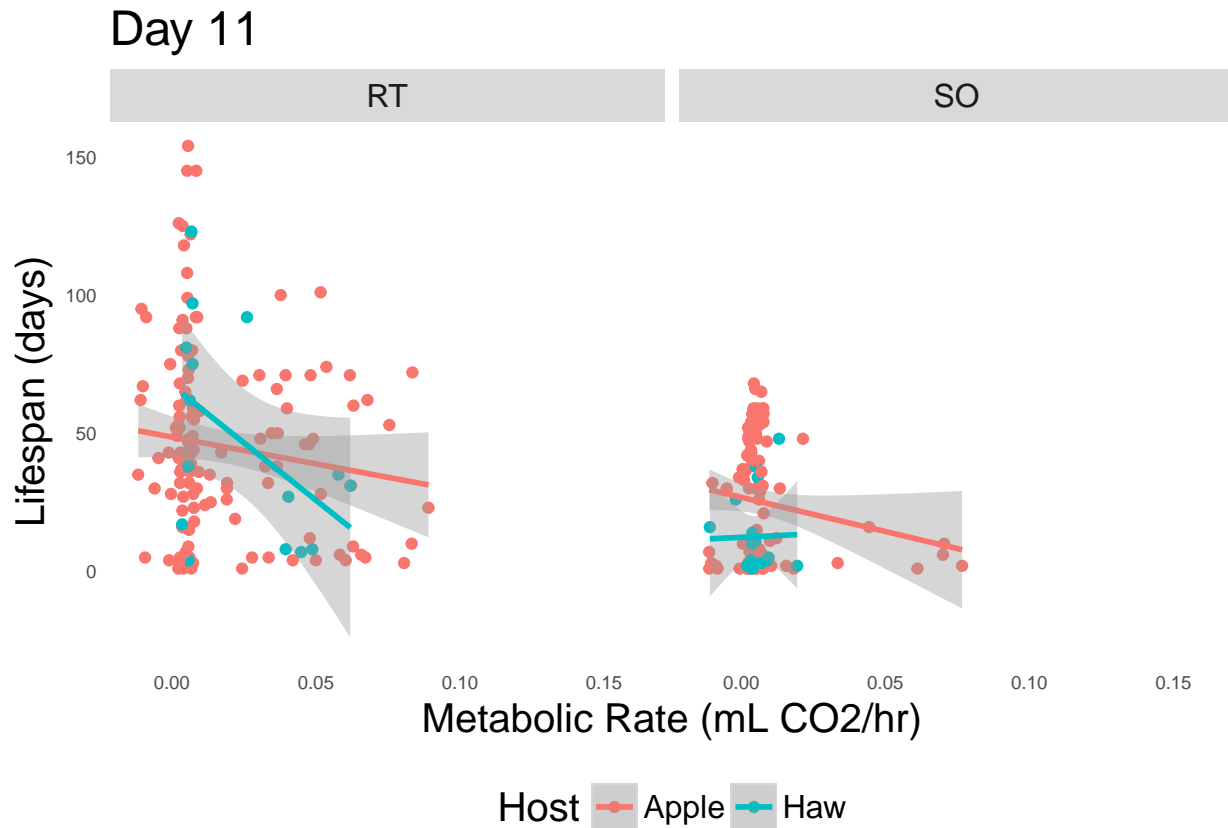


```
#Linear
ggplot(data5.treatsub,aes(x=MR11.cor, y=lifespan, colour=Host))+geom_point()+stat_smooth(method="lm") +
  axis.ticks.x=element_blank(),legend.position="bottom",
  axis.ticks.y=element_blank(),panel.background = element_blank(),
  panel.grid.major = element_blank(), panel.grid.minor = element_blank(),axis.text=element_text(s
```

```
## Don't know how to automatically pick scale for object of type difftime. Defaulting to continuous.
```

```
## Warning: Removed 928 rows containing non-finite values (stat_smooth).
```

```
## Warning: Removed 928 rows containing missing values (geom_point).
```

Day 15 Cohort

Repeat above, but with day 15, starting with time sequence

```
#glimpse(data$purge_time_2)
#hm(data$purge_time_2)
data$day15purge <- lubridate::hour(hm(data$purge_time_2))+lubridate::minute(hm(data$purge_time_2))/60

## Warning in .parse_hms(..., order = "HM", quiet = quiet): Some strings
## failed to parse, or all strings are NAs

## Warning in .parse_hms(..., order = "HM", quiet = quiet): Some strings
## failed to parse, or all strings are NAs

#Getting start and end (min and max) of purges and sample size for each host, cohort day, and tape
param <- data%>%
  group_by(cohort_day, tape)%>%
  summarise(max=max(day15purge, na.rm=TRUE), min=min(day10purge, na.rm=TRUE), n=length(cohort_day))

#goal: for this section, we want a sequence of times for day 15 purge
data2.15 <- data%>%
  group_by(cohort_day, tape)%>%
  mutate(.,day15purge.trans=seq(from = min(day15purge, na.rm=TRUE), to = max(day15purge, na.rm=TRUE), l
#glimpse(data2.15)
```

Calculating start and end time for total amount of hours of CO2 production

```
#glimpse(data$resp_time_2)
#hms(data$resp_time_2)
data2.15$day15resp <- lubridate::hour(hms(data$resp_time_2))+lubridate::minute(hms(data$resp_time_2))/60

## Warning in .parse_hms(..., order = "HMS", quiet = quiet): Some strings
## failed to parse, or all strings are NAs

## Warning in .parse_hms(..., order = "HMS", quiet = quiet): Some strings
## failed to parse, or all strings are NAs

#Obtaining total time in hours
data2.15$total_time_day15 <- (24 - data2.15$day15purge.trans) + data2.15$day15resp
```

Metabolic Rate Calculation

```
#getting denominator for mass-specific mr
data2.15$total_time_day15 * data$mass_day14

##      [1] 151.89085 244.99206 150.46594 234.57320  85.59947 167.88136
##      [7] 141.24237 206.76750 180.33549 107.08496 145.60199  87.54758
##     [13] 219.72889 188.20758 109.40677 166.33217 159.74371 148.21321
##     [19] 139.29352 132.69243 264.27349 145.03540 200.82687 195.30696
##     [25] 137.22043 205.96616 217.92167 169.64426 135.43424 185.96447
##     [31]      NA 189.15792 235.00100 217.26289 107.62898 177.09695
##     [37] 242.53165 217.52194 183.53694  90.97667 220.16532 155.78098
##     [43] 168.99835      NA 253.48203 220.72387 118.98165 138.53605
##     [49]      NA 222.12480 145.28668 208.30287      NA 152.61490
##     [55] 132.65624 118.17762 164.74597 142.90597 125.99792 239.47423
##     [61] 191.25882 137.47027 203.71383 149.26840  86.66776 205.22603
##     [67]  97.64181 170.79156 195.36288 154.11864 181.02458 219.17927
##     [73] 255.98020 170.75549 171.73724 185.57046  94.75507 113.18864
##     [79] 199.16352 217.87870 176.58789 231.42917 217.81565 147.70479
##     [85] 178.31970 151.90227 172.69412 235.22770      NA      NA
##     [91] 173.43143      NA 161.67000  91.22169 160.17619  88.70424
##     [97] 104.30327 154.93066 100.55314 135.34997 158.52950 142.55519
##    [103] 175.23779 139.58000 169.75765 141.66045 140.31098 131.80457
##    [109] 189.45019 141.89187 128.67182      NA 149.68498 167.49996
##    [115] 135.92331 173.37573 123.86904  91.83773 158.17120 115.64842
##    [121] 156.25751 142.34213 166.27355 136.34649 220.06983 167.20783
##    [127] 132.09796      NA 122.61788 146.20692 147.20671 145.89665
##    [133] 164.58226 115.15517 137.13308 188.94774 120.96411 206.39749
##    [139] 133.42545 123.53940 138.02871  93.64513 140.24610  76.19920
##    [145] 145.25555 132.18991 166.64653 204.62503 138.78969 129.51853
##    [151]      NA      NA 104.29522 153.22821 132.91935 125.37235
##    [157]  94.20718 187.64379 154.92192 109.38803 139.56381  95.03147
##    [163]  96.64490 193.94313 130.03819 171.59728 201.32973  75.63907
##    [169] 157.46692 118.95714 120.10717 170.30659 118.87453 188.22498
##    [175] 188.65869 129.95582  93.55194 117.67849 133.91280 105.32226
##    [181] 187.08785 113.11787 149.29462 103.57299 175.68833 138.54515
##    [187] 100.13237 158.88923 102.29142 161.39253 135.52780 125.88503
##    [193] 145.33989 125.99277 194.01531 120.26475 113.70353 133.49780
```

| | | | | | | | |
|----|-------|-----------|-----------|-----------|-----------|-----------|-----------|
| ## | [199] | 213.14419 | 156.11278 | 127.26062 | 151.08781 | 111.35458 | 145.92023 |
| ## | [205] | 77.42203 | 167.94754 | 159.35110 | 118.51955 | 181.54629 | NA |
| ## | [211] | NA | 210.92280 | NA | 239.07768 | 223.58864 | 248.24841 |
| ## | [217] | 147.92914 | 282.79218 | 160.03243 | 229.96373 | 67.66099 | 240.47365 |
| ## | [223] | 156.70936 | 185.37084 | 188.98484 | 167.57158 | 293.79324 | 169.37080 |
| ## | [229] | 227.94742 | 212.17937 | 193.65370 | 290.28204 | 204.75771 | 166.73981 |
| ## | [235] | 200.04701 | 237.14783 | 217.78539 | 233.92074 | 158.65920 | 235.80375 |
| ## | [241] | NA | 128.62508 | 195.07568 | 213.04145 | 128.77189 | 211.25375 |
| ## | [247] | 188.84655 | 169.16884 | NA | 289.22014 | 180.28914 | 209.93716 |
| ## | [253] | 174.98840 | 189.26678 | 216.32975 | 148.26766 | 176.24423 | NA |
| ## | [259] | NA | 181.98857 | 221.75317 | 157.50294 | 253.53012 | 325.74404 |
| ## | [265] | 84.33469 | 163.15817 | 232.00944 | 286.34917 | 222.37025 | 257.77629 |
| ## | [271] | 156.07641 | 177.37436 | 62.72913 | 106.17287 | 114.52053 | 249.50545 |
| ## | [277] | 115.89834 | 117.52800 | 243.54476 | 109.07969 | 145.23401 | 226.83525 |
| ## | [283] | 134.20610 | 192.24333 | 193.13049 | 237.25176 | 294.77150 | NA |
| ## | [289] | NA | 141.73390 | 153.19699 | 234.38103 | 206.95478 | 222.37235 |
| ## | [295] | 143.53541 | 208.47114 | 150.71091 | 232.84130 | 287.52325 | 169.39526 |
| ## | [301] | NA | 174.79913 | 242.34941 | NA | 145.05671 | 95.77601 |
| ## | [307] | NA | 213.37724 | 187.11563 | 244.50529 | 155.81848 | 161.84259 |
| ## | [313] | 182.03143 | 112.76965 | 205.86262 | 140.41052 | 255.69964 | 53.84486 |
| ## | [319] | 167.63052 | 140.58888 | 243.49872 | 100.45010 | NA | NA |
| ## | [325] | 190.53460 | 164.26525 | 297.64252 | 206.97028 | 250.82632 | 231.69705 |
| ## | [331] | 79.80375 | 176.82501 | 224.76932 | 242.91789 | 165.13514 | 228.38918 |
| ## | [337] | 212.14231 | 195.48392 | 299.80077 | 270.92393 | 162.20514 | 114.05928 |
| ## | [343] | 136.39667 | 273.04876 | 189.01511 | 185.60400 | 100.40243 | 118.61826 |
| ## | [349] | 201.64137 | 172.45530 | NA | 150.90512 | 164.05921 | NA |
| ## | [355] | 184.42703 | 145.44548 | 170.38304 | 215.17668 | 276.40605 | 209.48872 |
| ## | [361] | 159.02165 | 296.17420 | 196.56105 | 180.88194 | 166.41854 | NA |
| ## | [367] | NA | 106.67487 | 137.35168 | 88.76411 | 168.85505 | 118.39758 |
| ## | [373] | 142.94495 | 100.82789 | 166.33404 | 105.20512 | 127.10385 | 158.85221 |
| ## | [379] | 109.20588 | 95.85304 | 98.98832 | 109.00858 | 140.12302 | 157.83592 |
| ## | [385] | 119.82423 | 92.39381 | 104.71804 | 109.72455 | 153.32995 | 139.37617 |
| ## | [391] | 146.16490 | 94.09549 | 124.55999 | 135.25201 | 91.96828 | 96.74815 |
| ## | [397] | 110.90906 | 142.39795 | 95.36705 | 178.27287 | 117.45217 | 106.64339 |
| ## | [403] | 125.29231 | 139.78389 | 130.11250 | 138.06234 | 137.08574 | 132.11241 |
| ## | [409] | 107.38332 | 146.45435 | 112.94472 | 162.99451 | 143.87179 | 182.31400 |
| ## | [415] | 112.45118 | 137.21087 | 190.81286 | 157.89329 | 154.95165 | 130.99045 |
| ## | [421] | 213.85285 | 113.13184 | 106.97173 | 169.79517 | 95.25265 | 130.93571 |
| ## | [427] | 177.15243 | 103.61381 | 153.60891 | 94.38480 | 161.91280 | 176.81406 |
| ## | [433] | 156.80795 | 80.57306 | 125.48539 | 140.97935 | 124.67357 | 165.00013 |
| ## | [439] | 107.94375 | 174.47320 | NA | NA | 129.29240 | 98.38951 |
| ## | [445] | 138.61858 | 57.39732 | 132.48440 | 95.36688 | 144.12392 | 113.79814 |
| ## | [451] | 152.27066 | 125.64837 | 97.29221 | 112.23995 | 113.01147 | 112.71854 |
| ## | [457] | 134.97585 | 146.84429 | 41.69218 | 73.92982 | 127.53707 | 137.16758 |
| ## | [463] | 146.45643 | 174.32481 | 127.84530 | 153.82080 | 111.18386 | 163.63502 |
| ## | [469] | 136.81643 | 145.17647 | 118.76511 | 134.58213 | 102.46473 | 95.58704 |
| ## | [475] | 163.10545 | 106.27280 | 159.77890 | 115.69885 | 131.93143 | 173.26923 |
| ## | [481] | 146.03875 | 151.68940 | 149.68239 | 162.22895 | 141.29655 | 98.57041 |
| ## | [487] | 178.88688 | 99.90326 | 118.76540 | 99.20152 | 140.96040 | 179.11386 |
| ## | [493] | 122.69171 | 167.50831 | 133.38262 | 172.47963 | 178.95173 | 77.08516 |
| ## | [499] | 109.29514 | 168.64010 | 78.59833 | 117.25660 | 108.77277 | 170.22154 |
| ## | [505] | 84.95161 | 120.13937 | 219.97006 | 89.17671 | 200.95708 | 174.88358 |
| ## | [511] | 149.52354 | 118.90944 | 111.61225 | 107.44541 | 116.19183 | 64.28032 |
| ## | [517] | 112.54550 | NA | NA | 119.63233 | 176.70951 | 121.92029 |

| | | | | | | | |
|----|-------|-----------|-----------|-----------|-----------|-----------|-----------|
| ## | [523] | 97.48770 | 212.66488 | 190.25181 | 226.61630 | 176.78887 | 141.70688 |
| ## | [529] | 120.48896 | 149.72767 | 114.02913 | 162.22790 | 137.72680 | 167.46930 |
| ## | [535] | 198.21471 | 222.22414 | 144.31799 | 168.20951 | 122.05592 | 221.65199 |
| ## | [541] | 138.23713 | 130.08095 | 104.97054 | 108.50448 | 102.49981 | 152.72634 |
| ## | [547] | 114.30883 | 150.95225 | 256.50053 | 58.61762 | 234.15292 | 202.58850 |
| ## | [553] | 166.46562 | 102.30400 | 73.83530 | 176.43906 | 215.49304 | 172.10703 |
| ## | [559] | 149.25629 | 129.33603 | 177.53887 | 199.08242 | 168.44693 | 116.91421 |
| ## | [565] | 242.54057 | 156.51122 | 147.17754 | 138.86507 | 200.17635 | 217.63522 |
| ## | [571] | 119.46102 | 157.40750 | 125.33768 | 180.01474 | 174.39298 | 133.33263 |
| ## | [577] | 268.23300 | 95.02816 | 174.07383 | 99.68096 | 155.36923 | 155.35262 |
| ## | [583] | 202.45228 | 197.44912 | 95.08334 | 196.12843 | 168.90263 | 117.00722 |
| ## | [589] | 247.23883 | 133.80897 | 52.08961 | 168.40112 | 257.18133 | 184.64638 |
| ## | [595] | 158.13157 | NA | NA | 144.28510 | 161.79047 | 152.77798 |
| ## | [601] | 118.06475 | 148.65257 | 146.78588 | 221.23474 | 183.70374 | 146.20556 |
| ## | [607] | 153.38014 | 156.39290 | 100.28556 | 146.89129 | 43.57708 | 200.00809 |
| ## | [613] | 215.19587 | 200.90385 | 222.13625 | 159.76772 | 179.72952 | 235.60384 |
| ## | [619] | 143.83959 | 165.56589 | 169.19188 | 205.23061 | 136.60399 | 96.23867 |
| ## | [625] | 159.47762 | 141.69077 | 81.17816 | 230.40825 | 251.64143 | 246.48242 |
| ## | [631] | 144.95418 | 125.70535 | 169.03649 | 182.87705 | 147.04253 | 117.82683 |
| ## | [637] | 176.94333 | 187.24581 | 145.36384 | 146.38374 | 221.59893 | 123.19011 |
| ## | [643] | 76.97929 | 136.66649 | 115.38099 | 219.37054 | 230.25630 | 142.11877 |
| ## | [649] | 241.13554 | 147.05367 | 101.22302 | 125.97621 | 133.53677 | 129.80641 |
| ## | [655] | 98.53129 | 95.76844 | 163.06612 | 161.59682 | 132.83680 | 153.18395 |
| ## | [661] | 250.72770 | 150.98030 | 123.52719 | 139.50187 | 240.54657 | 125.51468 |
| ## | [667] | 190.46049 | 159.52057 | 146.56710 | 125.12594 | 152.86735 | 203.88902 |
| ## | [673] | 189.34695 | 119.51908 | NA | NA | 236.67933 | 168.91685 |
| ## | [679] | 165.23486 | 90.71093 | 188.43320 | 206.56077 | 202.94191 | 239.68368 |
| ## | [685] | 237.39638 | 274.29968 | 142.39446 | 104.20104 | 204.31214 | 101.18085 |
| ## | [691] | 82.51267 | 190.37467 | 112.13394 | 273.50354 | 212.95399 | 97.87156 |
| ## | [697] | 229.50360 | 168.34751 | 257.51921 | 112.89832 | 170.36071 | 279.51447 |
| ## | [703] | 158.15610 | 178.87214 | 196.34760 | 228.83767 | 150.31899 | NA |
| ## | [709] | NA | NA | 311.66841 | 182.39797 | 228.87327 | 245.15594 |
| ## | [715] | 160.90599 | NA | 314.77844 | 171.42461 | 203.05210 | 175.45414 |
| ## | [721] | 200.83334 | 245.11307 | 104.70096 | 180.36082 | 157.88619 | 144.06367 |
| ## | [727] | 213.12673 | 254.52983 | 176.09792 | 168.73682 | 164.26036 | 210.17226 |
| ## | [733] | 226.54403 | NA | 188.52128 | 182.89847 | 137.98934 | 116.97289 |
| ## | [739] | 186.88617 | 96.42372 | 147.47906 | 176.37094 | 233.86559 | 113.55891 |
| ## | [745] | 165.27355 | 274.09870 | 169.22480 | 109.41060 | 281.53706 | 262.80691 |
| ## | [751] | 320.78118 | 182.44865 | 129.90357 | 251.00006 | 216.98010 | 262.31409 |
| ## | [757] | 164.41188 | 170.58110 | 274.51861 | 230.10616 | 138.51344 | 264.17894 |
| ## | [763] | 101.88259 | 166.44373 | 202.28102 | 146.56193 | 192.58559 | 110.35523 |
| ## | [769] | 172.43764 | 134.46004 | 161.91695 | 178.52947 | 279.75390 | 273.48942 |
| ## | [775] | 179.44321 | 250.91752 | 222.00473 | 186.59735 | 194.25126 | 283.03854 |
| ## | [781] | 181.22456 | 166.61449 | 327.83293 | 179.75373 | 241.65299 | 210.24296 |
| ## | [787] | 91.35382 | 195.12751 | 209.76778 | 141.00363 | 167.89859 | 198.79763 |
| ## | [793] | 183.74838 | 142.50235 | 266.73237 | 255.94424 | 122.27842 | 199.87733 |
| ## | [799] | 177.35375 | 275.66366 | 174.81956 | 218.02028 | 172.37171 | 177.11704 |
| ## | [805] | 260.90900 | 217.01619 | 315.04518 | 149.98952 | 145.86244 | 87.65047 |
| ## | [811] | 224.42390 | 132.33197 | 155.25084 | 212.06302 | 187.50740 | NA |
| ## | [817] | NA | 323.41733 | 169.52080 | 134.15429 | 82.67114 | 221.31699 |
| ## | [823] | 138.49607 | 146.33448 | 144.16308 | 166.44108 | 103.64131 | 188.07572 |
| ## | [829] | 115.57467 | 172.78965 | 143.90616 | 147.30084 | 202.85473 | 201.70344 |
| ## | [835] | 196.93784 | 109.24496 | 98.80317 | 173.24057 | 73.20611 | 111.85256 |
| ## | [841] | 128.48519 | 203.65379 | 116.68310 | 63.96380 | 228.85606 | 171.33080 |

| | | | | | | | |
|----|--------|-----------|-----------|-----------|-----------|-----------|-----------|
| ## | [847] | 61.58388 | 203.32356 | 110.45579 | 109.52558 | 35.35027 | 30.06423 |
| ## | [853] | 120.86745 | 160.90368 | 179.67191 | 112.51395 | 149.79841 | 138.82326 |
| ## | [859] | 113.31152 | 113.91065 | 187.58133 | 179.81861 | 175.59594 | 177.68839 |
| ## | [865] | 130.71979 | 190.95666 | 107.41615 | 153.89902 | 77.57568 | 187.11961 |
| ## | [871] | 156.95540 | 115.77999 | 125.86917 | 166.54506 | 155.09568 | 140.92224 |
| ## | [877] | 231.20393 | 246.98775 | 201.31696 | 140.41804 | 120.57412 | 322.41481 |
| ## | [883] | 74.52438 | 190.33830 | 243.87155 | 215.68673 | 278.05120 | 184.20396 |
| ## | [889] | 282.16279 | 286.47307 | 281.79190 | 197.98659 | 194.85279 | NA |
| ## | [895] | NA | 198.57162 | 153.19246 | 110.29079 | 173.70045 | 188.78256 |
| ## | [901] | 153.36750 | 153.42593 | 120.23252 | 10.85755 | 67.72840 | 80.71602 |
| ## | [907] | 135.94162 | 116.41159 | 82.74139 | 156.12541 | NA | 74.40171 |
| ## | [913] | 119.59107 | 147.51089 | 134.32303 | 119.03206 | 131.75652 | NA |
| ## | [919] | 145.63663 | 121.61368 | 149.12887 | 209.06062 | 151.70809 | 115.55536 |
| ## | [925] | 201.16650 | 99.72531 | 142.23510 | 223.61960 | 93.18641 | 123.62597 |
| ## | [931] | 108.23865 | 130.73022 | 102.05851 | 205.89768 | 108.89413 | 88.75034 |
| ## | [937] | 154.85725 | 136.38411 | 118.33598 | 170.22920 | 158.20073 | 45.55562 |
| ## | [943] | 152.87023 | 135.86050 | 172.23985 | 192.00503 | 172.41881 | 211.03935 |
| ## | [949] | 150.95938 | 127.73553 | 45.61205 | 115.26292 | 223.55963 | 145.55478 |
| ## | [955] | 306.67086 | 289.74836 | 242.09249 | 249.96737 | 234.70742 | 169.80472 |
| ## | [961] | 234.88993 | 376.96925 | 281.22783 | 234.70519 | 219.96935 | 199.25986 |
| ## | [967] | 212.27750 | 285.11913 | 228.85886 | 290.39371 | NA | NA |
| ## | [973] | 172.65320 | 183.68117 | 153.53304 | 173.16056 | 218.29182 | 137.62199 |
| ## | [979] | 165.65117 | 130.70494 | 210.53709 | 191.39765 | 82.18379 | 137.79446 |
| ## | [985] | 112.79892 | 145.01462 | 114.28057 | 109.50767 | 186.99437 | 130.62650 |
| ## | [991] | 97.55314 | 157.44405 | 154.79311 | 201.94956 | 41.72071 | 116.78201 |
| ## | [997] | 242.28931 | 247.22546 | 223.45850 | 262.33558 | 214.46909 | 132.69326 |
| ## | [1003] | 183.51150 | 146.52745 | 229.40637 | 132.28060 | 128.12840 | 258.12294 |
| ## | [1009] | 213.89346 | 210.77450 | 206.93692 | 198.02815 | 133.14584 | 137.28838 |
| ## | [1015] | 89.87927 | 146.90913 | 322.06512 | 136.24043 | 281.08101 | 228.49547 |
| ## | [1021] | 319.65530 | 203.66199 | 194.20163 | 87.94538 | 119.63981 | 165.24507 |
| ## | [1027] | 223.05722 | 178.62372 | 85.92271 | 275.72205 | 145.84070 | 207.80827 |
| ## | [1033] | 350.34029 | 317.86052 | 96.80861 | 220.35617 | 249.10996 | 171.34718 |
| ## | [1039] | 328.78355 | NA | 87.96872 | 263.14647 | NA | 92.76666 |
| ## | [1045] | 284.18584 | 157.05615 | 160.41360 | 129.19200 | 204.66106 | 192.72184 |
| ## | [1051] | 155.20074 | 245.57807 | 168.26862 | 103.52014 | 133.07120 | 146.01605 |
| ## | [1057] | 184.02752 | 147.88937 | 116.65658 | 92.32872 | 143.53153 | 127.04844 |
| ## | [1063] | 192.41222 | 93.17674 | 209.88858 | 128.30998 | 151.49131 | 93.93060 |
| ## | [1069] | 150.97619 | 137.87455 | 73.70803 | 94.66387 | 182.46425 | 154.15399 |
| ## | [1075] | 231.37282 | 241.40623 | 193.30175 | 264.17044 | 222.21892 | 149.39019 |
| ## | [1081] | 155.57885 | 198.99142 | 55.31328 | 218.51867 | 164.37739 | 207.83587 |
| ## | [1087] | 95.92582 | 161.17564 | 203.59065 | 167.62576 | 157.06691 | 312.41478 |
| ## | [1093] | 320.44203 | 217.00516 | 179.08850 | 272.06279 | 290.17599 | 263.09970 |
| ## | [1099] | 301.36289 | 90.43548 | 215.14794 | 231.05793 | 151.55009 | 214.66855 |
| ## | [1105] | 245.46634 | 183.07296 | NA | 192.94184 | 246.54650 | NA |
| ## | [1111] | 286.82817 | 153.22174 | 222.28400 | 241.36079 | 332.28417 | 227.16926 |
| ## | [1117] | 249.44556 | 299.97528 | 334.62067 | 322.88511 | 182.84323 | 181.84777 |
| ## | [1123] | 241.92458 | 92.76750 | 88.69525 | 143.98714 | 158.37522 | 168.04372 |
| ## | [1129] | 167.24837 | 165.22573 | NA | 102.70816 | 233.88299 | 153.77298 |
| ## | [1135] | 147.09475 | 94.88459 | 97.82111 | 75.31639 | 103.79262 | 132.26950 |
| ## | [1141] | 131.70484 | 122.86012 | 111.56943 | 177.57999 | 65.04169 | 92.93503 |
| ## | [1147] | 102.30778 | 129.49114 | 137.45966 | NA | 127.19144 | 184.66296 |
| ## | [1153] | 126.92067 | 149.67364 | 59.87287 | 120.78404 | 146.47512 | 119.86114 |
| ## | [1159] | 101.53148 | 176.75276 | 113.64982 | NA | 99.57592 | 30.16799 |
| ## | [1165] | 173.08528 | 107.92759 | 108.15704 | 57.41938 | 69.65494 | NA |

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|----|--------|-----------|-----------|-----------|-----------|-----------|-----------|
| ## | [1171] | 102.85278 | 166.62065 | 177.47587 | 97.35236 | 112.54514 | 64.98080 |
| ## | [1177] | 78.49780 | 85.18404 | 139.86676 | 186.89877 | 90.47459 | 72.58539 |
| ## | [1183] | 166.94783 | 115.54217 | 54.80953 | 128.65189 | 93.57555 | 146.30151 |
| ## | [1189] | NA | NA | 119.16528 | 211.31282 | 98.11031 | 117.84093 |
| ## | [1195] | 119.80105 | 131.29539 | 157.27285 | 103.98303 | 111.03694 | 190.83216 |
| ## | [1201] | 106.52553 | 132.28334 | 73.99766 | 148.52123 | 188.46066 | 105.76365 |
| ## | [1207] | 150.75092 | 116.24582 | 172.27732 | 97.39833 | 129.09653 | NA |
| ## | [1213] | 155.50752 | 104.45643 | 141.38853 | NA | 123.47535 | NA |
| ## | [1219] | 124.42136 | 120.44197 | 142.03023 | 164.46617 | 168.92029 | 202.99025 |
| ## | [1225] | 111.61102 | 109.46183 | 167.95669 | 130.73166 | 201.91703 | 125.38394 |
| ## | [1231] | 73.68412 | 73.43288 | 158.84712 | 118.25260 | 160.24879 | 151.06416 |
| ## | [1237] | NA | 132.88299 | 127.60735 | 75.35524 | 152.17302 | 113.14416 |
| ## | [1243] | 168.03050 | 116.99026 | 178.99750 | 143.44193 | 181.78055 | 129.02225 |
| ## | [1249] | NA | 116.18379 | 53.63977 | 144.47140 | 172.36000 | 244.01667 |
| ## | [1255] | 218.99005 | 144.28601 | 159.90376 | 131.50459 | 104.05383 | 152.88184 |
| ## | [1261] | 116.89446 | 166.10940 | NA | 167.42960 | 158.55343 | 137.44062 |
| ## | [1267] | 154.28527 | 99.14746 | 124.24569 | 139.53881 | NA | 165.45042 |
| ## | [1273] | 159.90640 | 142.29002 | 79.91002 | 175.56989 | 147.75732 | 134.67985 |
| ## | [1279] | 234.01888 | 143.55330 | 127.20283 | 102.95157 | 107.88862 | 144.19424 |
| ## | [1285] | 99.46803 | 149.89772 | 219.22127 | 171.91043 | 89.04474 | 134.77723 |
| ## | [1291] | NA | NA | NA | 137.58710 | 166.13676 | 208.74563 |
| ## | [1297] | 145.75332 | 129.46490 | 147.67111 | 217.54619 | 197.68499 | 209.92611 |
| ## | [1303] | 197.96462 | 71.27602 | 139.69641 | 122.65191 | 189.16305 | 163.16684 |
| ## | [1309] | 135.26312 | 115.26506 | 168.64854 | 146.64188 | 156.49874 | 176.44331 |
| ## | [1315] | 202.32102 | 147.41599 | 149.97407 | 132.95669 | 152.96789 | NA |
| ## | [1321] | 199.81508 | 117.15915 | 117.50567 | 119.39058 | 132.41547 | 151.59490 |
| ## | [1327] | 182.56528 | 68.66843 | NA | NA | 121.83458 | 116.84727 |
| ## | [1333] | 163.00653 | 178.98124 | 109.34876 | 113.55042 | 77.75115 | 162.04099 |
| ## | [1339] | 96.31663 | 179.78079 | NA | 136.46086 | 198.04252 | NA |
| ## | [1345] | 112.12598 | 77.56403 | 87.06523 | 109.86987 | NA | 132.50968 |
| ## | [1351] | NA | NA | 111.29147 | 93.77000 | 131.89740 | 157.83322 |
| ## | [1357] | 110.42440 | 130.37814 | 195.87709 | 77.34572 | 75.20604 | 175.04821 |
| ## | [1363] | NA | NA | 120.90100 | 181.78367 | NA | 124.08068 |
| ## | [1369] | 198.06612 | 151.11344 | 174.98207 | 106.29763 | 76.63305 | 126.23087 |
| ## | [1375] | 178.10913 | NA | 151.40395 | 78.69247 | 132.52953 | NA |
| ## | [1381] | 136.69206 | 170.84502 | 83.74507 | 127.50795 | NA | 83.57440 |
| ## | [1387] | 80.36200 | 74.95620 | 190.11447 | 88.62713 | 111.68657 | 131.69102 |
| ## | [1393] | 88.81461 | 119.27532 | 114.94706 | NA | NA | NA |
| ## | [1399] | 85.11790 | 208.96956 | NA | 137.09795 | 100.60509 | NA |
| ## | [1405] | 94.15180 | 110.77646 | 112.77084 | 76.34200 | NA | 130.87021 |
| ## | [1411] | 85.11067 | 176.66107 | 160.55216 | 46.49058 | 122.77251 | 157.58435 |
| ## | [1417] | 114.85455 | 180.72516 | 47.55750 | 49.82075 | 222.28067 | 156.72825 |
| ## | [1423] | 192.32518 | 137.86516 | 118.75153 | 154.49490 | 171.13653 | 127.39541 |
| ## | [1429] | 41.67827 | 156.30701 | NA | NA | 102.52178 | 139.47623 |
| ## | [1435] | NA | NA | 119.66567 | 236.07220 | 121.74142 | 189.48305 |
| ## | [1441] | 162.62837 | NA | 116.84083 | 125.88978 | NA | 244.67069 |
| ## | [1447] | NA | 139.92927 | 113.31414 | 95.29627 | 48.50283 | 140.51240 |
| ## | [1453] | 146.43951 | 78.81370 | 218.29109 | 123.35742 | 185.91030 | NA |
| ## | [1459] | 112.36917 | 46.53600 | 58.05370 | NA | NA | 126.67133 |
| ## | [1465] | NA | NA | 21.98117 | 58.07681 | 97.73594 | 86.29161 |
| ## | [1471] | 104.33818 | 39.45489 | 104.97681 | 155.53929 | 128.79015 | 136.88200 |
| ## | [1477] | 160.23670 | 121.34268 | 90.06578 | 130.07233 | 106.71204 | NA |
| ## | [1483] | 104.33652 | NA | NA | 139.33920 | 141.37884 | NA |
| ## | [1489] | 170.06010 | 116.97167 | 160.57186 | NA | 169.83861 | 118.96600 |

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|----|--------|-----------|-----------|-----------|-----------|-----------|-----------|
| ## | [1495] | 125.37667 | 119.82736 | 86.32223 | 192.16016 | 133.57331 | 41.94566 |
| ## | [1501] | 105.04054 | 116.81096 | NA | NA | 209.23848 | 166.76409 |
| ## | [1507] | 90.18104 | 98.32764 | 177.78933 | 76.49713 | 189.06318 | 232.21556 |
| ## | [1513] | 198.26908 | 157.65782 | 181.37117 | 172.30585 | 59.78497 | 177.88338 |
| ## | [1519] | 183.95332 | 167.10854 | 230.02523 | NA | 215.54974 | 201.50898 |
| ## | [1525] | 112.80392 | 274.20991 | 226.16207 | 158.05350 | 103.14852 | 100.17539 |
| ## | [1531] | 159.97480 | 119.93497 | 165.56494 | 160.68316 | 164.00951 | 190.76995 |
| ## | [1537] | 183.05694 | NA | NA | 149.10296 | 214.80001 | 144.76424 |
| ## | [1543] | 178.52991 | 200.31255 | 175.95890 | 183.33887 | 178.24855 | 245.58114 |
| ## | [1549] | 147.72093 | 263.30008 | 117.47655 | 118.72364 | 245.91689 | 137.95587 |
| ## | [1555] | 230.63749 | 177.45229 | 196.65643 | 101.93301 | 236.50376 | 248.94311 |
| ## | [1561] | 242.58495 | 98.23182 | 245.84062 | 169.47697 | 172.44119 | 114.44906 |
| ## | [1567] | 166.00263 | 147.84572 | 192.37546 | 220.97363 | 77.67518 | 186.77626 |
| ## | [1573] | NA | NA | 79.43571 | 92.24309 | 75.48601 | 125.34606 |
| ## | [1579] | 161.20744 | 68.31486 | 218.24924 | 203.77884 | 78.94708 | 81.34468 |
| ## | [1585] | 96.84808 | 153.67885 | 170.69681 | 121.80736 | 144.49813 | 156.06282 |
| ## | [1591] | 73.20968 | 177.53706 | 197.98525 | 94.15520 | 140.17270 | 166.06029 |
| ## | [1597] | 129.86142 | 140.57108 | 117.42571 | 143.67859 | NA | NA |
| ## | [1603] | 148.85326 | 103.02119 | 152.28510 | 143.84563 | 136.54512 | 202.20190 |
| ## | [1609] | 124.28306 | 51.67115 | 191.41012 | 124.85152 | 132.82674 | 195.58169 |
| ## | [1615] | 192.28484 | 77.13881 | 171.38010 | 172.64505 | 178.81140 | 133.37580 |
| ## | [1621] | 185.69787 | NA | 121.39132 | 196.66217 | 126.38441 | 166.38630 |
| ## | [1627] | 67.91207 | NA | NA | NA | 160.26970 | 161.05106 |
| ## | [1633] | 224.63528 | 174.32319 | 97.34354 | 108.46619 | 193.05100 | 76.36842 |
| ## | [1639] | 226.22979 | 185.55198 | 94.76189 | 174.31425 | 190.58042 | 159.07028 |
| ## | [1645] | 231.45871 | 166.33221 | 126.24730 | 183.84861 | 217.99322 | 99.20873 |
| ## | [1651] | 252.68084 | 96.85710 | 81.13505 | 152.40973 | NA | NA |
| ## | [1657] | 123.73153 | 239.64653 | 154.33846 | 85.85031 | 265.68861 | 163.54675 |
| ## | [1663] | 250.33914 | NA | 179.42565 | 114.18896 | 182.68615 | 190.13307 |
| ## | [1669] | 190.54843 | 262.25482 | 119.48470 | 146.59442 | 169.41549 | 215.11515 |
| ## | [1675] | 129.80599 | 233.69133 | NA | 246.40602 | 95.98497 | 155.47802 |
| ## | [1681] | 196.01322 | NA | NA | 134.50125 | 219.59214 | 152.04994 |
| ## | [1687] | 180.25875 | 195.58610 | 156.19833 | 161.23809 | 197.13476 | NA |
| ## | [1693] | 75.56701 | 72.35841 | 116.12870 | 194.59992 | 128.69037 | 163.67792 |
| ## | [1699] | 141.03004 | 225.71903 | 188.40986 | 63.21329 | 76.09883 | 155.49287 |
| ## | [1705] | 82.06272 | NA | NA | 196.38645 | 142.28533 | 150.06583 |
| ## | [1711] | 91.09380 | 135.44795 | 114.54000 | 85.28640 | 185.15258 | 144.66357 |
| ## | [1717] | 105.45520 | 128.83733 | 142.08705 | 96.83333 | 144.21960 | 168.10950 |
| ## | [1723] | 120.18155 | 199.70853 | 190.39113 | 147.54600 | 217.05440 | 73.70763 |
| ## | [1729] | NA | NA | NA | 248.79800 | 182.77229 | 232.41392 |
| ## | [1735] | 244.36720 | 213.50815 | 221.64391 | 169.24760 | 166.72520 | 140.56383 |
| ## | [1741] | 242.17462 | 172.30871 | 77.61304 | 165.98987 | 177.25161 | 148.76411 |
| ## | [1747] | 164.73786 | 234.53318 | 253.50825 | 48.96582 | 183.86504 | 76.23554 |
| ## | [1753] | 129.85366 | 55.26739 | 53.70520 | 213.67313 | 166.38941 | 214.57532 |
| ## | [1759] | 179.90624 | 132.03194 | 112.45647 | NA | NA | 219.78250 |
| ## | [1765] | 138.01833 | 156.32913 | 156.26399 | 155.67097 | 217.61147 | 73.94129 |
| ## | [1771] | 72.29062 | 66.85996 | 145.28368 | 145.90492 | 217.95305 | 199.61770 |
| ## | [1777] | 215.61712 | 167.38887 | 245.40952 | 81.27106 | 240.01241 | 149.43238 |
| ## | [1783] | 233.98965 | 187.03385 | 63.66374 | 94.71431 | 161.29734 | 236.19405 |
| ## | [1789] | 248.68916 | 177.33225 | 233.25807 | 261.99605 | 95.61695 | NA |
| ## | [1795] | NA | 253.88243 | 190.76824 | 168.04693 | 182.70481 | 77.43916 |
| ## | [1801] | 116.19455 | 192.02371 | 161.06931 | 201.83333 | 232.88043 | 233.91636 |
| ## | [1807] | 233.74640 | 210.00292 | 130.27605 | 208.92148 | 243.84071 | 218.50840 |
| ## | [1813] | NA | NA | 153.58400 | 179.45459 | 219.89365 | 248.00314 |

```
## [1819] 214.50364 189.38241 153.78191 175.82525 93.24695 169.50176
## [1825] 257.13091 88.76400 182.33058 177.13056 262.24014 207.97490
## [1831] NA NA 178.46812 177.07151 246.56321 93.49180
## [1837] 100.34719 231.49734 74.55641 263.23700 151.34114 101.99627
## [1843] 187.35244 95.40382 225.77281 196.91302 137.03403 166.55808
## [1849] 251.21986 150.64306 NA NA 147.06193 145.58719
## [1855] 198.11833 87.37408 185.54678 217.81529 228.37900 172.61262
## [1861] 233.31954 211.46488 216.23507 184.86772 156.03840 226.67209
## [1867] 230.76247 185.24200 183.39856 NA NA 199.68795
## [1873] 164.27075 164.32162 200.03163 NA NA 226.41480
## [1879] 157.70811 235.94518 155.81043 NA NA 260.40647
## [1885] 190.04787 164.00737 274.88237 282.88944 209.70457 NA
## [1891] NA 275.67330 222.38083 166.19319 211.14833 149.70539
## [1897] NA NA 296.31403 301.30817 171.42998 241.50760
## [1903] NA NA NA NA NA NA
## [1909] NA
```

```
#Metabolic Rate
```

```
data2.15$MR15<- data$resp_day15/(data2.15$total_time_day15)
```

```
#Mass specific metabolic rate
```

```
data2.15$msMR15<- data$resp_day15/(data2.15$total_time_day15 * data$mass_day14)
```

Controlling for Blanks

```
data3.15 <- data2.15%>%
  group_by(cohort_day, tape)%>%
  filter(Site_name=="Blank")%>%
  summarise(mean.blank2=mean(MR15,na.rm=TRUE))
```

```
#check columns
```

```
#glimpse(data3.15)
```

```
data3.15$mean.blank2
```

```
## [1] 0.0005641578 0.0004986472 0.0004491015 0.0013998016 0.0007086216
## [6] 0.0004586952 0.0008301721 0.0010078476 0.0006573110 0.0004078814
## [11] 0.0006396536 0.0010072904 0.0009850978 0.0006266569 0.0013045121
## [16] 0.0005749230 0.0006262029 0.0003713112 0.0005021056 0.0006112672
## [21] 0.0004834431 0.0004612535 0.0004729824 0.0005167701 0.0005268374
## [26] 0.0002867896 0.0004687064 0.0003718163 0.0050939720 0.0012094708
## [31] 0.0037153831 0.0004513217 0.0005082236 0.0007341679 0.0004501632
## [36] 0.0002760230 0.0012459416 0.0017019482
```

```
#merge data3 and data2 by cohort day and tape
```

```
data4.15 <- inner_join(data2.15, data3.15, by=c("cohort_day", "tape"))
```

```
data4.15$mean.blank2
```

```
## [1] 0.0005641578 0.0005641578 0.0005641578 0.0005641578 0.0005641578
## [6] 0.0005641578 0.0005641578 0.0005641578 0.0005641578 0.0005641578
## [11] 0.0005641578 0.0005641578 0.0005641578 0.0005641578 0.0005641578
## [16] 0.0005641578 0.0005641578 0.0005641578 0.0005641578 0.0005641578
## [21] 0.0005641578 0.0005641578 0.0005641578 0.0005641578 0.0005641578
## [26] 0.0005641578 0.0005641578 0.0005641578 0.0005641578 0.0005641578
## [31] 0.0005641578 0.0005641578 0.0005641578 0.0005641578 0.0005641578
```


[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

```
mutate(MR15.cor = MR15 - mean.blank2, msMR15.cor = msMR15 - mean.blank2)
#glimpse(data5.15)

#Filter out negatives
data5.15.neg <- data5.15%>%
  filter(MR15.cor<0)
```

Calculate Eclosions Dates

```
#Focus only on RT and SO treatments
data5.15.treatsub<-data5%>%
  filter(treatment!="GC"&treatment!="")

data5.15$neweclosions <- difftime(as.Date(data5.15$eclosion_date), as.Date(data5.15$Eclosion_reference_date), units="days")
data5.15.treatsub$neweclosions<-difftime(as.Date(data5.15.treatsub$eclosion_date), as.Date(data5.15.treatsub$Eclosion_reference_date), units="days")
```

Test differences in eclosion timing between host and experiment for Day 15 (ANOVA)

```
#Two Way ANOVA
mod1<- aov(neweclosions ~ Host*treatment, data=data5.15.treatsub)
summary(mod1)
```

```
##              Df Sum Sq Mean Sq F value    Pr(>F)
## Host          1   3685    3685    9.705 0.00193 **
## treatment     1  82798   82798  218.056 < 2e-16 ***
## Host:treatment 1   9903    9903   26.080 4.45e-07 ***
## Residuals    579 219851     380
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## 606 observations deleted due to missingness
```

```
#narrow down to SO & RT
data5.15SO<-data5.15.treatsub%>%
  filter(treatment=="SO")
```

```
data5.15RT<-data5.15.treatsub%>%
  filter(treatment=="RT")
```

```
#One Way ANOVA for RT & SO
mod2RT15<-aov(neweclosions~Host, data=data5.15)
summary(mod2RT15)
```

```
##              Df Sum Sq Mean Sq F value    Pr(>F)
## Host          1   3685    3685    6.85 0.0091 **
## Residuals    581 312552     538
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## 1326 observations deleted due to missingness
```



```

mod2S015<-aov(neweclosures~Host, data=data5.15)
summary(mod2S015)

##              Df Sum Sq Mean Sq F value Pr(>F)
## Host          1   3685    3685    6.85 0.0091 **
## Residuals    581 312552     538
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## 1326 observations deleted due to missingness

#how balanced is it?
data5.S0%>%
  group_by(Host)%>%
  summarise(n=length(Host))

## # A tibble: 2 x 2
##   Host      n
##   <chr> <int>
## 1 Apple   428
## 2 Haw     163

```

This ANOVA test reflects the same results from the ANOVA done on Day 11.

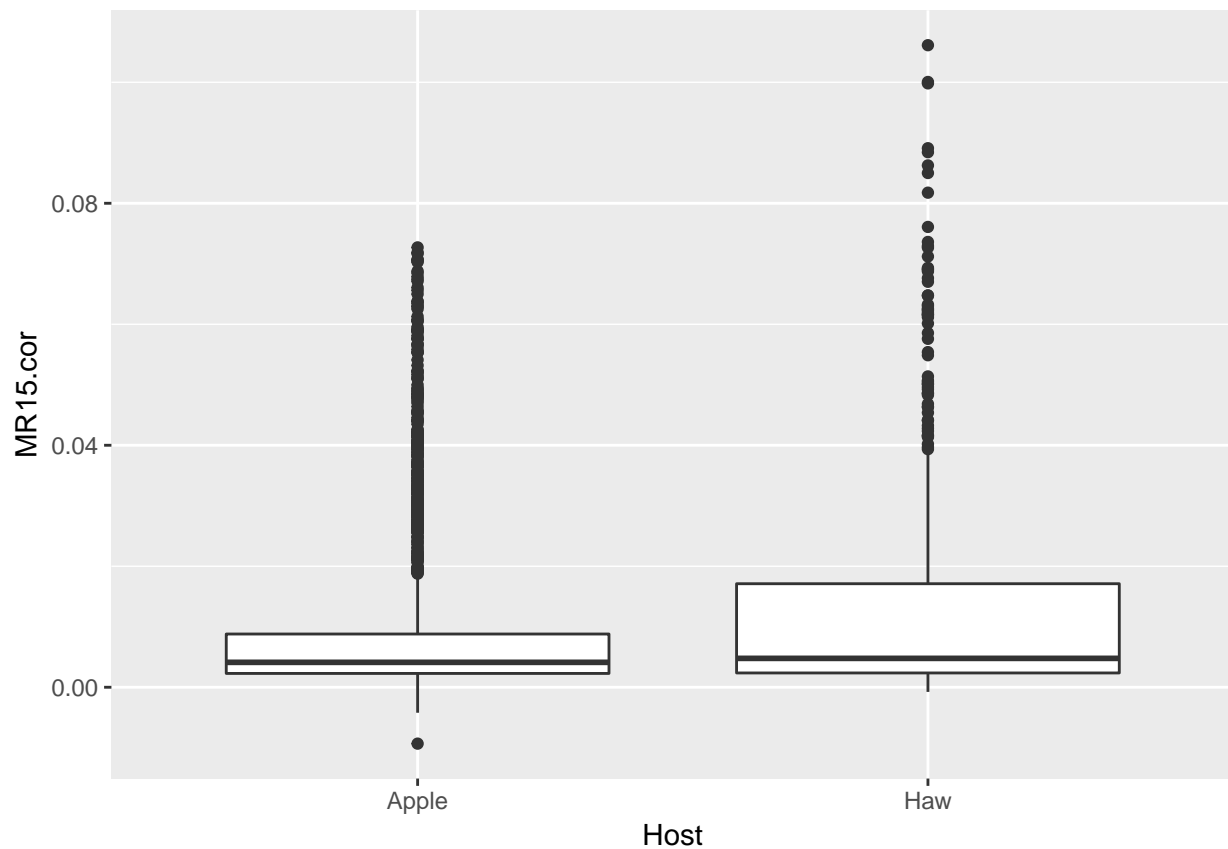
Figures looking at MR between hosts

```

ggplot(data5.15, aes(x=Host, y=MR15.cor))+geom_boxplot()

## Warning: Removed 63 rows containing non-finite values (stat_boxplot).

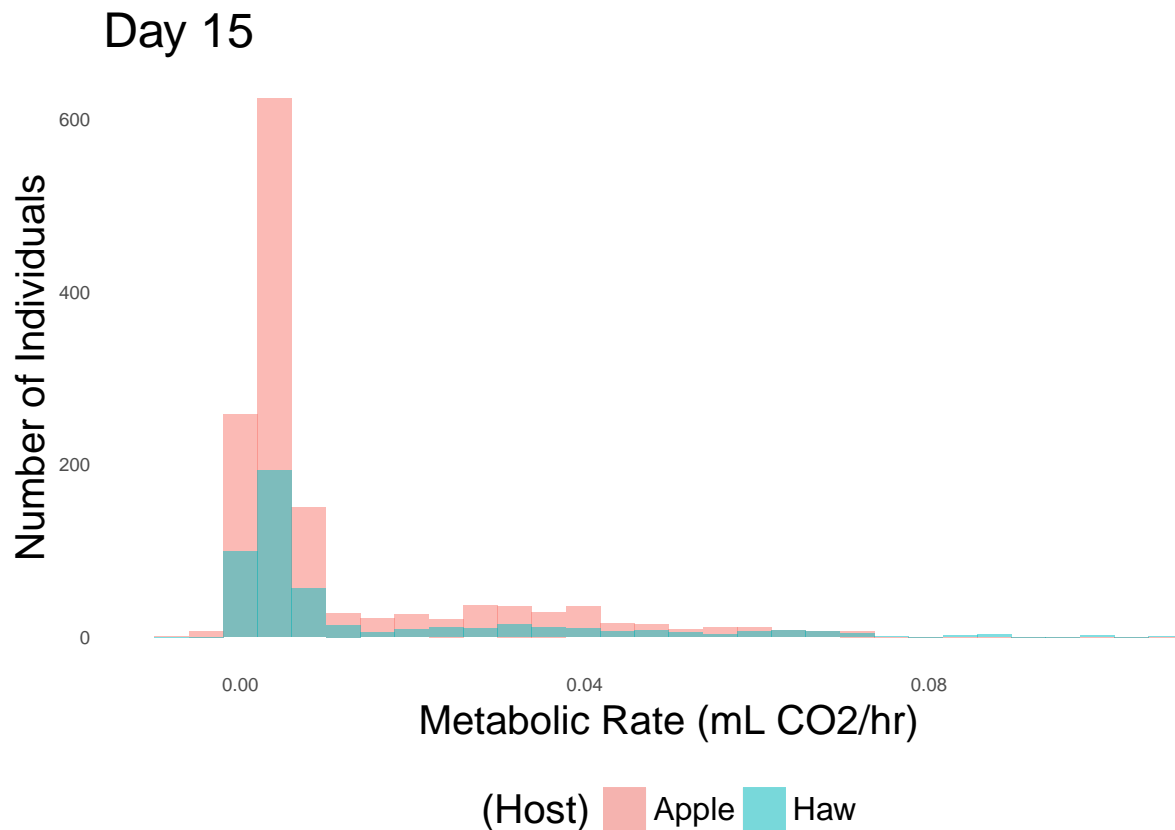
```



```
ggplot(data5.15, aes(x=MR15.cor, fill=(Host)))+geom_histogram(position = "identity", alpha=.5) +labs(x = "MR15.cor",
axis.ticks.x=element_blank(),legend.position="bottom",
axis.ticks.y=element_blank(),panel.background = element_blank(),
panel.grid.major = element_blank(), panel.grid.minor = element_blank(),axis.text=element_text(size=12))
```

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```

```
## Warning: Removed 63 rows containing non-finite values (stat_bin).
```



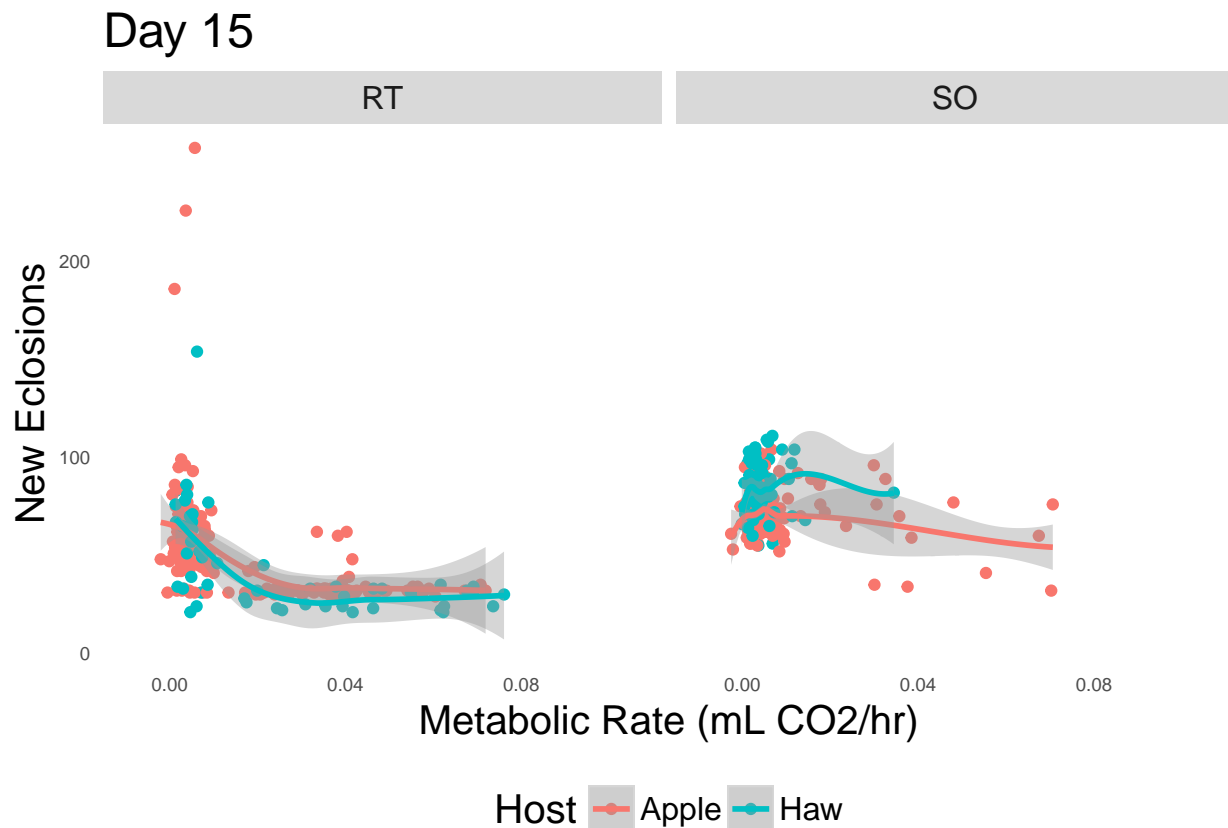
Explore relationship between eclosion days and MR

scatter plots

```
#Filter out GC and blank treatments
data5.15.treatsub<-data5.15%>%
  filter(treatment!="GC"&treatment!="")
```

```
ggplot(data5.15.treatsub, aes(x=MR15.cor, y=neweclosions, colour=Host))+geom_point()+stat_smooth(method="lm",
  axis.ticks.x=element_blank(),legend.position="bottom",
  axis.ticks.y=element_blank(),panel.background = element_blank(),
  panel.grid.major = element_blank(), panel.grid.minor = element_blank(),axis.text=element_text(size=12))
```

```
## Don't know how to automatically pick scale for object of type difftime. Defaulting to continuous.
## Warning: Removed 607 rows containing non-finite values (stat_smooth).
## Warning: Removed 607 rows containing missing values (geom_point).
```



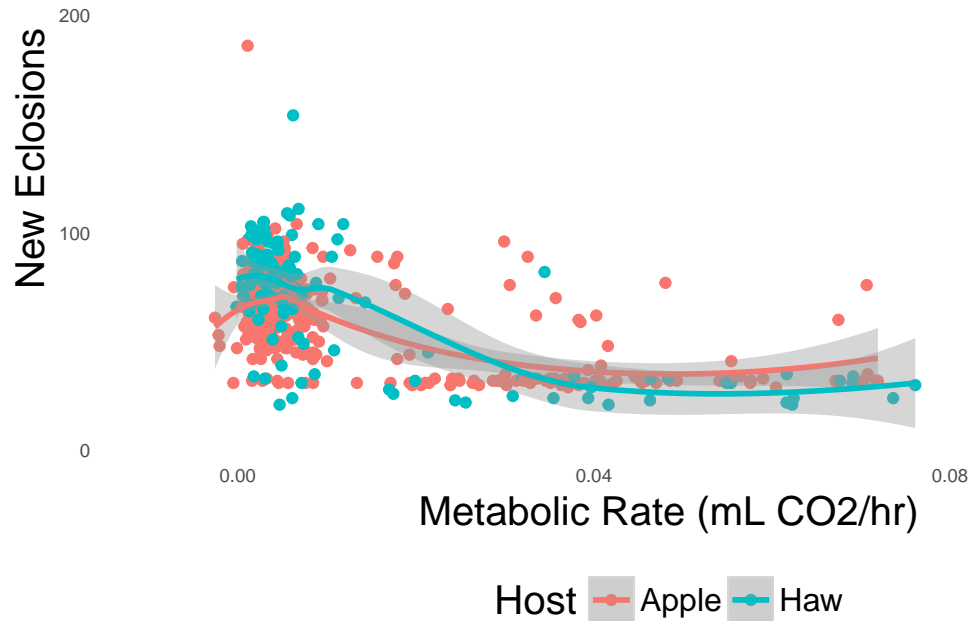
```
#plotting mr15 on the x axis and new eclosions on the y axis; color points by host, fit curve to data
ggplot(data5.15.treatsub,aes(x=MR15.cor, y=neweclosions, colour=Host))+geom_point()+stat_smooth(method=
  axis.ticks.x=element_blank(),legend.position="bottom",
  axis.ticks.y=element_blank(),panel.background = element_blank(),
  panel.grid.major = element_blank(), panel.grid.minor = element_blank(),axis.text=element_text(s
```

```
## Don't know how to automatically pick scale for object of type difftime. Defaulting to continuous.
```

```
## Warning: Removed 607 rows containing non-finite values (stat_smooth).
```

```
## Warning: Removed 607 rows containing missing values (geom_point).
```

Day 15



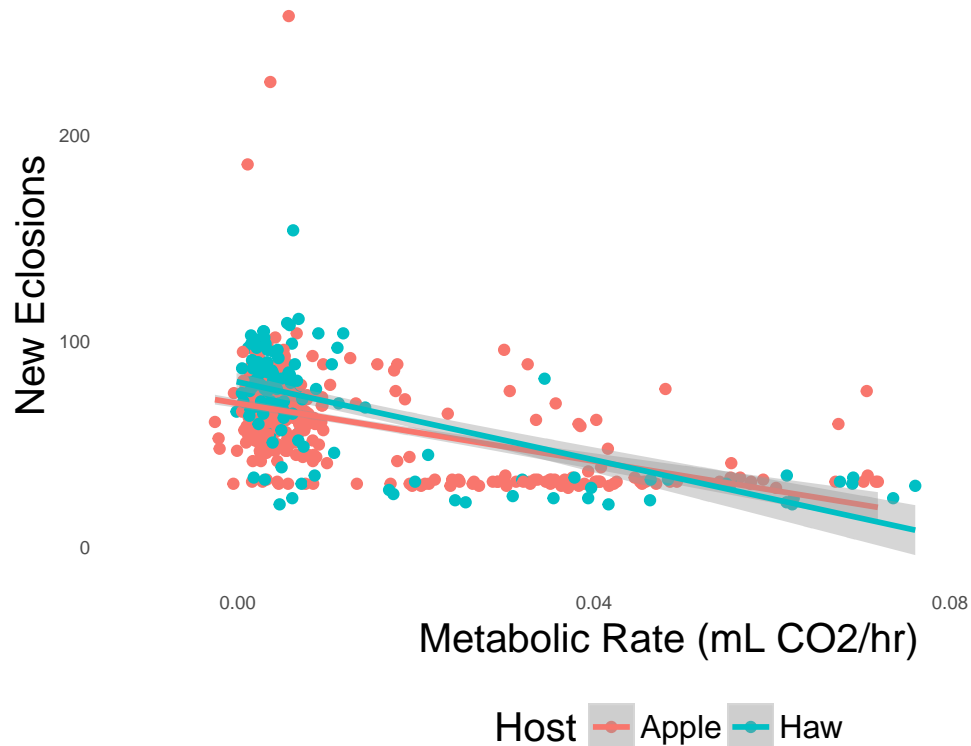
```
#same - fit linear curve to data
ggplot(data5.15.treatsub,aes(x=MR15.cor, y=neweclosions, colour=Host))+geom_point()+stat_smooth(method=
  axis.ticks.x=element_blank(),legend.position="bottom",
  axis.ticks.y=element_blank(),panel.background = element_blank(),
  panel.grid.major = element_blank(), panel.grid.minor = element_blank(),axis.text=element_text(s
```

```
## Don't know how to automatically pick scale for object of type difftime. Defaulting to continuous.
```

```
## Warning: Removed 607 rows containing non-finite values (stat_smooth).
```

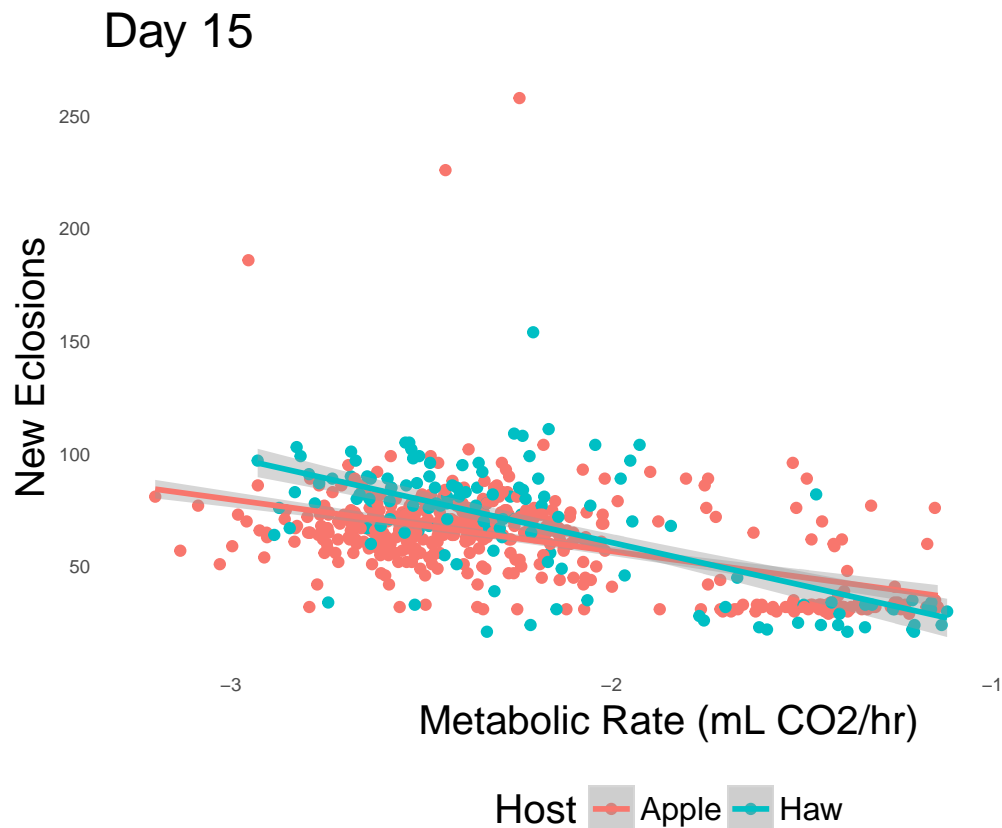
```
## Warning: Removed 607 rows containing missing values (geom_point).
```

Day 15



```
#could log transform mr to make it more linear
ggplot(data5.15.treatsub,aes(x=log10(MR15.cor), y=neweclosions, colour=Host))+geom_point()+stat_smooth(
  axis.ticks.x=element_blank(),legend.position="bottom",
  axis.ticks.y=element_blank(),panel.background = element_blank(),
  panel.grid.major = element_blank(), panel.grid.minor = element_blank(),axis.text=element_text(s
```

```
## Don't know how to automatically pick scale for object of type difftime. Defaulting to continuous.
## Warning in FUN(X[[i]], ...): NaNs produced
## Warning in FUN(X[[i]], ...): NaNs produced
## Warning: Removed 619 rows containing non-finite values (stat_smooth).
## Warning: Removed 619 rows containing missing values (geom_point).
```



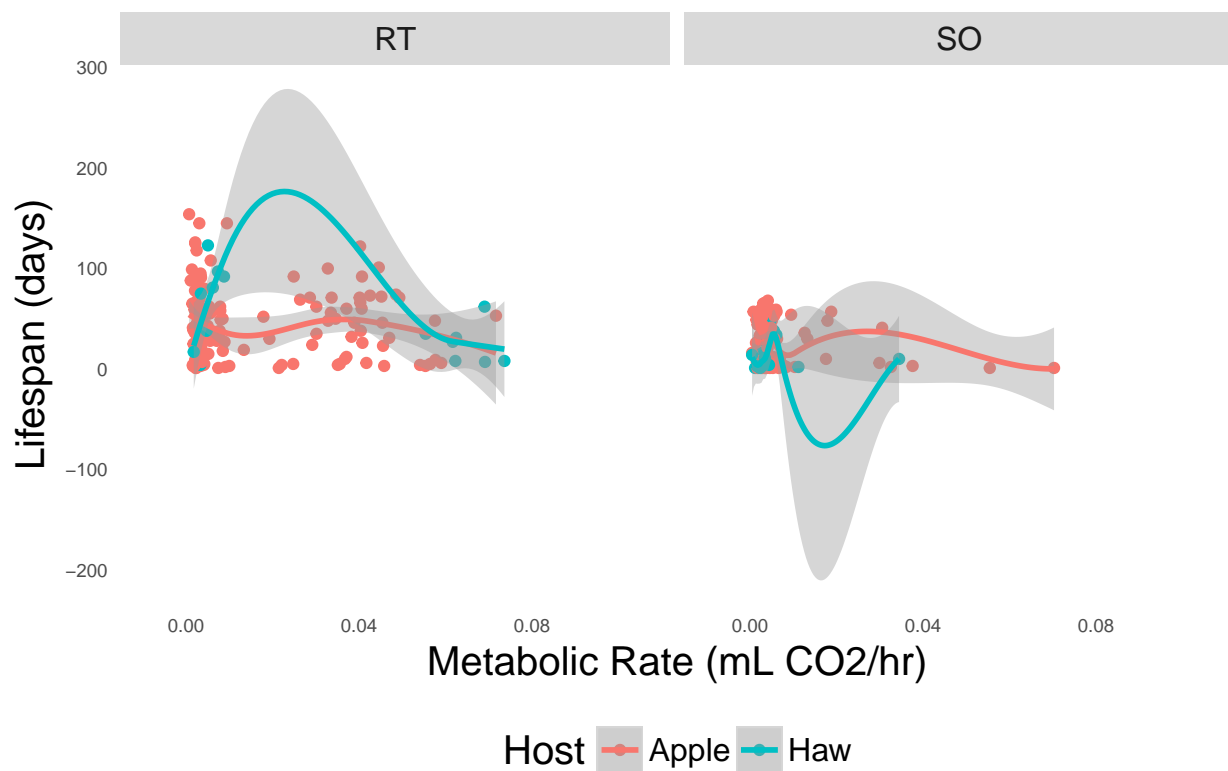
Figures associating MR with lifespan

```
#mr15 w lifespan
data5.15.treatsub$lifespan<-difftime(as.Date(data5.15.treatsub$Adult_death_date, na.rm=TRUE), as.Date(d

ggplot(data5.15.treatsub,aes(x= MR15.cor, y=lifespan, colour=Host))+geom_point()+stat_smooth(method="lo
  axis.ticks.x=element_blank(),legend.position="bottom",
  axis.ticks.y=element_blank(),panel.background = element_blank(),
  panel.grid.major = element_blank(), panel.grid.minor = element_blank(),axis.text=element_text(s

## Don't know how to automatically pick scale for object of type difftime. Defaulting to continuous.
## Warning: Removed 928 rows containing non-finite values (stat_smooth).
## Warning: Removed 928 rows containing missing values (geom_point).
```

Day 15

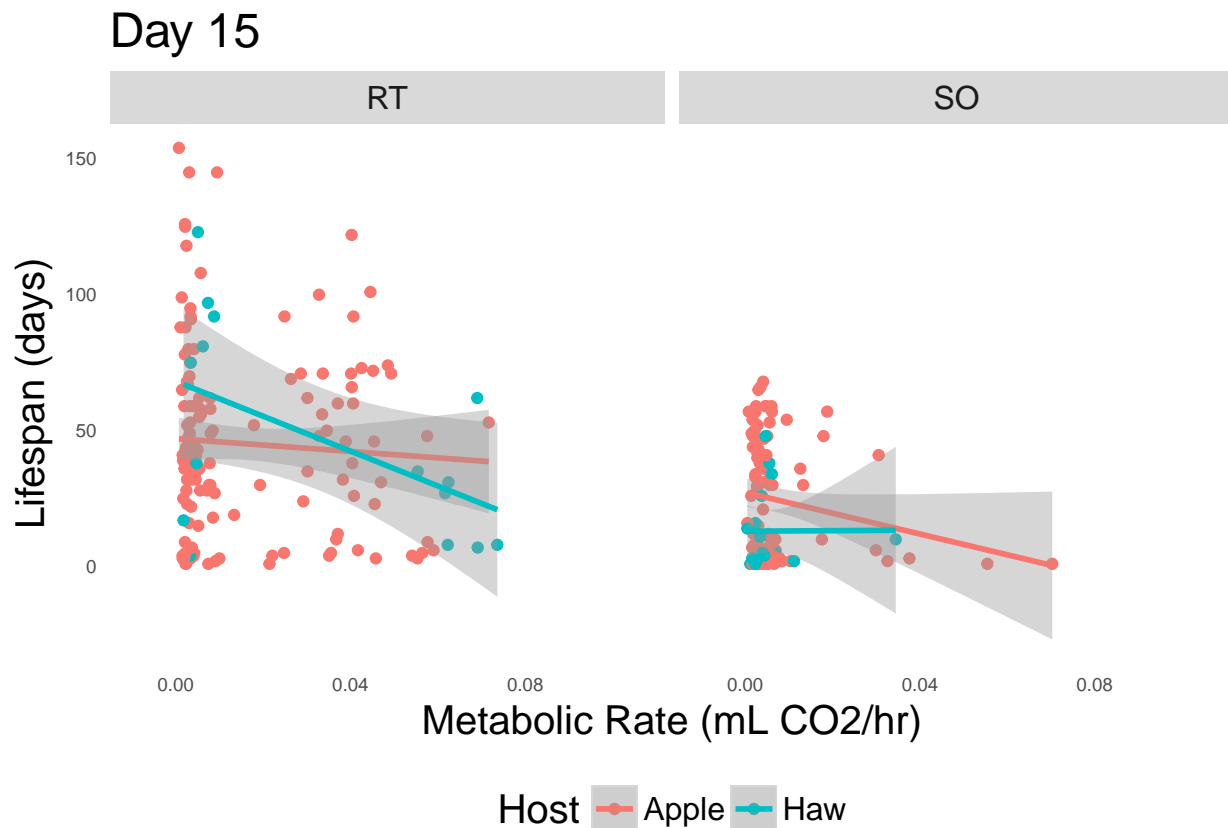


```
ggplot(data5.15.treatsub,aes(x=MR15.cor, y=lifespan, colour=Host))+geom_point()+stat_smooth(method="lm",
axis.ticks.x=element_blank(),legend.position="bottom",
axis.ticks.y=element_blank(),panel.background = element_blank(),
panel.grid.major = element_blank(), panel.grid.minor = element_blank(),axis.text=element_text(s
```

```
## Don't know how to automatically pick scale for object of type difftime. Defaulting to continuous.
```

```
## Warning: Removed 928 rows containing non-finite values (stat_smooth).
```

```
## Warning: Removed 928 rows containing missing values (geom_point).
```

Double Checking Eclosion for Pupal Deaths

```
#Assign names to each column and column bind in order to simplify and find data easily
host <- data[,11]
cd <- data[,10]
wid <- data[,29]
unid<-data[,26]
edate <- data[,27]
dcheck<-cbind(host, cd, wid, unid, edate)
```

Merging pupal death data sheet lifespans

Data available on github

```
pupdeaths<- fread("https://raw.githubusercontent.com/adnguyen/Circadian_rhythm_runs_seasonal_timing/master/pupdeaths.csv")
```

```
#Filter out non-eclosers
data5.na <- data5.15%>%
  mutate(neweclosions=as.numeric(neweclosions))%>%
  #glimpse()
  filter(is.na(neweclosions))

#Filter out individuals with adult lifespans
data5.ls <- data5.15.treatsub%>%
  filter(!is.na(lifespan))
```

```
#Merge the two sets of data (lifespan and non-eclosers) by "Host", "cohort_day", "well_id", and "uniqueID"
mergedat1 <- inner_join(pupdeaths, data5.ls, by = c("Host", "cohort_day", "well_id", "uniqueID"))
```

```
#column bind the merged data with the lifespan data to check same column dim
cbind(names(data5.ls), names(mergedat1)[-56])
```

```
## Warning in cbind(names(data5.ls), names(mergedat1)[-56]): number of rows of
## result is not a multiple of vector length (arg 2)
```

| ## | [,1] | [,2] |
|----|-----------------------------------|-----------------------------|
| ## | [1,] "Ind_ID" | "Host" |
| ## | [2,] "tape" | "cohort_day" |
| ## | [3,] "Site_name" | "well_id" |
| ## | [4,] "mass_day10" | "uniqueID" |
| ## | [5,] "purge_time_1" | "lifespan" |
| ## | [6,] "purge1" | "Ind_ID" |
| ## | [7,] "collection_date" | "tape" |
| ## | [8,] "day10" | "Site_name" |
| ## | [9,] "cohort_date" | "mass_day10" |
| ## | [10,] "cohort_day" | "purge_time_1" |
| ## | [11,] "Host" | "purge1" |
| ## | [12,] "Li-cor_1" | "collection_date" |
| ## | [13,] "resp_time_1" | "day10" |
| ## | [14,] "resp_day11" | "cohort_date" |
| ## | [15,] "mass_day14" | "Li-cor_1" |
| ## | [16,] "purge_time_2" | "resp_time_1" |
| ## | [17,] "resp_time_2" | "resp_day11" |
| ## | [18,] "resp_day15" | "mass_day14" |
| ## | [19,] "Li_cor2" | "purge_time_2" |
| ## | [20,] "treatment_day15" | "resp_time_2" |
| ## | [21,] "exit_fridge_date" | "resp_day15" |
| ## | [22,] "Eclosion_reference_date" | "Li_cor2" |
| ## | [23,] "notes" | "treatment_day15" |
| ## | [24,] "Resp_code" | "exit_fridge_date" |
| ## | [25,] "treatment" | "Eclosion_reference_date" |
| ## | [26,] "uniqueID" | "notes" |
| ## | [27,] "eclosion_date" | "Resp_code" |
| ## | [28,] "eclosion_days" | "treatment" |
| ## | [29,] "well_id" | "eclosion_date" |
| ## | [30,] "organism" | "eclosion_days" |
| ## | [31,] "Trikinetics_position" | "organism" |
| ## | [32,] "Trik_monitor" | "Trikinetics_position" |
| ## | [33,] "Trikinetics_entry_LD_time" | "Trik_monitor" |
| ## | [34,] "Trikinetic_exit_date" | "Trikinetics_entry_LD_time" |
| ## | [35,] "Trikinetics_exit_LD_time" | "Trikinetic_exit_date" |
| ## | [36,] "notes_2" | "Trikinetics_exit_LD_time" |
| ## | [37,] "Free_run_trik_monitor" | "notes_2" |
| ## | [38,] "Free_run_trik_position" | "Free_run_trik_monitor" |
| ## | [39,] "Free_run_entry_date" | "Free_run_trik_position" |
| ## | [40,] "Free_run_entry_time" | "Free_run_entry_date" |
| ## | [41,] "Free_run_exit_date" | "Free_run_entry_time" |
| ## | [42,] "Free_run_exit_time" | "Free_run_exit_date" |
| ## | [43,] "notes_3" | "Free_run_exit_time" |
| ## | [44,] "Adult_death_date" | "notes_3" |

```
## [45,] "day10purge"          "Adult_death_date"
## [46,] "day15purge"          "day10purge"
## [47,] "day15purge.trans"    "day15purge"
## [48,] "day15resp"           "day15purge.trans"
## [49,] "total_time_day15"    "day15resp"
## [50,] "MR15"                "total_time_day15"
## [51,] "msMR15"              "MR15"
## [52,] "mean.blank2"         "msMR15"
## [53,] "MR15.cor"            "mean.blank2"
## [54,] "msMR15.cor"          "MR15.cor"
## [55,] "neweclosures"        "msMR15.cor"
## [56,] "lifespan"            "Host"

#once the column dimensions are the same, merge the two sets of data
mergedata<- merge(data5.ls, mergedat1)

#Finally, row bind the data with the desired conditions
finalmerge<- rbind(data5.ls[,c("Host", "cohort_day", "well_id", "uniqueID", "lifespan", "MR15.cor", "msMR15.cor", "neweclosures", "lifespan")], mergedata)

#glimpse(finalmerge)
```

Negative binomial regression reanalyzed with pupal death data

```
#Filtering done to separate the two treatments
RT15.n <- finalmerge%>%
  filter(treatment=="RT")

S015.n <- finalmerge%>%
  filter(treatment=="S0")

#RT summary (no stat significance)
RT15.n$lifespan <- as.numeric(RT15.n$lifespan)
mod3.n<- glm.nb(lifespan~MR15.cor*Host, data=RT15.n)
summary(mod3.n)

##
## Call:
## glm.nb(formula = lifespan ~ MR15.cor * Host, data = RT15.n, init.theta = 0.4064870624,
##       link = log)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -1.9351  -1.3934  -0.1234   0.3376   1.2813
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)      3.5189     0.1575  22.345  <2e-16 ***
## MR15.cor          3.6204     7.0097   0.516   0.606
## HostHaw          -0.1046     0.4437  -0.236   0.814
## MR15.cor:HostHaw -7.0638    13.6632  -0.517   0.605
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
```

```
## (Dispersion parameter for Negative Binomial(0.4065) family taken to be 1)
##
##      Null deviance: 231.07  on 195  degrees of freedom
## Residual deviance: 230.28  on 192  degrees of freedom
## AIC: 1705
##
## Number of Fisher Scoring iterations: 1
##
##
##              Theta:  0.4065
##            Std. Err.:  0.0419
##
## 2 x log-likelihood: -1694.9600
```

```
#summary : HostHaw (in reference to apple) - as lifespan increases, the MR decreases by the estimate (l
#Coefficient describes relationship, i.e. for mod4SO, higher mass=higher life span according to this mo
#mod4 shows interaction b/w host and mr
mod4.nRT<-glm.nb(lifespan~MR15.cor*Host + mass_day14, data=RT15.n)
summary(mod4.nRT)
```

```
##
## Call:
## glm.nb(formula = lifespan ~ MR15.cor * Host + mass_day14, data = RT15.n,
##       init.theta = 0.4249870689, link = log)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -2.0723  -1.3914  -0.2088   0.3326   1.4933
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)      2.1633     0.3769   5.740 9.49e-09 ***
## MR15.cor          1.4843     7.0176   0.212 0.832488
## HostHaw          -0.4048     0.4368  -0.927 0.353964
## mass_day14        0.2047     0.0553   3.702 0.000214 ***
## MR15.cor:HostHaw -3.2856    13.4285  -0.245 0.806711
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for Negative Binomial(0.425) family taken to be 1)
##
##      Null deviance: 239.80  on 195  degrees of freedom
## Residual deviance: 230.27  on 191  degrees of freedom
## AIC: 1698.5
##
## Number of Fisher Scoring iterations: 1
##
##
##              Theta:  0.4250
##            Std. Err.:  0.0442
##
## 2 x log-likelihood: -1686.4600
```

```
S015.n$lifespan <- as.numeric(S015.n$lifespan)
mod4.nSO<-glm.nb(lifespan~MR15.cor*Host + mass_day14, data=S015.n)
```

```
summary(mod4.nS0)
```

```
##
## Call:
## glm.nb(formula = lifespan ~ MR15.cor * Host + mass_day14, data = S015.n,
##       init.theta = 0.8729265479, link = log)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -2.2740  -1.2062  -0.2847   0.5513   1.6254
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)      2.55542    0.42475   6.016 1.78e-09 ***
## MR15.cor        -30.63664   11.28046  -2.716  0.00661 **
## HostHaw         -1.13706    0.37668  -3.019  0.00254 **
## mass_day14         0.11472    0.05573   2.058  0.03956 *
## MR15.cor:HostHaw  33.18788   37.14903   0.893  0.37166
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for Negative Binomial(0.8729) family taken to be 1)
##
##      Null deviance: 139.26  on 111  degrees of freedom
## Residual deviance: 126.36  on 107  degrees of freedom
## (1 observation deleted due to missingness)
## AIC: 923.59
##
## Number of Fisher Scoring iterations: 1
##
##              Theta:  0.873
##             Std. Err.:  0.112
##
## 2 x log-likelihood:  -911.594
```

```
figmod4.nS0<-glm.nb(lifespan~MR15.cor*Host, data=S015.n)
summary(figmod4.nS0)
```

```
##
## Call:
## glm.nb(formula = lifespan ~ MR15.cor * Host, data = S015.n, init.theta = 0.8485587795,
##       link = log)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -2.3936  -1.2487  -0.4853   0.5522   1.5282
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)      3.3912    0.1375  24.659 < 2e-16 ***
## MR15.cor        -32.9921   11.4113  -2.891  0.00384 **
## HostHaw         -0.8280    0.3668  -2.257  0.02400 *
## MR15.cor:HostHaw  34.0782   37.5036   0.909  0.36353
```

```

## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for Negative Binomial(0.8486) family taken to be 1)
##
##      Null deviance: 135.74  on 111  degrees of freedom
## Residual deviance: 126.80  on 108  degrees of freedom
##   (1 observation deleted due to missingness)
## AIC: 925.26
##
## Number of Fisher Scoring iterations: 1
##
##
##           Theta:  0.849
##          Std. Err.:  0.108
##
## 2 x log-likelihood:  -915.265
figmod4.nRT<-glm.nb(lifespan~MR15.cor*Host, data=RT15.n)
summary(figmod4.nRT)

##
## Call:
## glm.nb(formula = lifespan ~ MR15.cor * Host, data = RT15.n, init.theta = 0.4064870624,
##        link = log)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -1.9351  -1.3934  -0.1234   0.3376   1.2813
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)      3.5189     0.1575  22.345  <2e-16 ***
## MR15.cor          3.6204     7.0097   0.516    0.606
## HostHaw          -0.1046     0.4437  -0.236    0.814
## MR15.cor:HostHaw -7.0638    13.6632  -0.517    0.605
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for Negative Binomial(0.4065) family taken to be 1)
##
##      Null deviance: 231.07  on 195  degrees of freedom
## Residual deviance: 230.28  on 192  degrees of freedom
## AIC: 1705
##
## Number of Fisher Scoring iterations: 1
##
##
##           Theta:  0.4065
##          Std. Err.:  0.0419
##
## 2 x log-likelihood:  -1694.9600
#mod4.1 shows interaction b/w host and msmr
mod4.1.nSO<-glm.nb(lifespan~msMR15.cor*Host, data=S015.n)

```

```
summary(mod4.1.nS0)
```

```
##
## Call:
## glm.nb(formula = lifespan ~ msMR15.cor * Host, data = S015.n,
##       init.theta = 0.848594471, link = log)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -2.3703  -1.2461  -0.3624   0.6299   1.5013
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)      3.2505     0.1172  27.734 < 2e-16 ***
## msMR15.cor     -182.2773    62.8655  -2.899  0.00374 **
## HostHaw        -0.6846     0.2971  -2.304  0.02123 *
## msMR15.cor:HostHaw 224.5755   356.0600   0.631  0.52822
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for Negative Binomial(0.8486) family taken to be 1)
##
##      Null deviance: 135.75  on 111  degrees of freedom
## Residual deviance: 126.85  on 108  degrees of freedom
## (1 observation deleted due to missingness)
## AIC: 925.31
##
## Number of Fisher Scoring iterations: 1
##
##              Theta:  0.849
##             Std. Err.: 0.108
##
## 2 x log-likelihood:  -915.308
```

```
mod4.1.nRT<-glm.nb(lifespan~msMR15.cor*Host, data=RT15.n)
summary(mod4.1.nRT)
```

```
##
## Call:
## glm.nb(formula = lifespan ~ msMR15.cor * Host, data = RT15.n,
##       init.theta = 0.4071281808, link = log)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -1.9507  -1.3966  -0.1284   0.3380   1.2736
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)      3.53929    0.13867  25.524 <2e-16 ***
## msMR15.cor      21.20632    44.94809   0.472   0.637
## HostHaw        -0.08528    0.39366  -0.217   0.828
## msMR15.cor:HostHaw -67.07964    78.15514  -0.858   0.391
## ---
```

```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for Negative Binomial(0.4071) family taken to be 1)
##
##      Null deviance: 231.38  on 195  degrees of freedom
## Residual deviance: 230.28  on 192  degrees of freedom
## AIC: 1704.7
##
## Number of Fisher Scoring iterations: 1
##
##              Theta:  0.4071
##             Std. Err.:  0.0420
##
## 2 x log-likelihood:  -1694.6580
#Likelihood ratio tests of Negative Binomial Models
m1.nRT <- update(mod4.nRT, . ~ . - prog)
anova(mod4.nRT, m1.nRT)

## Likelihood ratio tests of Negative Binomial Models
##
## Response: lifespan
##
##              Model      theta Resid. df
## 1              MR15.cor * Host + mass_day14 0.4249871      191
## 2 MR15.cor + Host + mass_day14 + MR15.cor:Host 0.4249871      191
##      2 x log-lik.   Test      df      LR stat. Pr(Chi)
## 1              -1686.46
## 2              -1686.46 1 vs 2      0 -3.760033e-08      1
#Checking model assumption -- is negative binomial regression (NBR) a good model for this data?
##Values close to 0 (estimates the dispersion parameter) strongly suggest the NBR model is more appropriate
mod5.n <- glm(lifespan ~ Host*MR15.cor, family = "poisson", data = S015.n)
pchisq(2 * (logLik(mod4.nRT) - logLik(mod5.n)), df = 1, lower.tail = FALSE)

## 'log Lik.' 6.470061e-238 (df=6)
mod5.n <- glm(lifespan ~ Host*MR15.cor, family = "poisson", data = S015.n)
pchisq(2 * (logLik(mod4.nS0) - logLik(mod5.n)), df = 1, lower.tail = FALSE)

## 'log Lik.' 0 (df=6)
#output below indicates that Hawthorn is 0.432 times less likely to live than apple
#confidence interval
(est <- cbind(Estimate = coef(mod4.nRT), confint(mod4.nRT)))

## Waiting for profiling to be done...
##
##              Estimate      2.5 %      97.5 %
## (Intercept)    2.1633428  1.2237554  3.1119932
## MR15.cor        1.4842992 -11.7040163  15.8181228
## HostHaw        -0.4048505  -1.2336952   0.5650535
## mass_day14      0.2047244   0.0676875   0.3465155
## MR15.cor:HostHaw -3.2855872 -29.3106300  25.7786008
#incident rate ratio - lifespan ratio
exp(est)
```



```
##               Estimate      2.5 %      97.5 %
## (Intercept)    8.70017236 3.399932e+00 2.246578e+01
## MR15.cor       4.41187261 8.260576e-06 7.408384e+06
## HostHaw       0.66707656 2.912145e-01 1.759542e+00
## mass_day14     1.22718686 1.070031e+00 1.414131e+00
## MR15.cor:HostHaw 0.03741861 1.864469e-13 1.568571e+11

#Filter out all the NAs in order to have same dimensions for predvalues
filS015.ms<- S015.n%>%
  filter(msMR15.cor!= "NA"&lifespan !="NA")

filRT15.ms<- RT15.n%>%
  filter(msMR15.cor!= "NA"&lifespan !="NA")

#Calculate the predicted values for msMR15
filS015.ms$predvalues <- predict(mod4.1.nS0, type = "response")

filRT15.ms$predvalues <- predict(mod4.1.nRT, type = "response")

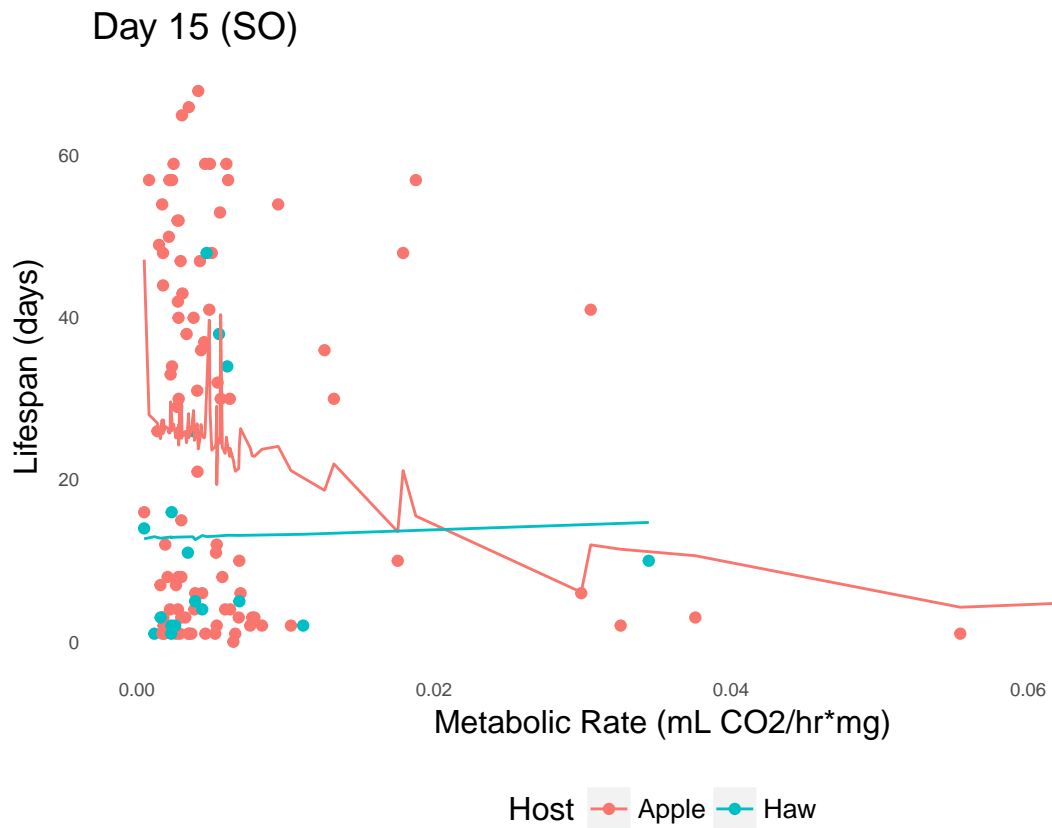
#Filter out all the NAs in order to have same dimensions for predvalues
filS015.ms<- S015.n%>%
  filter(msMR15.cor!= "NA"&lifespan !="NA")

filRT15.ms<- RT15.n%>%
  filter(msMR15.cor!= "NA"&lifespan !="NA")

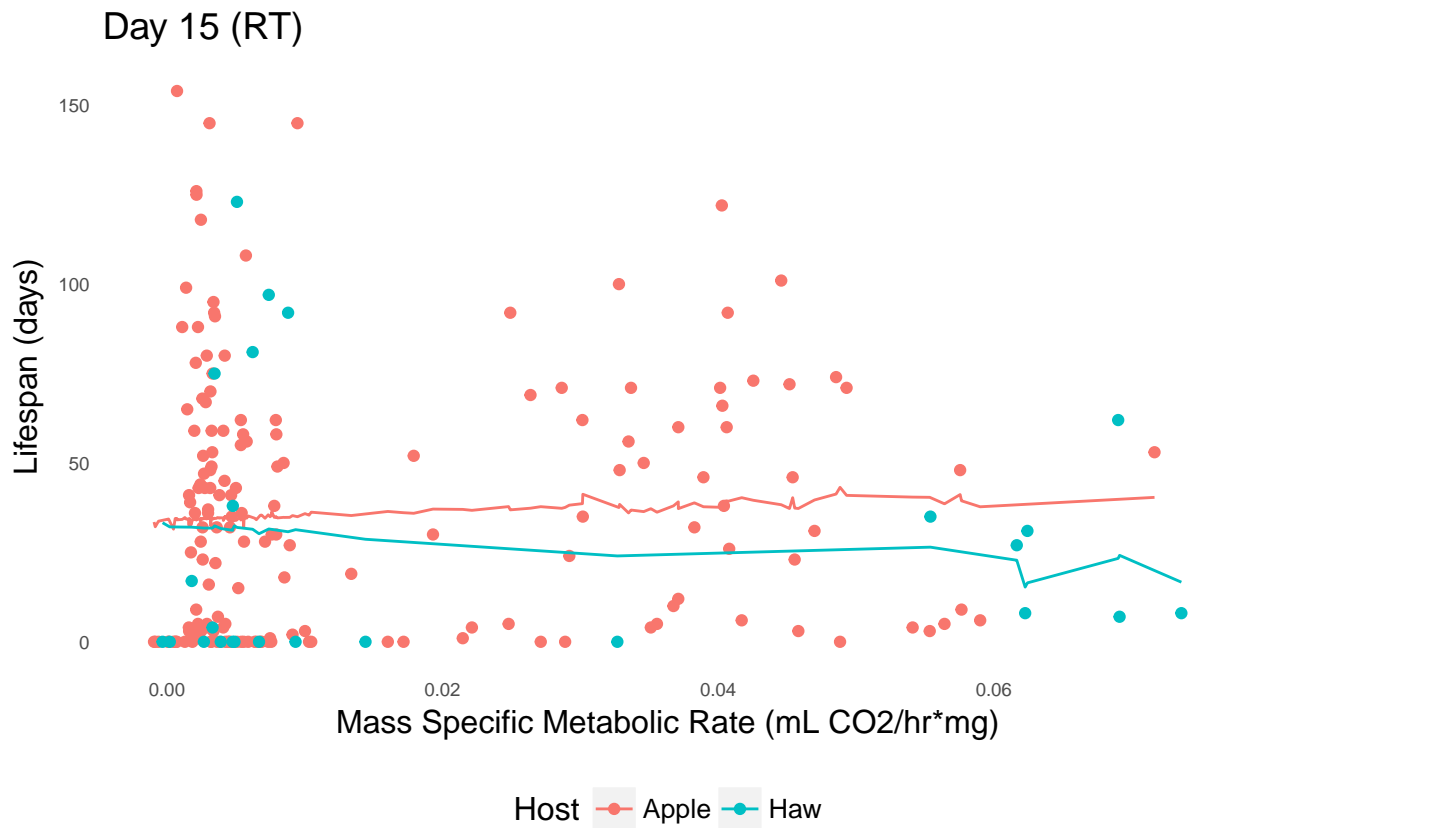
#Calculate the predicted values
filS015.ms$predvalues <- predict(mod4.1.nS0, type = "response")

filRT15.ms$predvalues <- predict(mod4.1.nRT, type = "response")

#overlay of real data(points) with predicted values(line) for MR
ggplot(filS015.ms,aes(x=MR15.cor, y=lifespan, colour=Host))+geom_point()+geom_line(data=filS015.ms, aes
  axis.ticks.x=element_blank(),legend.position="bottom",
  axis.ticks.y=element_blank(),panel.background = element_blank(),
  panel.grid.major = element_blank(), panel.grid.minor = element_blank(),axis.text=element_text(s
```

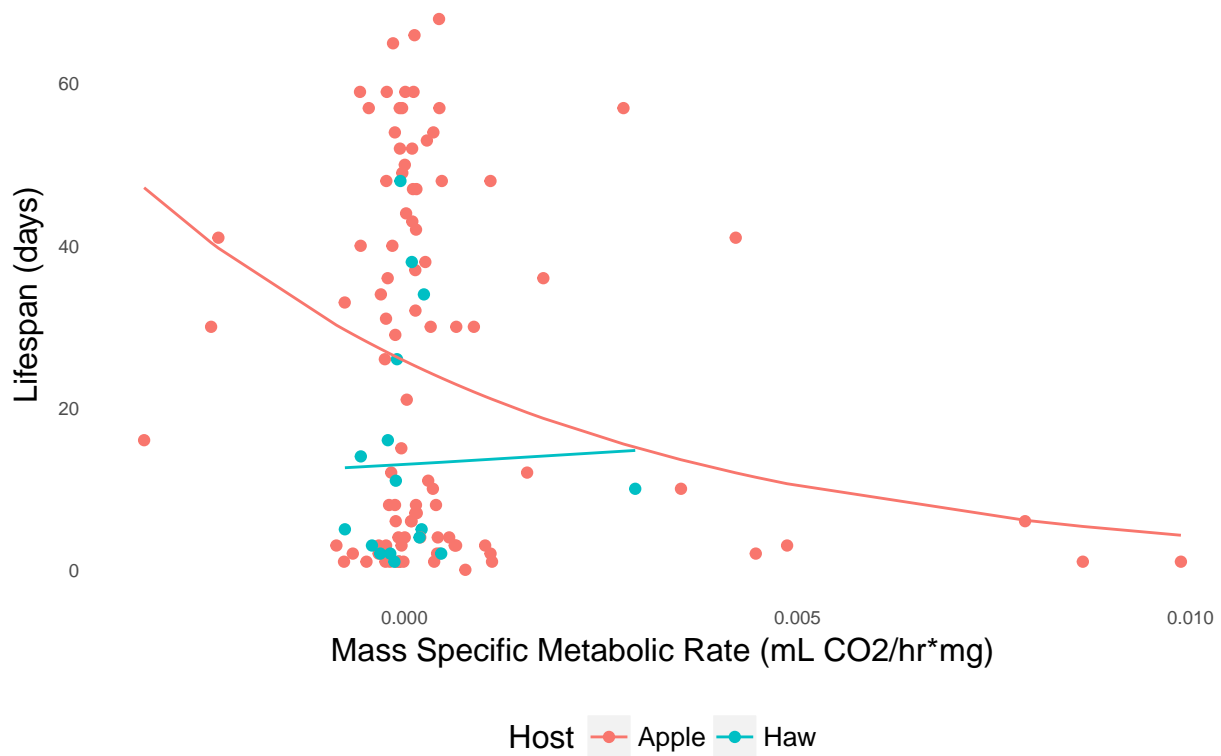


```
ggplot(filRT15.ms,aes(x=MR15.cor, y=lifespan, colour=Host))+geom_point()+geom_line(data=filRT15.ms, aes
axis.ticks.x=element_blank(),legend.position="bottom",
axis.ticks.y=element_blank(),panel.background = element_blank(),
panel.grid.major = element_blank(), panel.grid.minor = element_blank(),axis.text=element_text(s
```

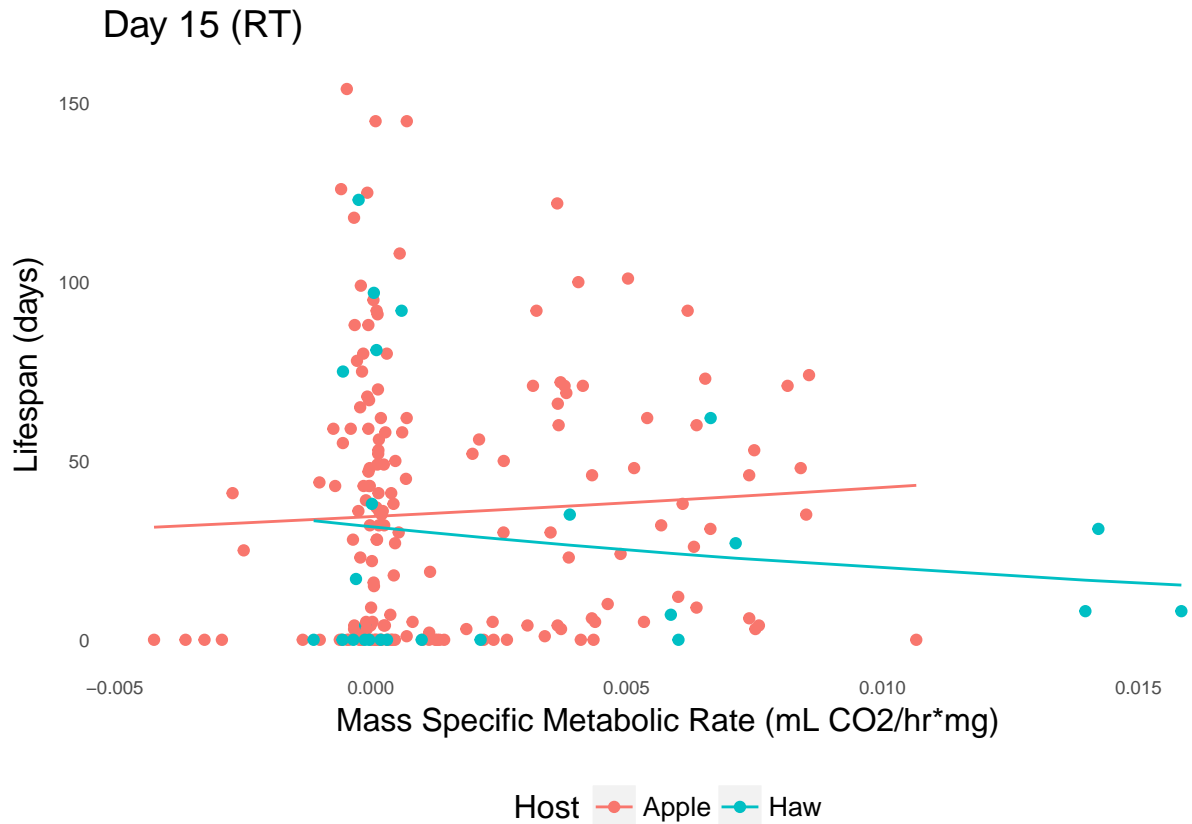


```
#overlay of real data(points) with predicted values(line) for msMR
ggplot(filS015.ms,aes(x=msMR15.cor, y=lifespan, colour=Host))+geom_point()+geom_line(data=filS015.ms, a
axis.ticks.x=element_blank(),legend.position="bottom",
axis.ticks.y=element_blank(),panel.background = element_blank(),
panel.grid.major = element_blank(), panel.grid.minor = element_blank(),axis.text=element_text(s
```

Day 15 (SO)



```
ggplot(filRT15.ms,aes(x=msMR15.cor, y=lifespan, colour=Host))+geom_point()+geom_line(data=filRT15.ms, a
axis.ticks.x=element_blank(),legend.position="bottom",
axis.ticks.y=element_blank(),panel.background = element_blank(),
panel.grid.major = element_blank(), panel.grid.minor = element_blank(),axis.text=element_text(s
```



Sam-
ples were split into two treatments, RT (favorable conditions) and SO (simulated overwintering). The dependent variable was lifespan, while the independent variables were metabolic rate in relation to host and mass. For this model, a negative coefficient of the log value would indicate a shorter lifespan since it produces a log-linear model in which a positive coefficient would indicate a (larger x and y value).

Based on the RT treatment, only mass showed significance which suggests an individual with a larger mass would have a longer lifespan (positive coefficient). Analysis of the SO treatment showed significance in host type, metabolic rate, and mass, in which a lower metabolic rate would suggest longer lifespan (neg. coefficient), and larger mass would suggest longer lifespan (pos. coefficient). Hawthorn individuals also had an overall lower lifespan compared to apple.

Negative binomial regression was shown to be a good model by comparing its log likelihood to a poisson distribution. Values close to 0 (which estimates the dispersion parameter) strongly suggests the NBR model is more appropriate than the Poisson model for both treatments.

Kaplan-Meier Survival Estimate

```
#install.packages(c("survival", "surminer"))
#install.packages(c("ggpubr", "magrittr"))
#Assign censored data and create new column
finalmerge$status[finalmerge$lifespan >= 0] <- 1

## Warning: Unknown or uninitialised column: 'status'.

#filter out unwanted treatments
fmerge <- finalmerge %>%
  filter(treatment != "")
```

```

#Filter out unwanted treatments
finalmerge <- finalmerge %>%
  filter(treatment != "")

#compute kaplan-Meier survival estimate-so compute the survival probability by host & treatment:
#fit1 computes survival probability by Host + Treatment

fit1 <- survfit(Surv(lifespan, status) ~ Host + treatment, data = finalmerge)
print(fit1)

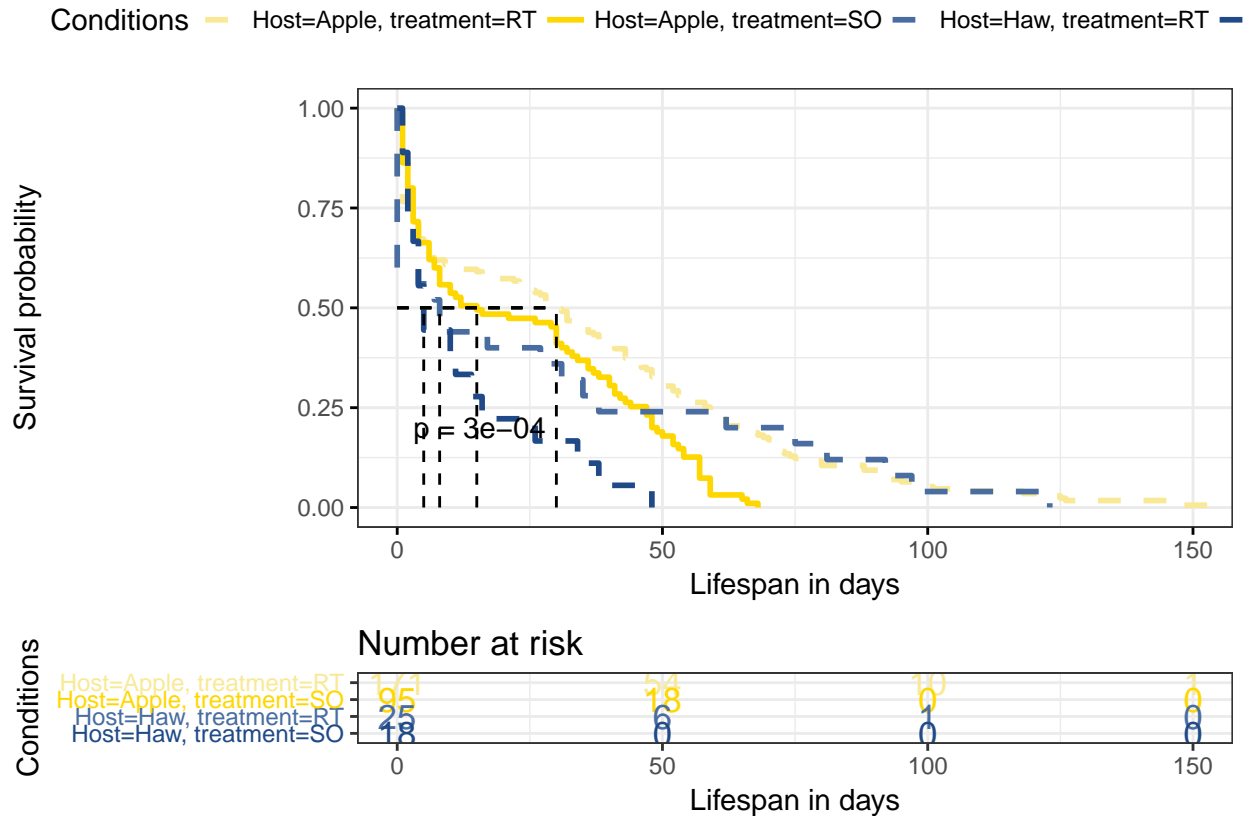
## Call: survfit(formula = Surv(lifespan, status) ~ Host + treatment,
##      data = finalmerge)
##
##              n events median 0.95LCL 0.95UCL
## Host=Apple, treatment=RT 171      171      30      22      39
## Host=Apple, treatment=SO  95       95      15       8      33
## Host=Haw, treatment=RT   25       25       8       0      38
## Host=Haw, treatment=SO   18       18       5       3      26

# Access to the sort summary table
summary(fit1)$table

##              records n.max n.start events      *rmean *se(rmean)
## Host=Apple, treatment=RT    171    171    171    171 33.91520  2.439589
## Host=Apple, treatment=SO    95     95     95    95 24.31579  2.271454
## Host=Haw, treatment=RT     25     25     25    25 27.04000  6.846483
## Host=Haw, treatment=SO     18     18     18    18 12.55556  3.296296
##              median 0.95LCL 0.95UCL
## Host=Apple, treatment=RT    30      22      39
## Host=Apple, treatment=SO    15       8      33
## Host=Haw, treatment=RT      8       0      38
## Host=Haw, treatment=SO      5       3      26

# Visualize curves (Change color, linetype by strata, risk.table color by strata)
ggsurvplot(fit1,
  pval = TRUE, conf.int = FALSE,
  risk.table = TRUE, # Add risk table
  risk.table.col = "strata", # Change risk table color by groups
  linetype = c("dashed","solid", "dashed", "dashed","solid"), # Change line type
  surv.median.line = "hv", # Specify median survival
  ggtheme = theme_bw(), # Change ggplot2 theme
  palette = c("#F9E996","#FFD700","#496DA1","#204A87"),
  xlab = "Lifespan in days",
  legend.title = "Conditions",
  pval.size=4)

```



The Kaplan-Meier Survival Estimate predicts the probability of survival from a specific time, so for this data, the estimates predict the probability of host survival at each time point in their lifespan. According to the figure, Hawthorn flies have an overall lower probability of survival over apple flies, which agrees with the previous analyses. This estimate is best used for categorical values, so MR and mass (which were all unique to the individual) were not good estimators and not used.

Cox Regression

```
#Split up treatments into: SO & RT
fmergeSO <- fmerge%>%
  filter(treatment == "SO")

fmergeRT <- fmerge%>%
  filter(treatment == "RT")

#Merge the data for censoring (all labeled as status 1 for dead)
covariates <- c("host", "treatment", "MR15.cor", "mass_day14")
#For RT
univ_formulasRT <- sapply(covariates,
  function(x) as.formula(paste('Surv(lifespan, status)~', fmergeRT)))
#For SO
univ_formulasSO <- sapply(covariates,
  function(x) as.formula(paste('Surv(lifespan, status)~', fmergeSO)))

univ_modelsRT <- lapply(univ_formulasRT, function(x){coxph(x, data = fmergeRT)})
univ_modelsSO <- lapply(univ_formulasSO, function(x){coxph(x, data = fmergeSO)})
```

```

#Multivariate case
#Testing interaction between Host, Treatment, and MR -- Shows interaction b/w treatment & Host
res.cox <- coxph(Surv(lifespan, status) ~ Host*treatment*MR15.cor + mass_day14, data = fmerge)
summary(res.cox)

## Call:
## coxph(formula = Surv(lifespan, status) ~ Host * treatment * MR15.cor +
##       mass_day14, data = fmerge)
##
##      n= 308, number of events= 308
##      (1 observation deleted due to missingness)
##
##              coef exp(coef) se(coef)      z
## HostHaw          2.178e-01  1.243e+00  2.887e-01  0.754
## treatmentSO       3.257e-01  1.385e+00  1.698e-01  1.918
## MR15.cor          -1.494e+00  2.244e-01  4.562e+00 -0.328
## mass_day14        -1.362e-01  8.727e-01  3.457e-02 -3.940
## HostHaw:treatmentSO  6.938e-01  2.001e+00  4.437e-01  1.564
## HostHaw:MR15.cor     5.393e+00  2.198e+02  8.805e+00  0.612
## treatmentSO:MR15.cor  1.858e+01  1.170e+08  1.163e+01  1.597
## HostHaw:treatmentSO:MR15.cor -2.660e+01  2.804e-12  3.695e+01 -0.720
##
##              Pr(>|z|)
## HostHaw          0.4506
## treatmentSO       0.0551 .
## MR15.cor          0.7432
## mass_day14        8.15e-05 ***
## HostHaw:treatmentSO  0.1179
## HostHaw:MR15.cor     0.5402
## treatmentSO:MR15.cor  0.1102
## HostHaw:treatmentSO:MR15.cor  0.4716
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##              exp(coef) exp(-coef) lower .95 upper .95
## HostHaw          1.243e+00  8.043e-01  7.060e-01  2.190e+00
## treatmentSO       1.385e+00  7.220e-01  9.929e-01  1.932e+00
## MR15.cor          2.244e-01  4.457e+00  2.935e-05  1.715e+03
## mass_day14        8.727e-01  1.146e+00  8.155e-01  9.338e-01
## HostHaw:treatmentSO  2.001e+00  4.997e-01  8.388e-01  4.775e+00
## HostHaw:MR15.cor     2.198e+02  4.550e-03  7.032e-06  6.870e+09
## treatmentSO:MR15.cor  1.170e+08  8.550e-09  1.469e-02  9.314e+17
## HostHaw:treatmentSO:MR15.cor  2.804e-12  3.566e+11  9.964e-44  7.892e+19
##
## Concordance= 0.611 (se = 0.021 )
## Rsquare= 0.107 (max possible= 1 )
## Likelihood ratio test= 34.75 on 8 df, p=3e-05
## Wald test              = 36.24 on 8 df, p=2e-05
## Score (logrank) test = 37.24 on 8 df, p=1e-05

#Coefficient describes relationship: for Cox Regression, coefficient describes relationship to likeliho
#Test interaction between Host, RT, and MR
res.coxRT <- coxph(Surv(lifespan, status) ~ Host*MR15.cor + mass_day14, data = fmergeRT)
summary(res.coxRT)

## Call:

```



```
## coxph(formula = Surv(lifespan, status) ~ Host * MR15.cor + mass_day14,
##       data = fmergeRT)
##
## n= 196, number of events= 196
##
##               coef exp(coef) se(coef)      z Pr(>|z|)
## HostHaw        0.26512   1.30358  0.29147  0.910 0.363049
## MR15.cor       -0.50118   0.60582  4.57571 -0.110 0.912782
## mass_day14     -0.15524   0.85621  0.04451 -3.488 0.000488 ***
## HostHaw:MR15.cor 2.90449  18.25589  8.85836  0.328 0.743002
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##               exp(coef) exp(-coef) lower .95 upper .95
## HostHaw          1.3036    0.76712 7.363e-01 2.308e+00
## MR15.cor          0.6058    1.65067 7.718e-05 4.755e+03
## mass_day14        0.8562    1.16794 7.847e-01 9.343e-01
## HostHaw:MR15.cor  18.2559    0.05478 5.262e-07 6.334e+08
##
## Concordance= 0.624 (se = 0.028 )
## Rsquare= 0.068 (max possible= 1 )
## Likelihood ratio test= 13.77 on 4 df, p=0.008
## Wald test = 13.57 on 4 df, p=0.009
## Score (logrank) test = 13.62 on 4 df, p=0.009
##
#Test interaction between Host, SO, and MR
res.coxSO <- coxph(Surv(lifespan, status) ~ Host*MR15.cor + mass_day14, data = fmergeSO)
summary(res.coxSO)

## Call:
## coxph(formula = Surv(lifespan, status) ~ Host * MR15.cor + mass_day14,
##       data = fmergeSO)
##
## n= 112, number of events= 112
## (1 observation deleted due to missingness)
##
##               coef exp(coef) se(coef)      z Pr(>|z|)
## HostHaw        1.175e+00  3.238e+00  3.826e-01  3.071 0.00213 **
## MR15.cor        3.096e+01  2.780e+13  1.176e+01  2.632 0.00849 **
## mass_day14     -1.032e-01  9.019e-01  5.506e-02 -1.874 0.06088 .
## HostHaw:MR15.cor -3.888e+01  1.305e-17  3.748e+01 -1.037 0.29962
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##               exp(coef) exp(-coef) lower .95 upper .95
## HostHaw        3.238e+00  3.088e-01  1.530e+00  6.853e+00
## MR15.cor        2.780e+13  3.597e-14  2.707e+03  2.855e+23
## mass_day14      9.019e-01  1.109e+00  8.097e-01  1.005e+00
## HostHaw:MR15.cor 1.305e-17  7.663e+16  1.627e-49  1.047e+15
##
## Concordance= 0.593 (se = 0.035 )
## Rsquare= 0.122 (max possible= 0.999 )
## Likelihood ratio test= 14.57 on 4 df, p=0.006
## Wald test = 16.07 on 4 df, p=0.003
## Score (logrank) test = 16.72 on 4 df, p=0.002
```

Samples were split into two treatments, RT (favorable conditions) and SO (simulated overwintering). The dependent variable was lifespan, while the independent variables were metabolic rate in relation to host and mass. For this model, the coefficient relates x variables (metabolic rate, host, and mass) to likelihood of death (hazard rate). The dependent variable is lifespan again.

The RT treatment only showed significance for mass (negative coefficient), in which a lower mass suggests a higher likelihood of death (i.e. longer lifespan).

The SO treatment showed significance for metabolic rate, host, and mass. The positive coefficient for the hawthorn flies and metabolic rate indicate higher likelihood of death if the individual is a host to hawthorn and/or the individual has a high metabolic rate. This agrees with the NBR model. However, the negative coefficient for mass indicates a higher likelihood of death with smaller masses which agrees with the NBR model.

Conclusions

Evolutionarily speaking, linking lower metabolic rate to longer lifespan makes sense since an organism who reserves its energy would take longer to develop, thus longer to die. For example early eclosers tend to have higher metabolic rate vs. late eclosers, so if a group of insects undergo diapause at the same time, those with higher metabolic rates risk eclosing too early in the season, before optimal conditions are available. Late eclosers have a higher energy reserve because they don't need as much energy with lower metabolic rates. This allows them to eclose at suitable seasonal time, lengthening their lifespan/hazard rate.

From this data, the cohorts placed in favorable conditions showed no significance when relating metabolic rate and lifespan. This is most likely because these cohorts had no survival need to undergo diapause which would have separated out the individuals with high and low metabolic rates. Those placed in simulated overwintering conditions would be affected by their metabolic rates – individuals with high metabolic rates would be more likely to eclose earlier and die earlier. Individuals with low metabolic rates would spend more time in diapause and eclose later, extending their lifespan.

According to the evolution theory described in Olshansky & Rattan (2009), higher metabolic rates are linked to species with “high extrinsic mortality” (e.g. targeted prey) since they must quickly develop and reproduce in order to pass down their genes. Lower metabolic rates were linked to individuals with lower extrinsic mortalities and could take their time to gradually develop into larger sizes and live longer. For this theory to hold true, lifespan would need to be considered as a function of metabolic rate.

However, the evolution theory does not consider seasonal timing in this case. It may be possible that both early and late eclosers miss the optimal season for mating by eclosing before or after resource availability. Therefore, individuals who have metabolic rates that are too high or too low would not be able to pass down their genes. However, if we assume these individuals are still able to live, but not reproduce, their lifespans would not be an ideal indicator for fitness.

Session Info

```
sessionInfo()
```

```
## R version 3.5.0 (2018-04-23)
## Platform: x86_64-apple-darwin15.6.0 (64-bit)
## Running under: macOS High Sierra 10.13.5
##
## Matrix products: default
## BLAS: /Library/Frameworks/R.framework/Versions/3.5/Resources/lib/libRblas.0.dylib
## LAPACK: /Library/Frameworks/R.framework/Versions/3.5/Resources/lib/libRlapack.dylib
```

```

##
## locale:
## [1] en_US.UTF-8/en_US.UTF-8/en_US.UTF-8/C/en_US.UTF-8/en_US.UTF-8
##
## attached base packages:
## [1] stats      graphics  grDevices  utils      datasets  methods   base
##
## other attached packages:
## [1] bindrcpp_0.2.2    gridExtra_2.3    survminer_0.4.2
## [4] ggpubr_0.1.6      magrittr_1.5     survival_2.42-3
## [7] BBmisc_1.11       MASS_7.3-49      DiagrammeR_1.0.0
## [10] curl_3.2          lubridate_1.7.4  ggplot2_2.2.1
## [13] data.table_1.11.4 dplyr_0.7.5
##
## loaded via a namespace (and not attached):
## [1] viridis_0.5.1     tidyr_0.8.1      jsonlite_1.5
## [4] viridisLite_0.3.0 splines_3.5.0    assertthat_0.2.0
## [7] yaml_2.1.19       pillar_1.2.3     backports_1.1.2
## [10] lattice_0.20-35   glue_1.2.0       downloader_0.4
## [13] digest_0.6.15     RColorBrewer_1.1-2 checkmate_1.8.5
## [16] colorspace_1.3-2  cmprsk_2.2-7     htmltools_0.3.6
## [19] Matrix_1.2-14     plyr_1.8.4       psych_1.8.4
## [22] XML_3.98-1.11     pkgconfig_2.0.1  broom_0.4.4
## [25] purrr_0.2.5       xtable_1.8-2     scales_0.5.0
## [28] brew_1.0-6        km.ci_0.5-2      KMSurv_0.1-5
## [31] tibble_1.4.2      influenceR_0.1.0 lazyeval_0.2.1
## [34] cli_1.0.0         mnormt_1.5-5     rgexf_0.15.3
## [37] crayon_1.3.4      evaluate_0.10.1  nlme_3.1-137
## [40] foreign_0.8-70    Rook_1.1-1       tools_3.5.0
## [43] hms_0.4.2         stringr_1.3.1    munsell_0.4.3
## [46] compiler_3.5.0    rlang_0.2.0      grid_3.5.0
## [49] rstudioapi_0.7    htmlwidgets_1.2  visNetwork_2.0.3
## [52] igraph_1.2.1      labeling_0.3      rmarkdown_1.9
## [55] gtable_0.2.0      reshape2_1.4.3   R6_2.2.2
## [58] zoo_1.8-1         knitr_1.20       survMisc_0.5.4
## [61] utf8_1.1.4        bindr_0.1.1      rprojroot_1.3-2
## [64] readr_1.1.1       stringi_1.2.2    parallel_3.5.0
## [67] Rcpp_0.12.17      tidyselect_0.2.4

```