

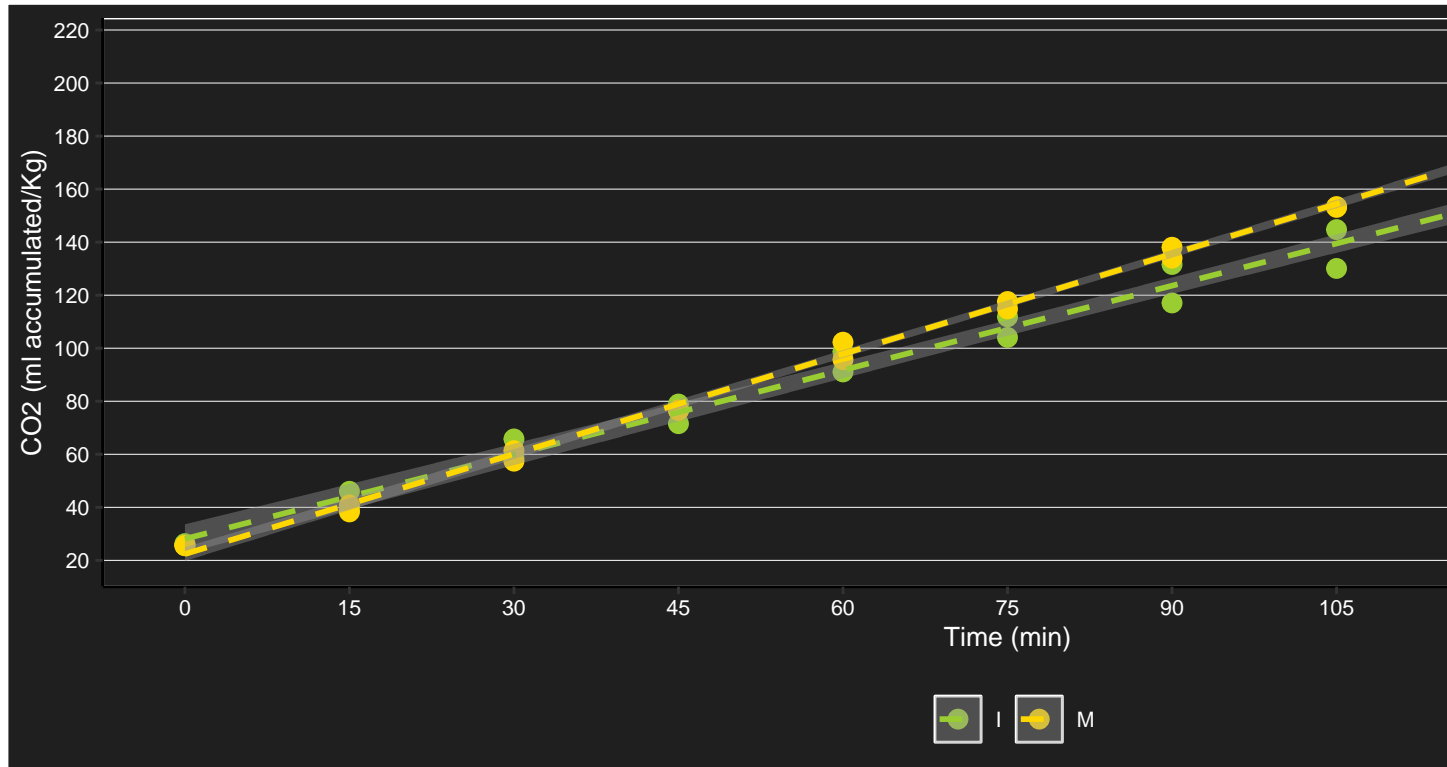
Ensayo 1

Respiración acumulada en frutos I y M



Immature and mature ubajay fruits were selected and randomly distributed in 4 jars, 2 immature and 2 mature, then respiration was quantified from accumulated CO₂ every 15 minutes for 150 minutes.

CO₂ acumulation



Descriptive table

```
## # A tibble: 20 x 7
## # Groups:   time_min [10]
##   time_min matu carbon_ac_n carbon_ac_Mean carbon_ac_sd carbon_ac_min carbon_ac_max
##   <fct>     <fct>      <int>      <dbl>      <dbl>      <dbl>      <dbl>
## 1 15      I          2        42.5        4.97        39.0        46.1
## 2 15      M          2        39.6        1.86        38.3        40.9
## 3 30      I          2        62.2        5.13        58.5        65.8
## 4 30      M          2        59.4        2.79        57.4        61.4
## 5 45      I          2        75.2        5.23        71.5        78.9
## 6 45      M          2        76.6        0.110       76.6        76.7
## 7 60      I          2        94.9        5.39        91.1        98.7
## 8 60      M          2        99.0        4.66        95.7       102.
## 9 75      I          2       108.         5.50       104.       112.
## 10 75     M          2       116.         1.97       115.       118.
## 11 90      I          2       124.        10.3       117.       132.
## 12 90     M          2       136.         2.91       134.       138.
## 13 105     I          2       137.        10.4       130.       145.
## 14 105     M          2       153.         0.221      153.       153.
## 15 120     I          2       157.        10.5       150.       164.
```

##	16	120	M	2	176.	4.77	172.	179.
##	17	135	I	2	170.	10.6	163.	178.
##	18	135	M	2	190.	6.59	185.	194.
##	19	150	I	2	186.	15.4	176.	197.
##	20	150	M	2	213.	3.02	211.	215.

Correlations over time



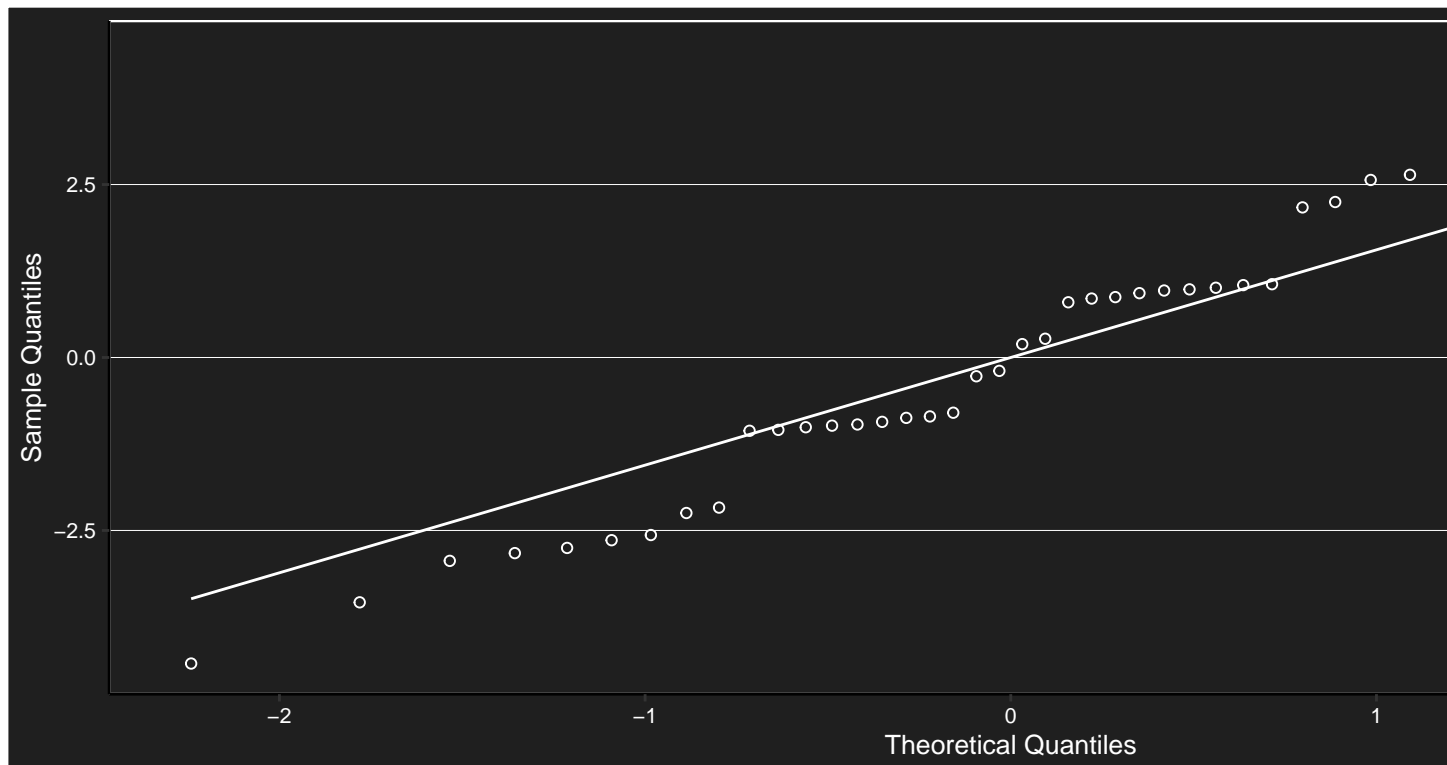
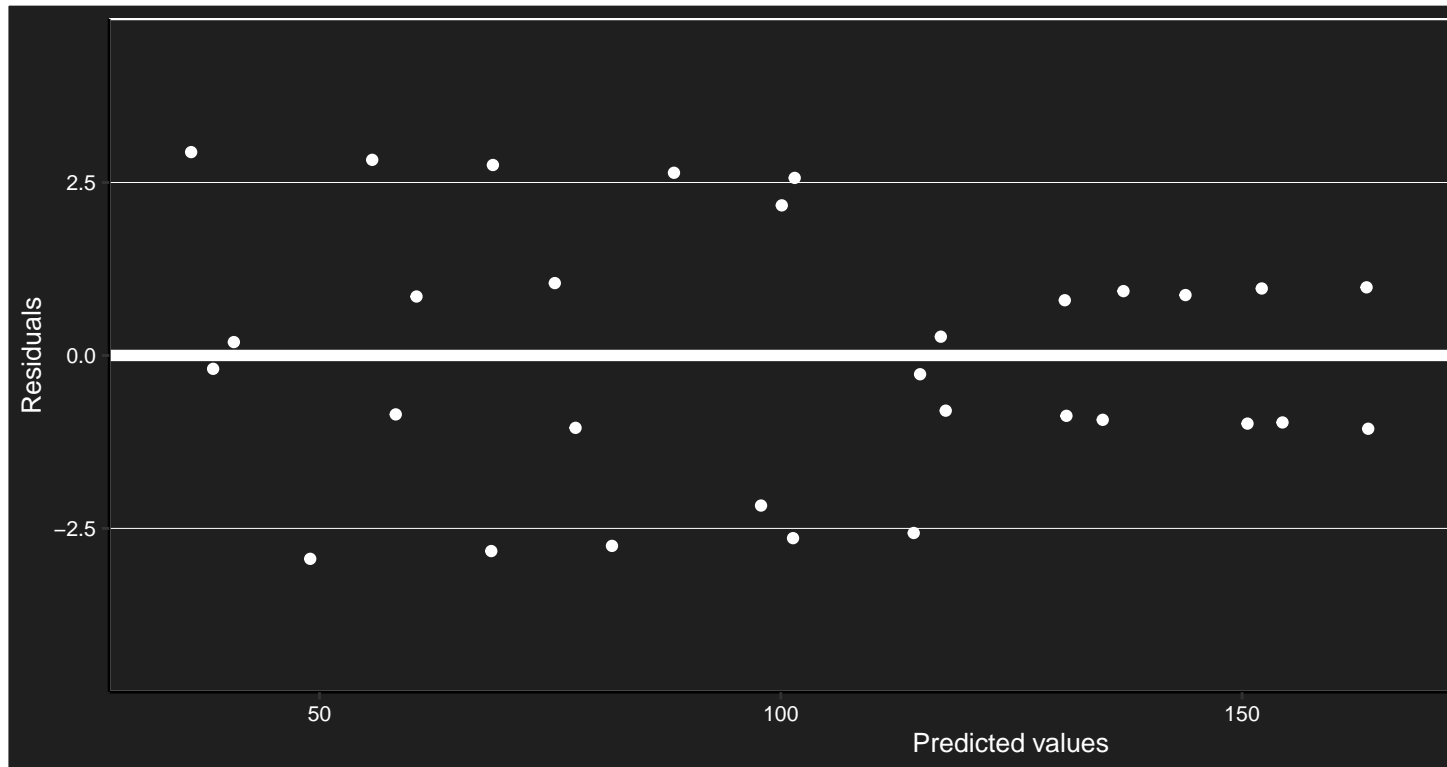
Covariance matrix

##		15	30	45	60	75	90	105	120	135	150
##	15	12.28	12.95	7.38	7.79	2.21	7.31	1.73	2.14	2.50	1.66
##	30	12.95	13.93	7.77	9.75	3.59	9.42	3.25	5.23	6.22	4.90
##	45	7.38	7.77	9.77	11.46	13.46	23.37	25.37	27.07	27.76	38.97
##	60	7.79	9.75	11.46	22.55	24.26	38.94	40.66	51.74	56.09	68.14
##	75	2.21	3.59	13.46	24.26	34.14	52.90	62.77	73.57	77.63	102.21
##	90	7.31	9.42	23.37	38.94	52.90	83.40	97.35	112.92	118.82	157.38
##	105	1.73	3.25	25.37	40.66	62.77	97.35	119.47	134.76	140.36	191.45
##	120	2.14	5.23	27.07	51.74	73.57	112.92	134.76	159.43	168.68	220.61
##	135	2.50	6.22	27.76	56.09	77.63	118.82	140.36	168.68	179.37	231.42
##	150	1.66	4.90	38.97	68.14	102.21	157.38	191.45	220.61	231.42	309.86

Marginal model with first-order autoregressive structure

```
## gls(model = (carbon_ac) ~ time_min * matu + basal, data = resp2w,
##      correlation = corAR1(form = ~1 | rep))
```

Assumptions



```
##
## Shapiro-Wilk normality test
##
## data: e
## W = 0.9761627, p-value = 0.549824
```

Model coefficients

```
##      (Intercept)      time_min30      time_min45      time_min60      time_min75
## -1088.309859117    19.625054307    32.708423844    52.333478151    65.416847689
##      time_min90      time_min105      time_min120      time_min135      time_min150
##      81.789723552    94.873093090    114.498147397    127.581516934    143.954392798
##      matuM      basal      time_min30:matuM      time_min45:matuM      time_min60:matuM
##      24.020301930    43.217057762      0.171220178      4.327249699      7.055345302
## time_min75:matuM time_min90:matuM time_min105:matuM time_min120:matuM time_min135:matuM
##      11.211374823    14.634773444    18.790802965    21.518898568    22.485337162
## time_min150:matuM
##      29.098326710
```

Anova

```
## Denom. DF: 19
##      numDF      F-value p-value
## (Intercept)      1 12422.59816 <.0001
## time_min      9  497.67981 <.0001
## matu      1  23.52445 0.0001
## basal      1  18.30623 0.0004
## time_min:matu      9  4.64451 0.0024
```

Simple effects

```
## $emmeans
## time_min = 15:
##      matu      emmean      SE      df      lower.CL      upper.CL
## I      29.0559310 3.99641289 4.38 18.3274684 39.7843936
## M      53.0762329 3.99641289 4.26 42.2426703 63.9097956
##
## time_min = 30:
##      matu      emmean      SE      df      lower.CL      upper.CL
## I      48.6809853 3.99641289 4.37 37.9473496 59.4146210
## M      72.8725074 3.99641289 4.38 62.1474410 83.5975738
##
## time_min = 45:
##      matu      emmean      SE      df      lower.CL      upper.CL
## I      61.7643548 3.99641289 4.23 50.9066429 72.6220667
## M      90.1119065 3.99641289 4.30 79.3151857 100.9086272
##
## time_min = 60:
##      matu      emmean      SE      df      lower.CL      upper.CL
## I      81.3894091 3.99641289 4.35 70.6412810 92.1375372
## M      112.4650564 3.99641289 4.37 101.7314207 123.1986921
##
```

```

## time_min = 75:
##   matu      emmean      SE    df    lower.CL    upper.CL
##   I      94.4727787 3.99641289 4.28  83.6565866 105.2889707
##   M     129.7044554 3.99641289 4.31 118.9126401 140.4962708
##
## time_min = 90:
##   matu      emmean      SE    df    lower.CL    upper.CL
##   I     110.8456545 3.99641289 4.38 100.1171919 121.5741172
##   M     149.5007299 3.99641289 4.26 138.6671673 160.3342926
##
## time_min = 105:
##   matu      emmean      SE    df    lower.CL    upper.CL
##   I     123.9290241 3.99641289 4.37 113.1953884 134.6626598
##   M     166.7401290 3.99641289 4.38 156.0150626 177.4651953
##
## time_min = 120:
##   matu      emmean      SE    df    lower.CL    upper.CL
##   I     143.5540784 3.99641289 4.23 132.6963665 154.4117903
##   M     189.0932789 3.99641289 4.30 178.2965581 199.8899996
##
## time_min = 135:
##   matu      emmean      SE    df    lower.CL    upper.CL
##   I     156.6374479 3.99641289 4.35 145.8893198 167.3855760
##   M     203.1430870 3.99641289 4.37 192.4094513 213.8767227
##
## time_min = 150:
##   matu      emmean      SE    df    lower.CL    upper.CL
##   I     173.0103238 3.99641289 4.28 162.1941317 183.8265159
##   M     226.1289524 3.99641289 4.31 215.3371371 236.9207678
##
## Degrees-of-freedom method: satterthwaite
## Results are given on the ( not the response) scale.
## Confidence level used: 0.95
##
## $contrasts
## time_min = 15:
##   contrast      estimate      SE    df t.ratio p.value
##   I - M     -24.0203019 7.19762359 6.29  -3.337  0.0146
##
## time_min = 30:
##   contrast      estimate      SE    df t.ratio p.value
##   I - M     -24.1915221 7.19762359 6.30  -3.361  0.0141
##
## time_min = 45:
##   contrast      estimate      SE    df t.ratio p.value
##   I - M     -28.3475516 7.19762359 4.51  -3.938  0.0135
##
## time_min = 60:
##   contrast      estimate      SE    df t.ratio p.value
##   I - M     -31.0756472 7.19762359 6.28  -4.317  0.0045
##
## time_min = 75:
##   contrast      estimate      SE    df t.ratio p.value
##   I - M     -35.2316768 7.19762359 6.24  -4.895  0.0024

```

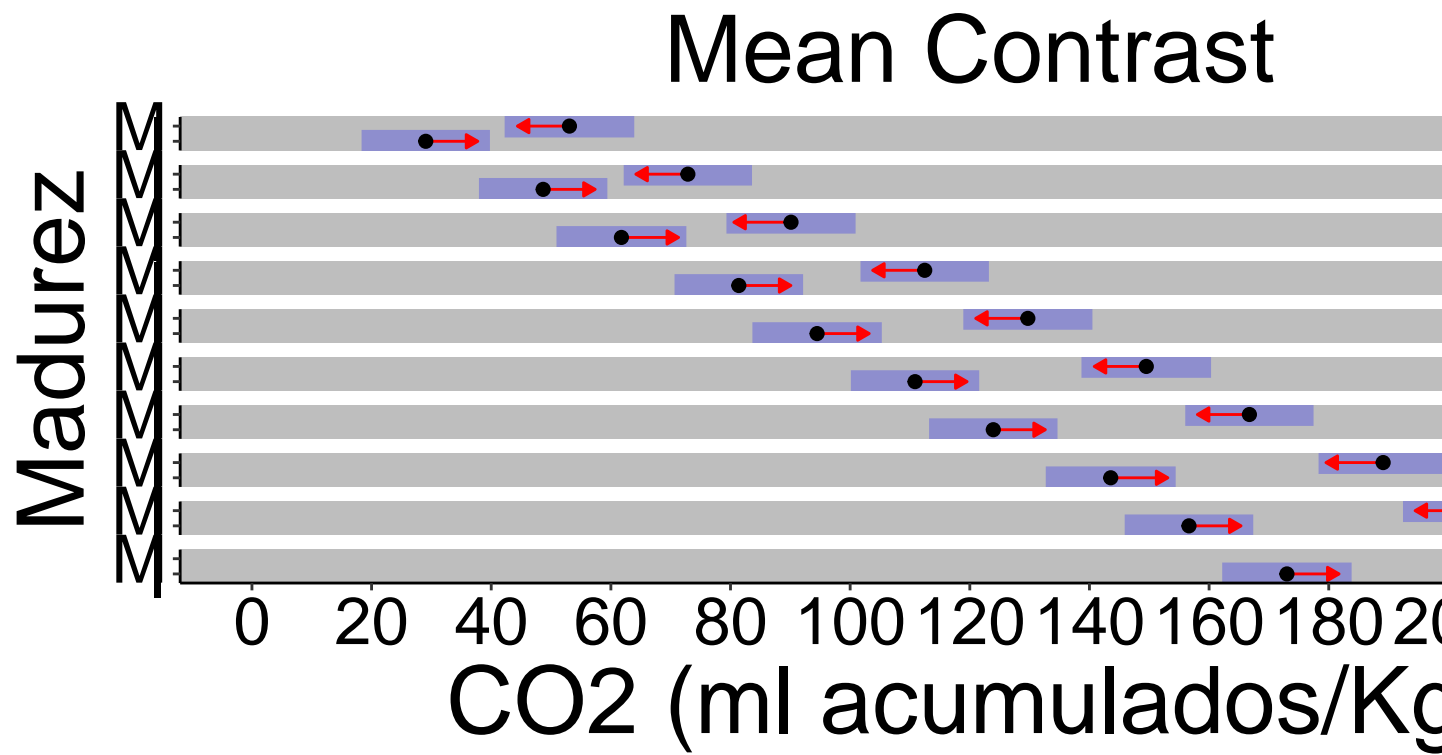
```

##
## time_min = 90:
## contrast      estimate          SE    df t.ratio p.value
## I - M      -38.6550754  7.19762359  6.29  -5.371  0.0015
##
## time_min = 105:
## contrast      estimate          SE    df t.ratio p.value
## I - M      -42.8111049  7.19762359  6.30  -5.948  0.0008
##
## time_min = 120:
## contrast      estimate          SE    df t.ratio p.value
## I - M      -45.5392005  7.19762359  4.51  -6.327  0.0021
##
## time_min = 135:
## contrast      estimate          SE    df t.ratio p.value
## I - M      -46.5056391  7.19762359  6.28  -6.461  0.0005
##
## time_min = 150:
## contrast      estimate          SE    df t.ratio p.value
## I - M      -53.1186286  7.19762359  6.24  -7.380  0.0003
##
## Note: contrasts are still on the ( scale
## Degrees-of-freedom method: satterthwaite

```

Statistically significant differences were found in the CO₂ respiration rate in each time between immature and mature *Hexachlamys edulis* fruits.

Comparison chart



Fitted model plot

