

# Male gain curve analyses

June 20, 2006

## 1 Introduction

Gain curves describe the relationship between the allocation of resources to a function and the fitness benefits returned by this allocation. Although gain curves can take many different forms, in general they are characterized by power laws of the form  $y = ax^b$ . Here we present the results of two experiments that quantify the male gain curve for canola.

## 2 Array-level gain curve

In this experiment we quantified total pollen deposition in arrays with different anther numbers and inflorescence sizes.

### 2.1 The data

Figure 1 plots the probability that an individual pollen grain reaches a stigma for different anther and flower numbers.

### 2.2 Linear models

We start with a simple, additive linear model of the probability of pollen export for anther number and inflorescence size.

```
> export.add <- lm(probExport ~ anthers + flowers, data = arrayData)
> summary(export.add)
```

Call:

```
lm(formula = probExport ~ anthers + flowers, data = arrayData)
```

Residuals:

	Min	1Q	Median	3Q	Max
	-0.0068616	-0.0024845	-0.0006813	0.0017046	0.0108378

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
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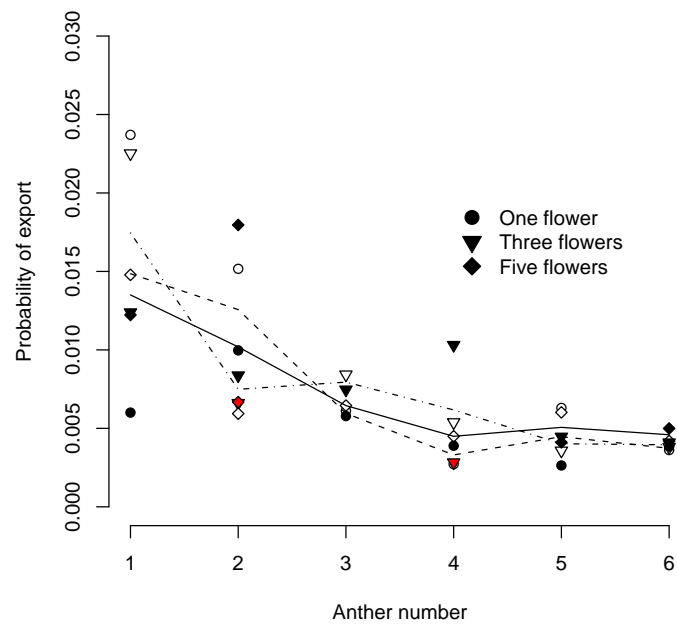


Figure 1: The gain curve for canola based on an array-level analysis of pollen deposition.

```

(Intercept) 0.0148836 0.0022024 6.758 1.06e-07 ***
anthers      -0.0021266 0.0003774 -5.634 2.83e-06 ***
flowers      0.0001103 0.0008145 0.135 0.893
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

```

Residual standard error: 0.003902 on 33 degrees of freedom
Multiple R-Squared: 0.4911, Adjusted R-squared: 0.4602
F-statistic: 15.92 on 2 and 33 DF, p-value: 1.445e-05

```

Anthers are a significant effect in this additive model, but an interaction model may be a better explanation for these data.

```

> export.int <- update(export.add, . ~ anthers * flowers)
> summary(export.int)

```

```

Call:
lm(formula = probExport ~ anthers + flowers + anthers:flowers,
    data = arrayData)

```

```

Residuals:
    Min       1Q   Median       3Q      Max
-0.007626 -0.002452 -0.000759  0.001857  0.010073

```

```

Coefficients:
              Estimate Std. Error t value Pr(>|t|)
(Intercept)  0.0170013  0.0039287   4.327 0.000138 ***
anthers      -0.0027394  0.0010120  -2.707 0.010800 *
flowers      -0.0009340  0.0017967  -0.520 0.606769
anthers:flowers 0.0003041  0.0004653   0.654 0.518052
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

```

Residual standard error: 0.003936 on 32 degrees of freedom
Multiple R-Squared: 0.4978, Adjusted R-squared: 0.4507
F-statistic: 10.57 on 3 and 32 DF, p-value: 5.492e-05

```

```

> anova(export.int, export.add)

```

Analysis of Variance Table

```

Model 1: probExport ~ anthers + flowers + anthers:flowers
Model 2: probExport ~ anthers + flowers
  Res.Df    RSS Df Sum of Sq    F Pr(>F)
1     32 0.00049579
2     33 0.00050241 -1 -0.00000662 0.4272 0.5181

```

There is no evidence for a significant interaction between anther number and inflorescence size for this linear approach. However, if we consider pollinator behaviour by including the number of flowers visited and amount of pollen transferred, we obtain slightly different conclusions.

```
> exp.stp <- stepAIC(export.int, scope = list(upper = ~anthers * flowers *
+      meanFlowerVisits * arrayRemoval, lower = ~1))
```

Start: AIC= -394.94

```
probExport ~ anthers + flowers + anthers:flowers
```

	Df	Sum of Sq	RSS	AIC
+ arrayRemoval	1	6.670e-05	4.291e-04	-398.14
- anthers:flowers	1	6.618e-06	5.024e-04	-396.47
<none>			4.958e-04	-394.94
+ meanFlowerVisits	1	3.653e-08	4.958e-04	-392.95

Step: AIC= -398.14

```
probExport ~ anthers + flowers + arrayRemoval + anthers:flowers
```

	Df	Sum of Sq	RSS	AIC
<none>			4.291e-04	-398.14
+ flowers:arrayRemoval	1	1.306e-05	4.160e-04	-397.26
- anthers:flowers	1	4.624e-05	4.753e-04	-396.46
+ meanFlowerVisits	1	2.907e-06	4.262e-04	-396.39
+ anthers:arrayRemoval	1	2.424e-06	4.267e-04	-396.35
- arrayRemoval	1	6.670e-05	4.958e-04	-394.94

```
> dropterm(exp.stp, test = "F")
```

Single term deletions

Model:

```
probExport ~ anthers + flowers + arrayRemoval + anthers:flowers
```

	Df	Sum of Sq	RSS	AIC	F Value	Pr(F)
<none>			4.291e-04	-398.14		
arrayRemoval	1	6.670e-05	4.958e-04	-394.94	4.82	0.03576 *
anthers:flowers	1	4.624e-05	4.753e-04	-396.46	3.34	0.07723 .

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

```
> exp.aov <- aov(probExport ~ anthers * flowers + arrayRemoval, data = arrayData)
> summary(exp.aov)
```

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
anthers	1	0.00048451	0.00048451	35.0037	1.557e-06 ***
flowers	1	0.00000028	0.00000028	0.0202	0.88793

```

arrayRemoval      1 0.00002708 0.00002708  1.9566   0.17180
anthers:flowers    1 0.00004624 0.00004624  3.3405   0.07723 .
Residuals         31 0.00042909 0.00001384
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

> coef(exp.aov)

      (Intercept)      anthers      flowers  arrayRemoval anthers:flowers
1.680757e-02    -2.633648e-03   -5.469398e-03   1.373010e-08   9.823610e-04

```

We find that the total amount of pollen removed in the array is an important predictor of the probability of pollen export. Furthermore, with this addition, the interaction between anther number and inflorescence size becomes marginally significant.

Another approach is to log transform the probability of pollen export to investigate non-linear changes with anther number.

```

> export.log <- update(export.int, log(probExport) ~ .)
> summary(export.log)

Call:
lm(formula = log(probExport) ~ anthers + flowers + anthers:flowers,
    data = arrayData)

```

```

Residuals:
      Min       1Q   Median       3Q      Max
-0.71680 -0.22690  0.02449  0.21408  0.65618

```

```

Coefficients:
              Estimate Std. Error t value Pr(>|t|)
(Intercept)   -4.05913    0.38610  -10.513 6.57e-12 ***
anthers        -0.32149    0.09945   -3.233 0.00284 **
flowers        -0.05191    0.17657   -0.294 0.77067
anthers:flowers  0.03429    0.04573    0.750 0.45879
---

```

```

Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

```

Residual standard error: 0.3868 on 32 degrees of freedom
Multiple R-Squared:  0.5955,    Adjusted R-squared:  0.5576
F-statistic: 15.7 on 3 and 32 DF,  p-value: 1.868e-06

```

## 2.3 Selfing

The count of transgenic pollen on stigmas provides an estimate of the total amount of self pollen deposited within plants.

```
> self.lm <- lm(selfing ~ anthers * flowers, data = arrayData)
> summary(self.lm)
```

Call:

```
lm(formula = selfing ~ anthers * flowers, data = arrayData)
```

Residuals:

	Min	1Q	Median	3Q	Max
	-141.691	-40.937	-6.457	54.320	97.264

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	-248.91	178.42	-1.395	0.1905
anthers	86.40	39.70	2.176	0.0522 .
flowers	219.83	74.01	2.970	0.0127 *
anthers:flowers	-45.09	16.66	-2.706	0.0204 *

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 74.2 on 11 degrees of freedom

Multiple R-Squared: 0.5165, Adjusted R-squared: 0.3846

F-statistic: 3.917 on 3 and 11 DF, p-value: 0.03981

These results suggest some influence of anther number and inflorescence size on selfing. However, the plot of these data (Figure 2) suggests more analyses are required.

## 2.4 Non-linear regression

Typically, gain curves are analysed with power-law distributions.

Formula:  $I(\text{probExport}/\text{anthers}) \sim a * \text{anthers}^{(b - 1)}$

Parameters:

	Estimate	Std. Error	t value	Pr(> t )
a	0.015056	0.002859	5.266	0.000365 ***
b	-0.596431	0.551845	-1.081	0.305164

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.004077 on 10 degrees of freedom

Formula:  $I(\text{probExport}/\text{anthers}) \sim a * \text{anthers}^{(b - 1)}$

Parameters:

	Estimate	Std. Error	t value	Pr(> t )
a	0.017365	0.001588	10.934	3.01e-07 ***

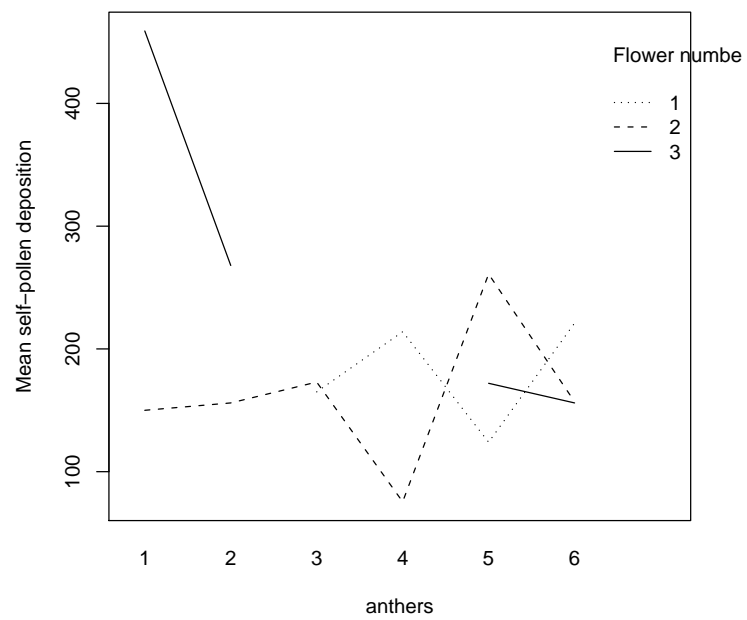


Figure 2: Self pollen deposition.

```

b -0.943255    0.359687   -2.622    0.0237 *
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.002255 on 11 degrees of freedom

Formula: I(probExport/anthers) ~ a * anthers^(b - 1)

Parameters:
      Estimate Std. Error t value Pr(>|t|)
a   0.013591   0.001212  11.216 1.37e-06 ***
b  -0.518122   0.248459  -2.085  0.0667 .
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.001732 on 9 degrees of freedom

```

With the exception of  $b$  for three-flowered plants, the estimate of each parameter is significantly different from zero. However, plotting the predicted and observed values (Figure 3) shows that these power-law distributions are significantly underfitting the values for large anther numbers.

### 3 Pollen shadow

In the second experiment, a focal plant with transgenic pollen was surrounded by concentric squares of wild-type plants. Pollen export from the focal plant was estimated by counting transgenic pollen on wild-type stigmas.

We'll start with a look at the pollinator visits in each array.

The following object(s) are masked from visits ( position 32 ) :

```
array_id arrive bout depart donor_anthers duration plant_id recipient_anthers
```

Call:

```
lm(formula = as.integer(duration) ~ array_id/(as.integer(donor_anthers) +
as.integer(recipient_anthers)), data = visits)
```

Residuals:

	Min	1Q	Median	3Q	Max
	-13.203	-6.342	-3.203	3.092	65.233

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	9.056408	0.603822	14.998	<2e-16 ***
array_id	-0.033354	0.068805	-0.485	0.628
array_id:as.integer(donor_anthers)	0.058831	0.031241	1.883	0.060 .



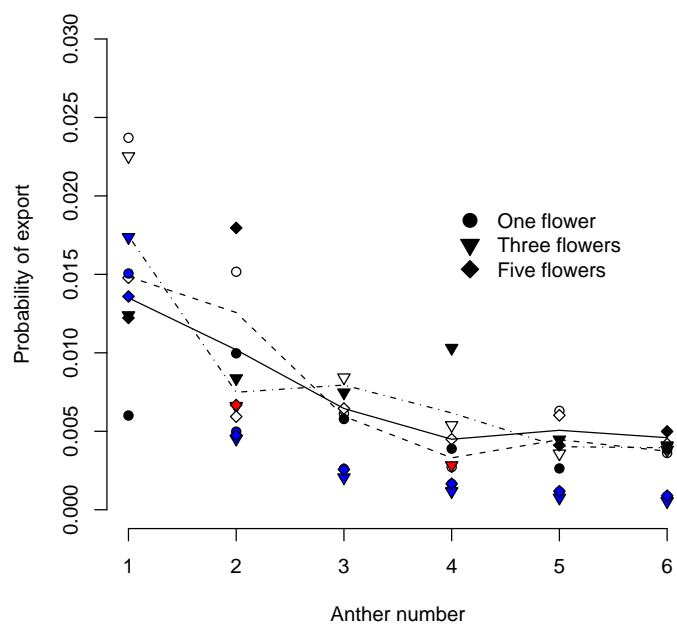


Figure 3: Predicted values from the non-linear regression.

```
array_id:as.integer(recipient_anthers) 0.005446 0.023742 0.229 0.819
---
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 9.952 on 968 degrees of freedom
Multiple R-Squared: 0.01903, Adjusted R-squared: 0.01599
F-statistic: 6.26 on 3 and 968 DF, p-value: 0.0003293
```

This provides marginal evidence that pollinators respond to the number of donor anthers.

### 3.1 Selfing

As with the gain curve experiment, we can look at self-pollen deposition for different anther numbers.

The following object(s) are masked from visits :

```
array_id donor_anthers recipient_anthers
```

Call:

```
lm(formula = gus_pollen ~ donor_anthers, data = selfing)
```

Residuals:

	Min	1Q	Median	3Q	Max
	-474.721	-145.546	3.553	114.841	444.128

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	189.32	157.10	1.205	0.240
donor_anthers	73.85	43.64	1.692	0.104

Residual standard error: 248.2 on 24 degrees of freedom

Multiple R-Squared: 0.1066, Adjusted R-squared: 0.06939

F-statistic: 2.864 on 1 and 24 DF, p-value: 0.1035

Call:

```
lm(formula = gus_pollen ~ wt_pollen, data = selfing)
```

Residuals:

	Min	1Q	Median	3Q	Max
	-469.54	-154.88	-54.35	175.30	492.30

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	390.4423	88.7636	4.399	0.000192 ***
wt_pollen	0.6961	0.9791	0.711	0.483942

```

---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 259.8 on 24 degrees of freedom
Multiple R-Squared:  0.02063,    Adjusted R-squared:  -0.02018 
F-statistic: 0.5055 on 1 and 24 DF,  p-value: 0.4839

Call:
lm(formula = gus_pollen ~ donor_anthers + recipient_anthers,
    data = selfing)

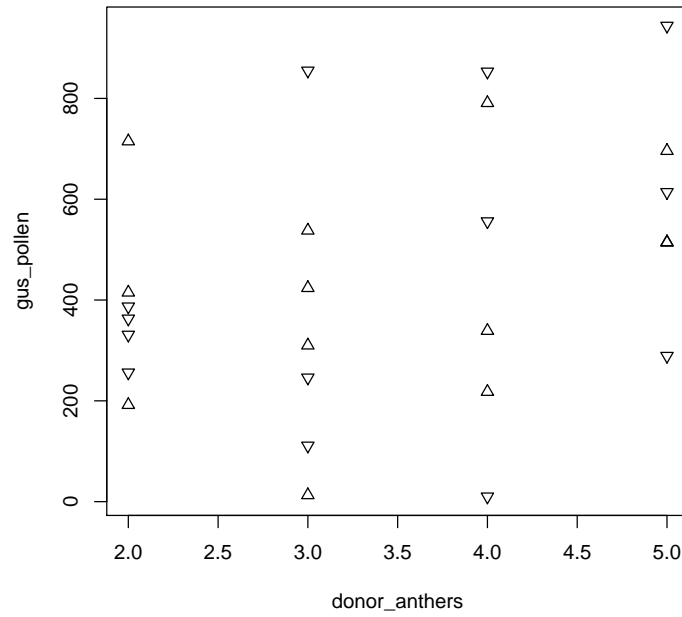
Residuals:
    Min       1Q   Median       3Q      Max
-482.906 -137.523   2.289  123.050  436.191

Coefficients:
              Estimate Std. Error t value Pr(>|t|)
(Intercept)    188.474    160.477   1.174   0.252
donor_anthers    82.140     67.912   1.210   0.239
recipient_anthers -8.042     49.720  -0.162   0.873

Residual standard error: 253.4 on 23 degrees of freedom
Multiple R-Squared:  0.1076,    Adjusted R-squared:  0.03003 
F-statistic: 1.387 on 2 and 23 DF,  p-value: 0.2699

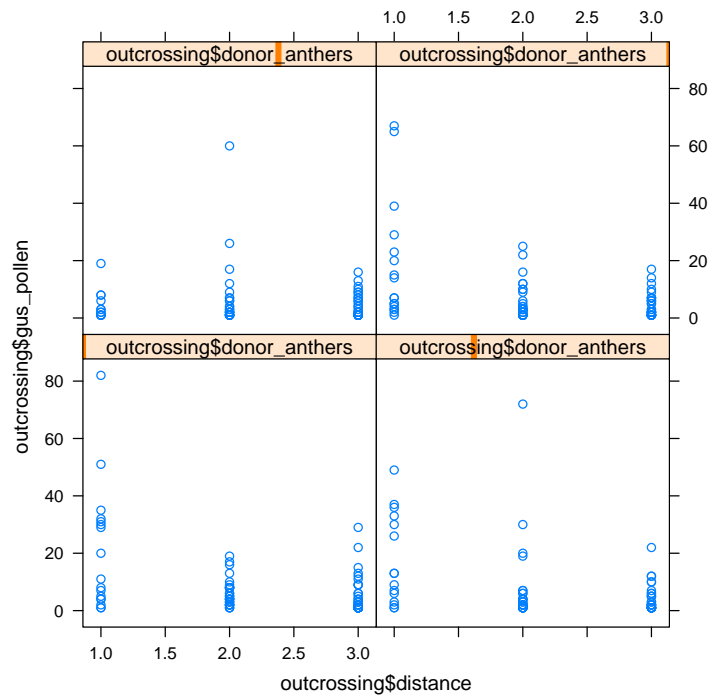
```

We find no evidence that selfing is correlated with anther number or pollen export.



### 3.2 Distance

A plot of the distance pollen travels for each number of donor anthers shows significant scatter, but perhaps an interaction.



Call:  
lm(formula = distance ~ donor\_anthers \* recipient\_anthers, data = outcrossing)

Residuals:

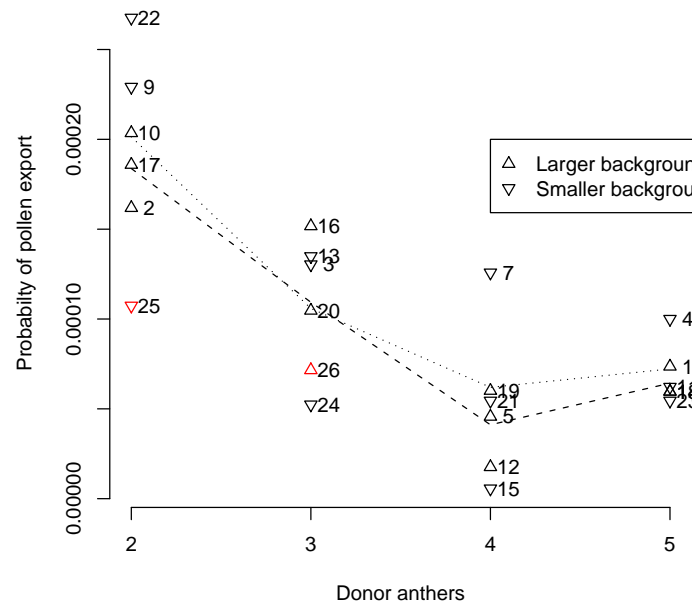
Min	1Q	Median	3Q	Max
-1.3280	-0.2946	-0.2475	0.7296	0.7537

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	2.47632	0.30889	8.017	1.81e-14 ***
donor_anthers	-0.05978	0.10752	-0.556	0.579
recipient_anthers	-0.06101	0.09996	-0.610	0.542
donor_anthers:recipient_anthers	0.01611	0.02657	0.606	0.545

---  
Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.7545 on 336 degrees of freedom  
Multiple R-Squared: 0.001375, Adjusted R-squared: -0.007542  
F-statistic: 0.1542 on 3 and 336 DF, p-value: 0.927



Start: AIC= -472.43  
pollen\_export ~ donor\_anthers + background + donor\_anthers:background

	Df	Sum of Sq	RSS	AIC
- donor_anthers:background	1	1.675e-12	4.860e-08	-474.43
+ mean_visit	1	5.769e-09	4.283e-08	-473.46
+ distance	1	4.159e-09	4.444e-08	-472.57
<none>			4.860e-08	-472.43

Step: AIC= -474.43  
pollen\_export ~ donor\_anthers + background

	Df	Sum of Sq	RSS	AIC
- background	1	6.916e-10	4.929e-08	-476.09
+ mean_visit	1	5.748e-09	4.285e-08	-475.45
+ distance	1	3.891e-09	4.471e-08	-474.43
<none>			4.860e-08	-474.43
+ donor_anthers:background	1	1.675e-12	4.860e-08	-472.43
- donor_anthers	1	5.522e-08	1.038e-07	-458.21

Step: AIC= -476.09  
pollen\_export ~ donor\_anthers

	Df	Sum of Sq	RSS	AIC
+ mean_visit	1	5.797e-09	4.349e-08	-477.09
<none>			4.929e-08	-476.09
+ distance	1	3.336e-09	4.595e-08	-475.77
+ background	1	6.916e-10	4.860e-08	-474.43
- donor_anthers	1	5.522e-08	1.045e-07	-460.05

Step: AIC= -477.09

pollen\_export ~ donor\_anthers + mean\_visit

	Df	Sum of Sq	RSS	AIC
<none>			4.349e-08	-477.09
+ distance	1	1.855e-09	4.164e-08	-476.13
- mean_visit	1	5.797e-09	4.929e-08	-476.09
+ donor_anthers:mean_visit	1	1.730e-09	4.176e-08	-476.06
+ background	1	6.427e-10	4.285e-08	-475.45
- donor_anthers	1	6.016e-08	1.037e-07	-458.25

Single term deletions

Model:

pollen\_export ~ donor\_anthers + mean\_visit

	Df	Sum of Sq	RSS	AIC	F Value	Pr(>F)
<none>			4.349e-08	-477.09		
donor_anthers	1	6.016e-08	1.037e-07	-458.25	29.05	2.404e-05 ***
mean_visit	1	5.797e-09	4.929e-08	-476.09	2.80	0.1092

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
donor_anthers	1	5.5221e-08	5.5221e-08	26.6620	4.078e-05 ***
mean_visit	1	5.7970e-09	5.7970e-09	2.7988	0.1092
Residuals	21	4.3494e-08	2.0710e-09		

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

	(Intercept)	donor_anthers	mean_visit
	2.172152e-04	-4.564110e-05	4.320045e-06

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
donor_anthers	1	5.5221e-08	5.5221e-08	24.4978	8.9e-05 ***
background	1	6.9200e-10	6.9200e-10	0.3068	0.5861
mean_visit	1	5.7480e-09	5.7480e-09	2.5499	0.1268
donor_anthers:background	1	2.3000e-11	2.3000e-11	0.0103	0.9204
Residuals	19	4.2828e-08	2.2540e-09		

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

So, we find that pollen export declines with the number of donor anthers and increases with pollinator visits.