

Dispersal Beta: Finding Ideal Parameters

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September 30, 2015

1 Summary

This paper explores the range of values and accuracy of the *beta* (dispersal) parameter in SORTIE-ND for adult trees in our validation plots. For each set of parameters in the 081815c runs, I varied them by 10% to test whether adjusting the parameters would increase the overall model fit. This will also give us an idea of how much swing these parameters have within the simulations.

For each species/step combination, I'll need to evaluate whether the parameters improve or hurt the model fit. I'll be using a general linear model that regresses the expected values (the "realPlots" means) against the simulated values of the model. The model improves as the slope approaches 1. If realPlots data are on the y-axis, then points or lines that fall above the "1" demarkation line are *underpredicting* the true value; and points or lines that fall below the "1" demarkation line are *overpredicting* the true value.

We'll need to view all of the data – data for the 90, 100, and 110 percent values of the parameters – before we can conduct the analysis.

View the Rnw document to view the code; otherwise, I am only printing outputs to save some space and make this document more readable.

2 Basal Area: At the nintieith percentile

Call:

```
lm(formula = SimAbsBA ~ ExpAbsBA, data = PlotMeans)
```

Residuals:

	Min	1Q	Median	3Q	Max
	-25.5307	0.4451	1.8166	1.9417	11.1927

Coefficients:

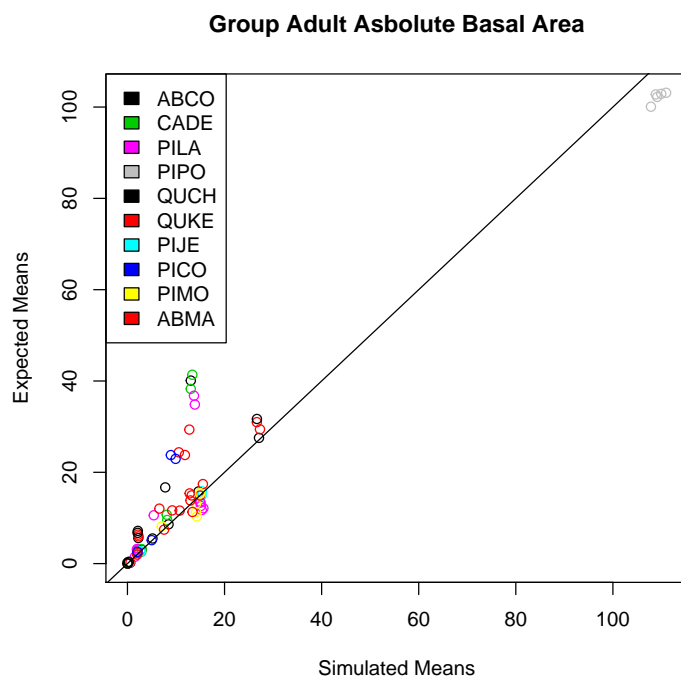
	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	-1.88002	0.75967	-2.475	0.015 *
ExpAbsBA	0.98567	0.02868	34.373	<2e-16 ***

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

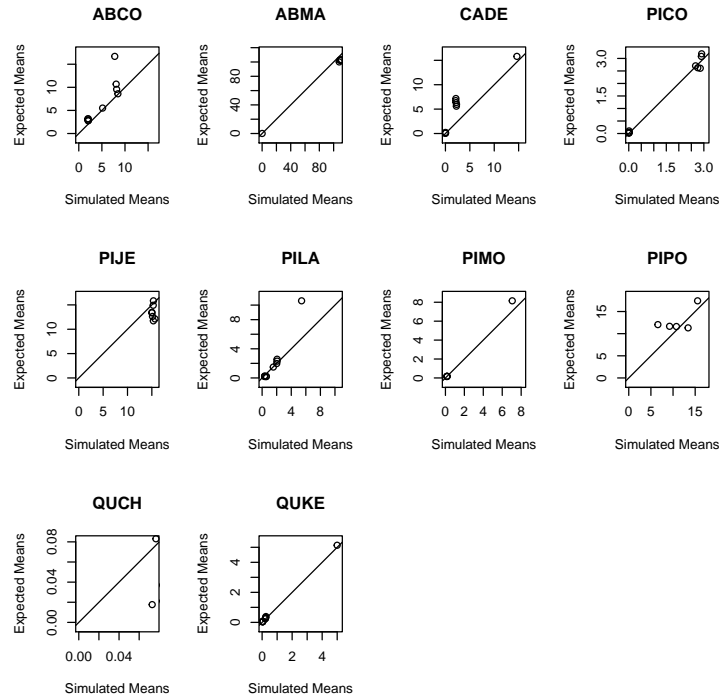
Residual standard error: 6.561 on 100 degrees of freedom

Multiple R-squared: 0.922, Adjusted R-squared: 0.9212

F-statistic: 1181 on 1 and 100 DF, p-value: < 2.2e-16



Now, how are the individual species doing?



> sppSlopes

	species	ba90
1	ABCO	3.0700509
2	ABMA	0.9345975
3	CADE	1.0286246
4	PICO	0.9962182
5	PIJE	-0.9722587
6	PILA	2.3945161
7	PIMO	0.9452041
8	PIPO	0.4690694
9	QUCH	-0.7826302
10	QUKE	1.0128360

3 At the original parameter designation

Call:

```
lm(formula = SimAbsBA ~ ExpAbsBA, data = PlotMeans)
```

Residuals:

	Min	1Q	Median	3Q	Max
	-24.5872	0.6374	1.8231	2.1861	11.1059

Coefficients:

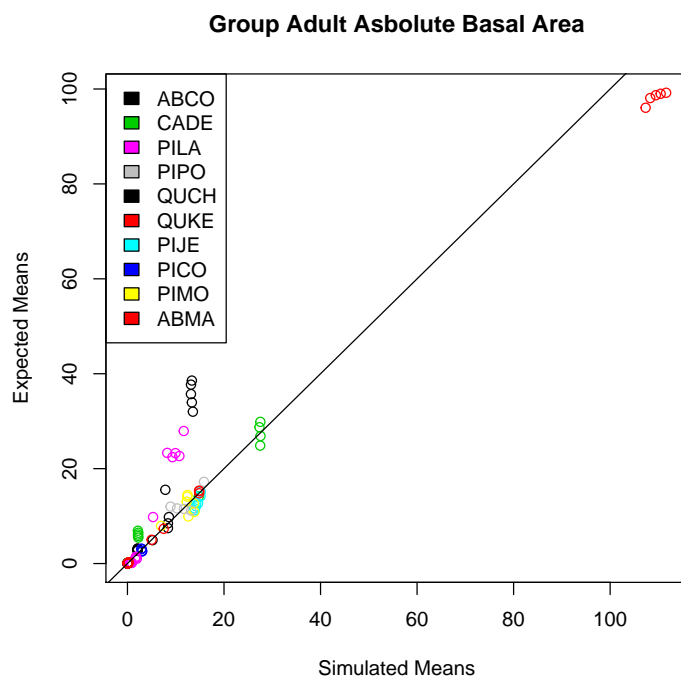
	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	-1.8487	0.7418	-2.492	0.0143 *
ExpAbsBA	1.0316	0.0293	35.208	<2e-16 ***

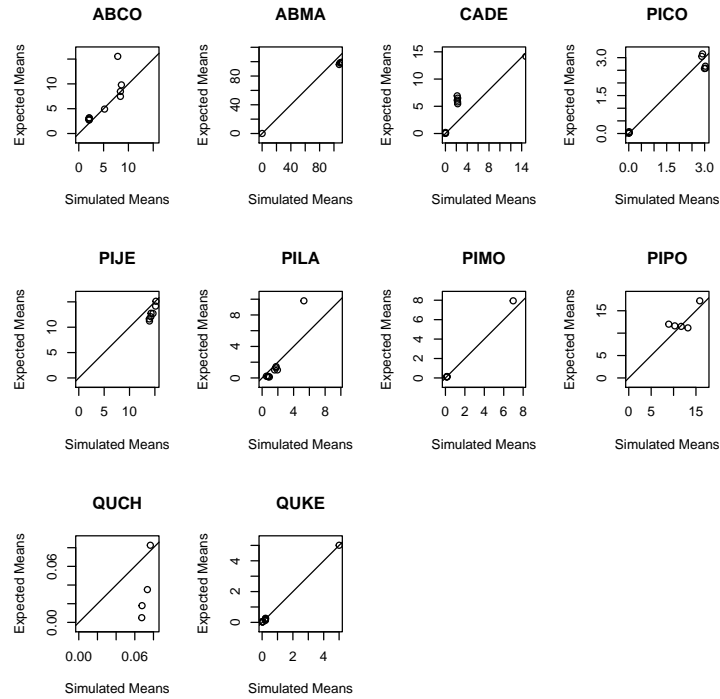
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 6.421 on 100 degrees of freedom

Multiple R-squared: 0.9254, Adjusted R-squared: 0.9246

F-statistic: 1240 on 1 and 100 DF, p-value: < 2.2e-16





> sppSlopes

	species	ba90	ba100
1	ABCO	3.0700509	2.8679859
2	ABMA	0.9345975	0.8968508
3	CADE	1.0286246	0.9240230
4	PICO	0.9962182	0.9295847
5	PIJE	-0.9722587	2.4811348
6	PILA	2.3945161	2.5907967
7	PIMO	0.9452041	0.9567542
8	PIPO	0.4690694	0.6919119
9	QUCH	-0.7826302	7.0914178
10	QUKE	1.0128360	1.0172449

4 At the one hundred and tenth percentile

Call:

```
lm(formula = SimAbsBA ~ ExpAbsBA, data = PlotMeans)
```

Residuals:

	Min	1Q	Median	3Q	Max
	-26.2569	0.5016	1.8413	1.9484	12.6616

Coefficients:

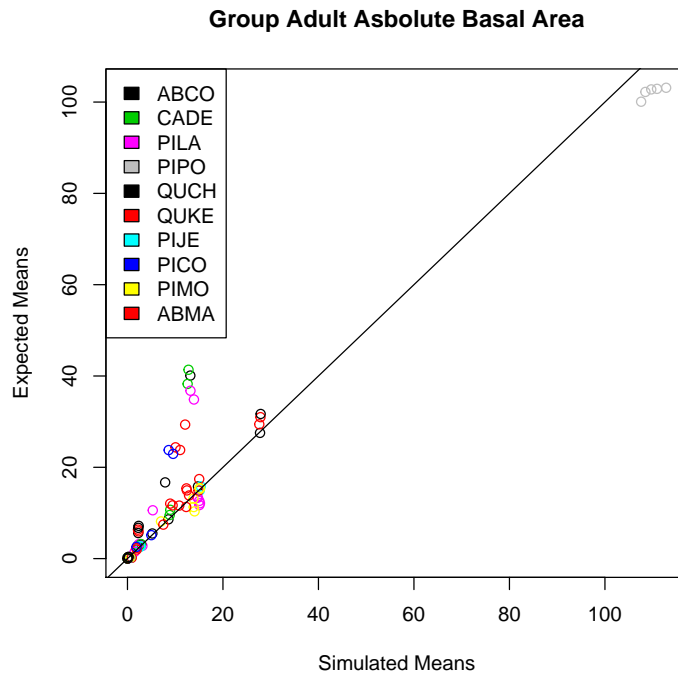
	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	-1.88329	0.77332	-2.435	0.0166 *
ExpAbsBA	0.98940	0.02919	33.894	<2e-16 ***

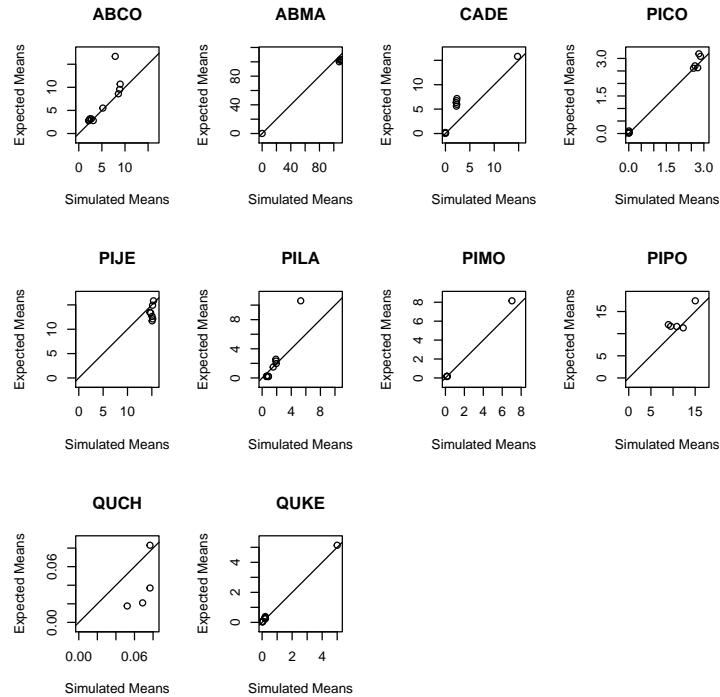
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 6.679 on 100 degrees of freedom

Multiple R-squared: 0.9199, Adjusted R-squared: 0.9191

F-statistic: 1149 on 1 and 100 DF, p-value: < 2.2e-16





> sppSlopes

	species	ba90	ba100	ba110
1	ABCO	3.0700509	2.8679859	3.2140277
2	ABMA	0.9345975	0.8968508	0.9291240
3	CADE	1.0286246	0.9240230	1.0000032
4	PICO	0.9962182	0.9295847	1.0287976
5	PIJE	-0.9722587	2.4811348	1.7046706
6	PILA	2.3945161	2.5907967	2.5827189
7	PIMO	0.9452041	0.9567542	0.9671009
8	PIPO	0.4690694	0.6919119	0.8387516
9	QUCH	-0.7826302	7.0914178	1.7139325
10	QUKE	1.0128360	1.0172449	1.0054535

5 Adult Density: At the ninetieth percentile

Call:

```
lm(formula = SimAbsDen ~ ExpAbsDen, data = PlotMeans)
```

Residuals:

	Min	1Q	Median	3Q	Max
	-1060.66	-108.88	-82.30	-20.85	1172.87

Coefficients:

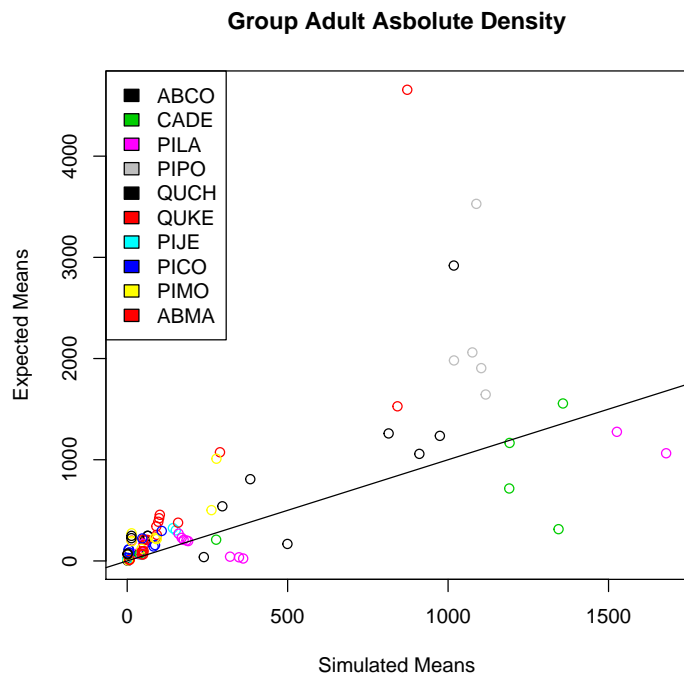
	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	83.95551	32.85628	2.555	0.0121 *
ExpAbsDen	0.39722	0.03787	10.489	<2e-16 ***

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

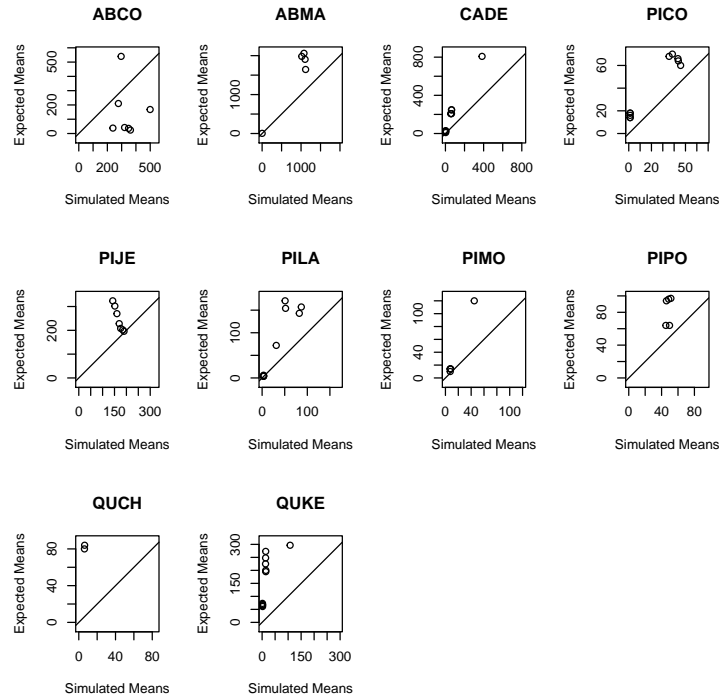
Residual standard error: 286.4 on 100 degrees of freedom

Multiple R-squared: 0.5238, Adjusted R-squared: 0.5191

F-statistic: 110 on 1 and 100 DF, p-value: < 2.2e-16



Now, how are the individual species doing?



> sppSlopes

	species	ba90	ba100	ba110	den90
1	ABCO	3.0700509	2.8679859	3.2140277	0.9199434
2	ABMA	0.9345975	0.8968508	0.9291240	2.0413250
3	CADE	1.0286246	0.9240230	1.0000032	2.4190038
4	PICO	0.9962182	0.9295847	1.0287976	1.2006573
5	PIJE	-0.9722587	2.4811348	1.7046706	-2.9059395
6	PILA	2.3945161	2.5907967	2.5827189	2.7015985
7	PIMO	0.9452041	0.9567542	0.9671009	3.6933377
8	PIPO	0.4690694	0.6919119	0.8387516	1.9794173
9	QUCH	-0.7826302	7.0914178	1.7139325	-13.7199921
10	QUKE	1.0128360	1.0172449	1.0054535	2.6207394

6 At the original parameter designation

Call:

```
lm(formula = SimAbsDen ~ ExpAbsDen, data = PlotMeans)
```

Residuals:

Min	1Q	Median	3Q	Max
-776.11	-73.57	-48.99	-4.74	687.39

Coefficients:

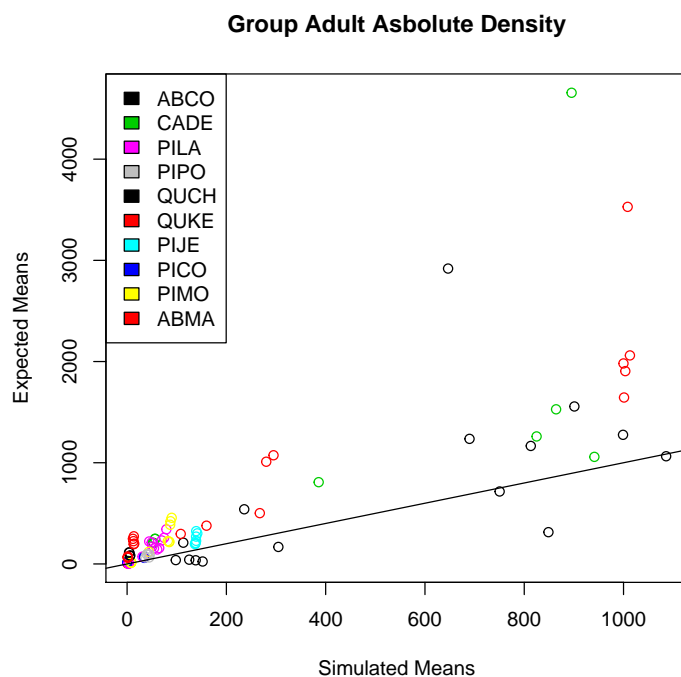
	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	52.13734	22.09152	2.36	0.0202 *
ExpAbsDen	0.34780	0.02546	13.66	<2e-16 ***

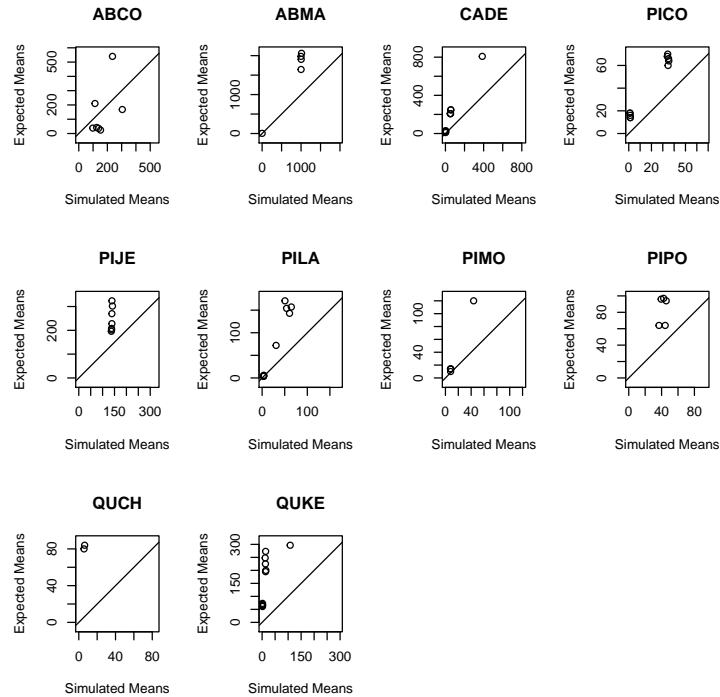
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 192.5 on 100 degrees of freedom

Multiple R-squared: 0.651, Adjusted R-squared: 0.6475

F-statistic: 186.6 on 1 and 100 DF, p-value: < 2.2e-16





> sppSlopes

	species	ba90	ba100	ba110	den90	den100
1	ABCO	3.0700509	2.8679859	3.2140277	0.9199434	1.366993
2	ABMA	0.9345975	0.8968508	0.9291240	2.0413250	2.218563
3	CADE	1.0286246	0.9240230	1.0000032	2.4190038	2.345902
4	PICO	0.9962182	0.9295847	1.0287976	1.2006573	1.479854
5	PIJE	-0.9722587	2.4811348	1.7046706	-2.9059395	27.477350
6	PILA	2.3945161	2.5907967	2.5827189	2.7015985	3.557684
7	PIMO	0.9452041	0.9567542	0.9671009	3.6933377	4.056143
8	PIPO	0.4690694	0.6919119	0.8387516	1.9794173	1.296860
9	QUCH	-0.7826302	7.0914178	1.7139325	-13.7199921	-15.618990
10	QUKE	1.0128360	1.0172449	1.0054535	2.6207394	2.577005

7 At the one hundred and tenth percentile

Call:

```
lm(formula = SimAbsDen ~ ExpAbsDen, data = PlotMeans)
```

Residuals:

Min	1Q	Median	3Q	Max
-2517.0	-363.4	-298.4	-217.8	2984.8

Coefficients:

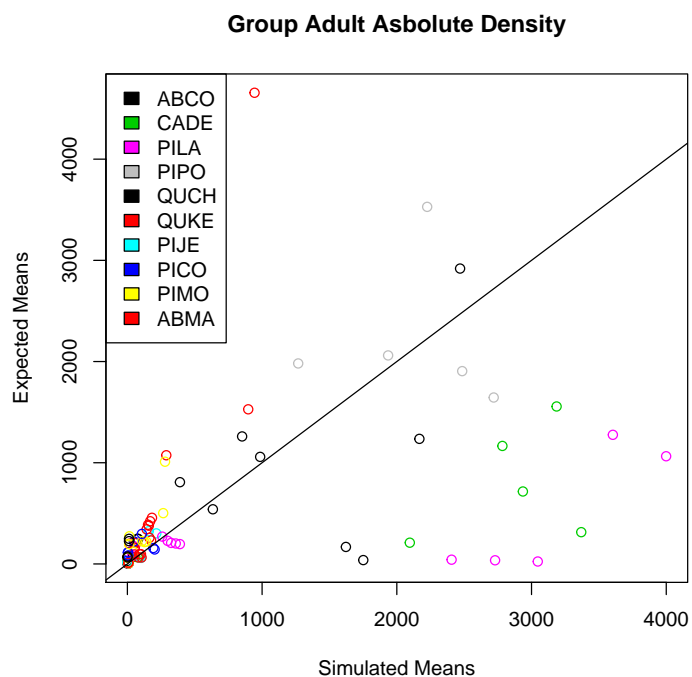
	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	289.5353	101.0477	2.865	0.00508 **
ExpAbsDen	0.6811	0.1165	5.848	6.29e-08 ***

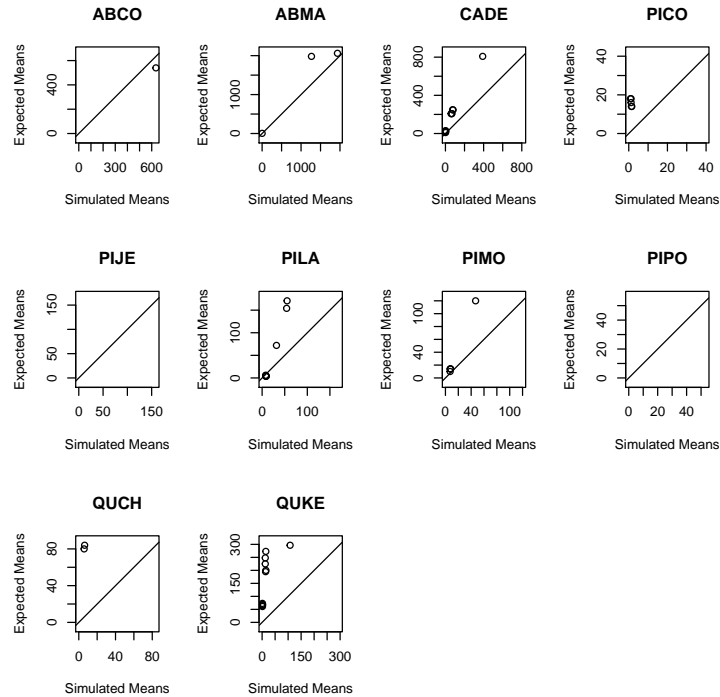
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 880.7 on 100 degrees of freedom

Multiple R-squared: 0.2548, Adjusted R-squared: 0.2474

F-statistic: 34.2 on 1 and 100 DF, p-value: 6.289e-08





```
> sppSlopes
```

	species	ba90	ba100	ba110	den90	den100	den110
1	ABCO	3.0700509	2.8679859	3.2140277	0.9199434	1.366993	0.2191432
2	ABMA	0.9345975	0.8968508	0.9291240	2.0413250	2.218563	0.7642356
3	CADE	1.0286246	0.9240230	1.0000032	2.4190038	2.345902	2.2804536
4	PICO	0.9962182	0.9295847	1.0287976	1.2006573	1.479854	0.5540442
5	PIJE	-0.9722587	2.4811348	1.7046706	-2.9059395	27.477350	-0.6207801
6	PILA	2.3945161	2.5907967	2.5827189	2.7015985	3.557684	1.0074069
7	PIMO	0.9452041	0.9567542	0.9671009	3.6933377	4.056143	2.3065268
8	PIPO	0.4690694	0.6919119	0.8387516	1.9794173	1.296860	-0.2537711
9	QUCH	-0.7826302	7.0914178	1.7139325	-13.7199921	-15.618990	-14.0318101
10	QUKE	1.0128360	1.0172449	1.0054535	2.6207394	2.577005	2.5996866

```
> write.csv(sppSlopes, file=paste(parName, ".csv", sep=""))
```