Data Analysis 2

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If we want an abstract it will go here. References are in the form Astley (1987) or (Astley 1987). For more information see here.

Introduction

Our clients conducted an experiment to determine the effect pine tissues, precipitation levels, time, and the interaction of these variables effects starch content. In total, 408 entries were recorded. The experiment was replicated at two locations as well and not all measurements within each replication were taken from the same sample location. (dont like that last line)

We intend to analysis the results of this data below. We will review the variables, fit multiple models, and make a suggestion to the client. The data set, data.csv, and all other files used in this project can be found on our Github page.

Exploring the Data

Variables

In the data set provided by the client there are four tissue types which are abbreviated as END, IT, LM, and UM. This can be found in the tissu column. The two precipitation levels, control and drought, are in the treatment column. As the column name may suggest, this will be considered the treatment,. The time component of the experiment is not simply one variable. The time column consists of six different times, with six being denoted by the first six letters of the alphabet. In addition to time, the column dayPeriod indicates whether the measurement was taken in the day or at night. Time points C and D appear to correspond to a dayPeriod of night, while all other time points are during the day. Note, the measurements for the starch contents can be found in the StarchNscTissue and each sample number can be found in the sample column.

The data set provided by the client also includes variables that indicate the physical location of where the measurement was taken within a sample. These are represented the columns row, col, and chamber with the latter being in the form row-col for each respective entry. The possible values of row and col range from one to four. Also, since the experiment was carried out at two locations which is represented by the campagne column.

Changes made to the variables in the original data set

Note there were a couple of problems with the original data set. Initially the time column included a seventh time, A'. Since this did not follow the format of the other time points and had substantially fewer occurrences in the data, we assumed this was a mistake. Therefore, we manually changed all occurrences of A' to A.

The other potential issue was in the chamber column. As stated above this column should be a combination of row and col, but the original data set was treating it as a date. For example if one sample has the values row = 1 and col = 4, the result of chamber should be 1-4. Instead the original data set was showing January 4th. We chose to manually change this to the correct format as well.

Summary Statistics

While some of the variables outlined above are numeric, most can be treated as categorical. The lone exception to this is the starch content. The table below shows some summary statistics for the starch content. This includes not only the summaries of all 408 measurements, but also the summaries based on the two values of campagne and dayPeriod.

Group	N	Mean	Median	SD	Min	Max
Overall	408	1.924902	1.429527	1.733284	0.0191182	7.898429
campagne: 1	184	1.340544	1.245685	1.008316	0.0191182	6.480553
campagne: 2	224	2.404911	1.677605	2.033619	0.2029488	7.898429
dayPeriod: Day	280	1.895429	1.357646	1.730086	0.0191182	7.898429
dayPeriod: Night	128	1.989375	1.483575	1.745326	0.0656625	7.537576

Figure 1: Summary statistics of starch content.

For starch contents across all measurements, the values range from about 0.019 to 7.898 with a median of roughly 1.430 and a mean of 1.925. The location of the median and mean with respect to the minimum and maximum is an early sign that the starch contents could be skewed and thus non-normal in distribution.

When comparing the two locations (campagne) where the experiment was replicated, we can see the 184 measurements from the first location seems to have lower values on average than the 224 measurements from location 2. There is a smaller difference in these metrics when comparing measurements taken in the day versus those taken in the night. Note over twice as many measurements were taken in the day.

To generate a table of summary statistics that account for more of the variables see Appendix A - R Code. That table is not included here due to its larger size.

As previously noted, the table above indicates the starch contents may be skewed and thus non-normal. This can be evaluated through a histogram and Q-Q plot. The histogram below supports our suspicion that the data is skewed and the Q-Q plot confirms the measure is non-normal. Note, all 408 measurements of starch content are used in the plots.

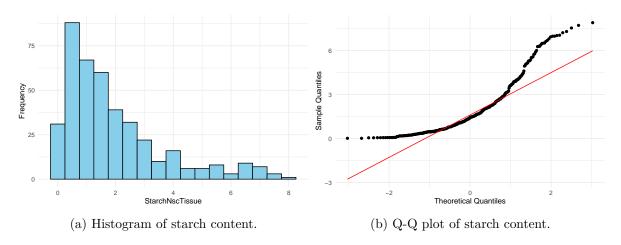


Figure 2: Plots used to check normallity assumption.

Relationships among variables

EXPLORE TRT AND TISSUE TYPES

Now let's see how some of the other variables relate to the starch content. First we can look at the four tissue types. To do this we will use the boxplot below. It appears the tissue types END and IT are similar to each other, as are LM and UM. The two pairs seem quite a bit different though as LM and UM have both far higher values than the other two. This indicates the tissue type could be significant.

Another variable of that could have a major impact is the treatment. If some samples get more water than others it would make sense to see more growth. It is also possible that the time could impact the effect the water has on the starch content. Below is a bar chart that separates measurements first by day and night, and then by the treatment while still showing

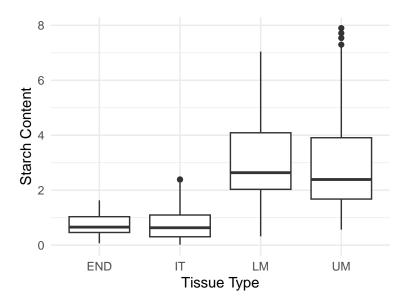


Figure 3: Boxplot of starch contents by tissue types

the differences in tissue type. Remember time points C and D are at night and the rest are during the day.

In the graph above we can see the starch content for measurements with the tissue types LM and UM are higher when given the control treatment instead of the drought treatment. This is not as clear with the other two tissue types. Additionally, the effect day and night have on the starch contents are not clear, as we saw in the summary statistics table above.

Potential models

The replication mentioned previously suggests a mixed model approach is needed. This is due to the replication being a random effect. The simplest case of this type of model is a linear mixed model. To use this, the residuals of the model must be approximately normally distributed.

How explanatory variables can be used

(talk about nesting vs non-nesting methods I guess. Just introduce the idea before we actually make the models.)

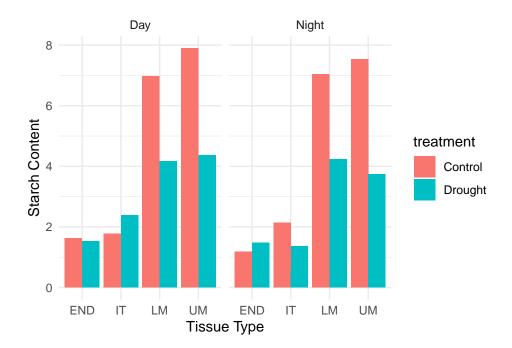


Figure 4: Barchat of starch content vs. tissue types, separating by treatment and day or night.

Model: Mixed Effects Model with Interactions

The first model we want to consider is a linear mixed model with fixed effects treatment, tissue type, and the period of the day, along with random effects for the larger location (campagne), the sample specific location (chamber), and the sample itself. Additionally, this model includes interaction terms for the fixed effects. This can be expressed as

$$y_{ijklm} = \mu + \tau_i + \alpha_j + \beta_k + (\tau\alpha)_{ij} + (\tau\beta)_{ik} + (\alpha\beta)_{jk} + (\tau\alpha\beta)_{ijk} + u_l + v_m + w_n + \epsilon_{ijklm}$$

where y_{ijklm} represents the starch content, μ is the overall mean, τ_i is the fixed effect for the ith treatment, α_j is the fixed effect for the jth tissue type, and β_k is the fixed effect for the period of the day. For the random effects u_l is the effect for the campagne variable, v_m is the effect for chamber, and w_n is the effect for the sample. The error term is ϵ_{ijklm} . The remaining terms represent the interaction between the fixed effects. For instance $(\tau \alpha)_{ij}$ is the interaction effect of the treatment and tissue type, while $(\tau \alpha \beta)_{ijk}$ represents the three-way interaction of all fixed effects in the model.

The model was applied in SAS and all code can be found in *Appendix B - SAS Code*. The figure below shows three tables that are a part of the SAS output. The *Fit Statistics* tables suggests we have a reasonably fitting model. Note these values can also be used for comparison later.

	Covaria	nce Param	eter Es	tim	ates		
	Cov Par				nate		
	campag	ne	1	.75	E-18		
	chambe	r		0.1	1694		
	sample		4	1.89	BE-6		
	Residua	il		0.9	9277		
		Fit Statis	4!				
	2 Doe I	og Likelih		11	50.3		
		aller is Be			56.3		
	•	maller is B			56.3		
		aller is Be		11	52.4		
	Type 3	B Tests of F	ixed E	ffec	ts		
Effect		Num DF	Den	DF	F Va	lue	Pr > F
treatment		1	3	86	6	.26	0.0128
tissu		3	3	86	172	.71	<.0001
treatment*t	issu	3	3	86	13	.06	<.0001
dayPeriod		1	3	86	2	.94	0.0874
treatment*c	•	1	3	86		.18	0.6731
tissu*dayPe	eriod	3	3	86	2	.14	0.0950
acca dayi							

Figure 5: SAS output of COvariance Parameter Estimates, Fit Statistics, and Type 3 Tests of Fixed Effects for the first proposed model

The Type 3 Tests of Fixed Effects reports what fixed effects are registering as significant. With p-values less than 0.0001 both the tissue and the treatment by tissue interaction are highly significant. The treatment effect on its own is still significant at a significance level of 5%. The day period and its interaction with the tissue type are marginally significant, but neither are at the 5% level. The remaining interactions are not significant either.

Conclusion

The model suggests that **tissue type and treatment** are the primary factors affecting starch content, with a significant interaction between them. The **day period** does not appear to play a substantial role. For practical recommendations, you might focus on the main effects and the interaction between treatment and tissue type.

• Treatment and tissue type have significant effects on starch content, with tissue type having the strongest effect.

- The interaction between treatment and tissue is also significant, indicating that the impact of treatment on starch content depends on the type of tissue.
- DayPeriod shows a weak effect, and interactions involving dayPeriod are not significant at the 5% level.

These results suggest that the model effectively captures differences in starch content across treatments and tissue types, with some minor time-of-day effects. Further exploration could involve focusing on treatment and tissue type differences, as well as investigating any practical relevance of the marginal effects of dayPeriod.

				Least S	quares Mea	ıns					
Effect	treatment	tissu	dayPeriod	Estimate	Standard Error	DF	t Value	Pr > t	Alpha	Lower	Upper
treatment	Control			1.3348	0.3383	386	3.95	<.0001	0.05	0.6698	1.9999
treatment	Drought			0.5624	0.3394	386	1.66	0.0983	0.05	-0.1048	1.2297
dayPeriod			Day	0.8603	0.3036	386	2.83	0.0048	0.05	0.2634	1.4573
dayPeriod			Night	1.0369	0.3083	386	3.36	0.0008	0.05	0.4308	1.6431
tissu		END		-0.2229	0.3145	386	-0.71	0.4788	0.05	-0.8412	0.3954
tissu		IT		-0.2106	0.3145	386	-0.67	0.5035	0.05	-0.8288	0.4077
tissu		LM		2.2571	0.3145	386	7.18	<.0001	0.05	1.6389	2.8754
tissu		UM		1.9708	0.3145	386	6.27	<.0001	0.05	1.3526	2.5891

Figure 6: Fig-2

Interpretation:

The Least Squares Means analysis shows that:

- 1. **Treatment Effect**: Starch content is significantly higher in the control group than in the drought group, with only the control group showing a statistically significant mean.
- 2. **DayPeriod Effect**: Both day and night collection periods show significant starch content, with slightly higher values at night.
- 3. **Tissue Effect**: LM and UM tissues have significantly higher starch content than END and IT tissues, which do not differ significantly from zero.

In summary, **tissue type**, **treatment**, **and collection period** all affect starch content, with **LM and UM tissues** and the **control treatment** showing notably higher levels.

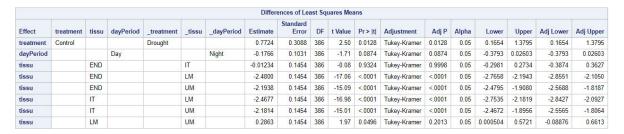


Figure 7: Fig-3

Interpretation:

The **Differences of Least Squares Means** table shows pairwise comparisons for the fixed effects in the model, with Tukey-Kramer adjustments for multiple comparisons. Here's an interpretation of the results for each factor:

Treatment Effect (Control vs. Drought):

There is a significant difference in starch content between the Control and Drought treatments, with the Control group having a higher mean starch content.

DayPeriod Effect (Day vs. Night)

The difference between Day and Night is not statistically significant at the 0.05 level, though it is close. This suggests that the time of collection (Day vs. Night) does not have a significant effect on starch content.

Tissue (Tissu) Effects

- Pairwise Comparisons:
 - END vs. IT: No significant difference (p = 0.9934).
 - END vs. LM: Significant difference with LM having a higher mean starch content than END (p < 0.0001).
 - END vs. UM: Significant difference with UM having a higher mean starch content than END (p < 0.0001).
 - IT vs. LM: Significant difference with LM having a higher mean starch content than IT (p < 0.0001).

- IT vs. UM: Significant difference with UM having a higher mean starch content than IT (p < 0.0001).
- LM vs. UM: No significant difference (p = 0.2013).

• Interpretation:

- LM and UM tissues have significantly higher starch content compared to END and IT tissues.
- There is no significant difference between **END** and **IT**, nor between **LM** and **UM**.

Summary of Findings

- 1. **Treatment**: Control has significantly higher starch content than Drought.
- 2. **DayPeriod**: No significant difference in starch content between Day and Night.
- 3. **Tissue**: Significant differences are present between groups, with LM and UM tissues having higher starch content than END and IT tissues. There is no significant difference within END vs. IT or within LM vs. UM.

In summary, **treatment** and **tissue type** are the main drivers of differences in starch content, with **Control** and the **LM/UM tissues** having higher values. The effect of **DayPeriod** is not significant.

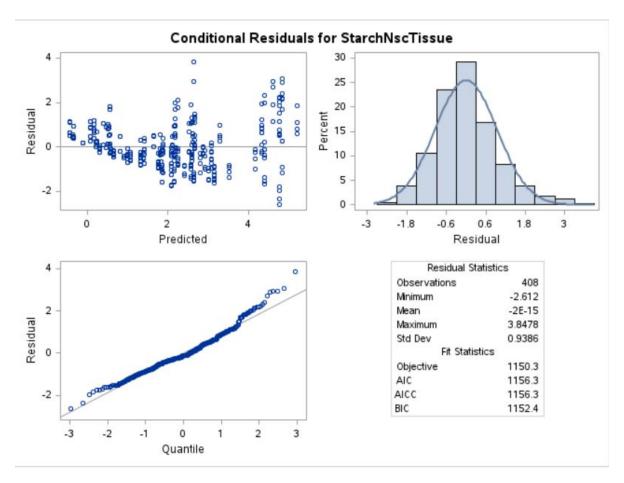


Figure 8: Fig-4

Interpretation:

The **Conditional Residuals** plot provides diagnostic checks for the model fit by examining residuals. Here's an interpretation based on the different plots:

1. Residuals vs. Predicted Plot (Top Left):

This plot shows the residuals plotted against the predicted values. Ideally, residuals should be randomly scattered around zero without any discernible pattern.

• In this plot, some spread appears to increase with the predicted values, suggesting potential heteroscedasticity (non-constant variance), as the residuals fan out slightly for higher predicted values.

2. Histogram of Residuals (Top Right):

This histogram with an overlaid normal curve shows the distribution of residuals.

• The residuals appear approximately normally distributed, but there is some skewness in the tails, particularly on the right. This suggests that the normality assumption is reasonably met, though not perfectly.

3. Q-Q Plot (Bottom Left):

This quantile-quantile (Q-Q) plot compares the distribution of residuals to a theoretical normal distribution.

• The residuals largely follow the straight line, especially in the middle, indicating approximate normality. However, there are deviations at the tails (especially the upper tail), which indicates some potential outliers or non-normality in the extremes.

4. Residual Statistics (Bottom Right):

The summary statistics indicate the mean residual is close to zero, and the standard deviation (0.9386) suggests moderate variability in residuals.

Summary

- Normality: The residuals are approximately normally distributed, though there are some deviations in the tails.
- **Homoscedasticity**: There may be slight heteroscedasticity, as seen in the residuals vs. predicted plot, where residual variance increases with predicted values.
- Model Fit: The model appears reasonably well-fitted, though a closer look at the tails or any potential outliers may improve the model further.

Overall, the residual diagnostics suggest that the model mostly meets the assumptions of normality and homoscedasticity, but some minor deviations could be investigated further if they impact model accuracy.

Interpretation

##Nested Model

Output

Covaria	nce Param	eter E	stim	ates	
Cov Parm	n		Est	imate	
campagn	e		0	.5207	
chamber((campagne)	0	.2477	
sampl(ca	mpag*char	nbe)	0.0	00819	
Residual			0	.9277	
-				100	
	Fit Statis	stics			
-2 Res L	og Likeliho	bod	11	51.9	
AIC (Sm	naller is Bet	tter)	11	59.9	
AICC (S	maller is B	etter)	11	60.0	
BIC (Sm	naller is Bet	tter)	11	54.6	
Туре	3 Tests of F	ixed E	Effec	ts	
Effect	Num DF	Den	DF	F Value	Pr > F
treatment	1		386	4.38	0.0371
tissu	3	- 5	386	172.72	<.0001
treatment*tissu	3	1	386	13.06	<.0001
dayPeriod	1	ŝ	386	2.93	0.0877
treatment*dayPeriod	1	1	386	0.17	0.6823
tissu*dayPeriod	3	5	386	2.14	0.0950
treatm*tissu*dayPeri	3		386	0.45	0.7153

Figure 9: Fig-1

Interpretation:

1. Covariance Parameter Estimates

- campagne: The estimated variance due to campagne is 0.5207, indicating that differences between locations (campagne) contribute to the overall variance in starch content.
- chamber(campagne): The estimated variance due to chamber nested within campagne is 0.2477, suggesting that variation between chambers within each location also affects starch content.
- sample(campagne*chamber): The estimated variance due to sample nested within campagne and chamber is 0.000819, which is relatively small, implying limited variability due to individual samples within chambers.

• **Residual**: The residual variance is 0.9277, which represents the unexplained variability after accounting for fixed effects and random effects.

2. Fit Statistics

The value of the AIC, AICC, and BIC are comparatively higher than the mixed effect model.

3. Type 3 Tests of Fixed Effects

This table tests the significance of each fixed effect and their interactions.

- Treatment (p = 0.0371): Significant at the 0.05 level, indicating that the treatment effect (Control vs. Drought) has a statistically significant impact on starch content.
- Tissu (p < 0.0001): Highly significant, suggesting that different tissue types have a strong effect on starch content.
- Treatment*Tissu Interaction (p < 0.0001): Significant, indicating that the effect of treatment on starch content varies by tissue type.
- DayPeriod (p = 0.0877): Not significant at the 0.05 level, implying that the collection time (Day vs. Night) does not have a significant impact on starch content.
- Treatment*DayPeriod Interaction (p = 0.6731): Not significant, suggesting that there is no interaction between treatment and day period.
- Tissu*DayPeriod Interaction (p = 0.0950): Marginally significant, indicating a potential interaction between tissue type and day period, but it is not below the 0.05 significance level.
- Treatment *Tissu*DayPeriod Interaction (p = 0.7153): Not significant, indicating that there is no three-way interaction among treatment, tissue type, and day period.

Summary of Findings

- 1. **Significant Effects**: Treatment and tissue type are significant main effects, with a significant interaction between them. This suggests that starch content varies by treatment and tissue type, with the impact of treatment depending on the type of tissue.
- 2. Non-Significant Effects: DayPeriod does not have a significant effect, and there are no significant interactions involving DayPeriod with treatment or tissue type.
- 3. Random Effects: Variability due to campagne and chamber within campagne are notable, while the sample-level variability is minimal.

In conclusion, **treatment** and **tissue type** are the primary factors affecting starch content, with the interaction indicating that the effect of treatment depends on tissue type. The time of collection (Day vs. Night) does not significantly affect starch content in this model.

				Least S	quares Mea	ns					
Effect	treatment	tissu	dayPeriod	Estimate	Standard Error	DF	t Value	Pr > t	Alpha	Lower	Upper
treatment	Control			2.2854	0.5725	386	3.99	<.0001	0.05	1.1598	3.4110
treatment	Drought			1.5169	0.5726	386	2.65	0.0084	0.05	0.3911	2.6427
tissu		END		0.7296	0.5496	386	1.33	0.1851	0.05	-0.3509	1.8101
tissu		IT		0.7420	0.5496	386	1.35	0.1778	0.05	-0.3385	1.8225
tissu		LM		3.2097	0.5496	386	5.84	<.0001	0.05	2.1292	4.2902
tissu		UM		2.9234	0.5496	386	5.32	<.0001	0.05	1.8429	4.0039
dayPeriod			Day	1.8129	0.5430	386	3.34	0.0009	0.05	0.7454	2.8805
dayPeriod			Night	1.9894	0.5465	386	3.64	0.0003	0.05	0.9149	3.0639

Figure 10: Fig-2

Interpretation:

Summary

- 1. **Treatment**: Control condition has higher starch content than Drought, and both are significantly different from zero.
- 2. **Tissue**: LM and UM tissues have significantly higher starch content compared to END and IT, which do not show significant starch content.
- 3. **DayPeriod**: Both Day and Night periods show significant starch content, with a slightly higher mean during the Night.

In summary, treatment, tissue type, and collection time all influence starch content, with the Control treatment, LM and UM tissues, and Night period showing higher values.

							Differ	ences of Le	ast Sq	uares Mea	ins							
Effect	treatment	tissu	dayPeriod	_treatment	_tissu	_dayPeriod	Estimate	Standard Error	DF	t Value	Pr > t	Adjustment	Adj P	Alpha	Lower	Upper	Adj Lower	Adj Upper
treatment	Control			Drought			0.7685	0.3673	386	2.09	0.0371	Tukey-Kramer	0.0371	0.05	0.04639	1.4905	0.04639	1.4905
tissu		END			IT		-0.01234	0.1453	386	-0.08	0.9324	Tukey-Kramer	0.9998	0.05	-0.2981	0.2734	-0.3874	0.3627
tissu		END			LM		-2.4800	0.1453	386	-17.06	<.0001	Tukey-Kramer	<.0001	0.05	-2.7658	-2.1943	-2.8551	-2.1050
tissu		END			UM		-2.1938	0.1453	386	-15.09	<.0001	Tukey-Kramer	<.0001	0.05	-2.4795	-1.9080	-2.5688	-1.8187
tissu		IT			LM		-2.4677	0.1453	386	-16.98	<.0001	Tukey-Kramer	<.0001	0.05	-2.7535	-2.1819	-2.8427	-2.0927
tissu		IT			UM		-2.1814	0.1453	386	-15.01	<.0001	Tukey-Kramer	<.0001	0.05	-2.4672	-1.8956	-2.5565	-1.8064
tissu		LM			UM		0.2863	0.1453	386	1.97	0.0496	Tukey-Kramer	0.2013	0.05	0.000505	0.5721	-0.08876	0.6613
dayPeriod			Day			Night	-0.1764	0.1031	386	-1.71	0.0877	Tukey-Kramer	0.0877	0.05	-0.3791	0.02622	-0.3791	0.02622

Figure 11: Fig-3

Interpretation:

Summary of Findings

- 1. **Treatment**: Control has a significantly higher starch content than Drought.
- 2. **Tissue**: LM and UM tissues have significantly higher starch content compared to END and IT tissues. However, there is no significant difference between END vs. IT or between LM vs. UM.
- 3. DayPeriod: No significant difference in starch content between Day and Night.

In summary, **treatment** and **tissue type** are the main drivers of differences in starch content, with **Control** and **LM/UM tissues** showing higher values. The **DayPeriod** does not significantly impact starch content

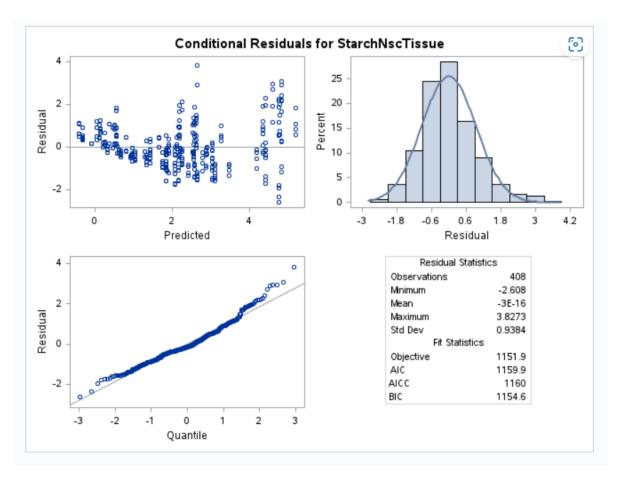


Figure 12: Fig-4

##GLMM Model:

Summary

- Normality: Residuals are approximately normally distributed, as suggested by the histogram and Q-Q plot, though slight deviations are present in the tails.
- **Homoscedasticity**: There appears to be some heteroscedasticity, with residuals showing increasing variance for higher predicted values, as indicated in the residuals vs. predicted plot.
- Model Fit: The model appears reasonably well-fitted overall, though slight adjustments or transformations could be considered if the heteroscedasticity impacts model accuracy.

In conclusion, the model generally meets the assumptions of normality and homoscedasticity, but there are minor deviations that may warrant further investigation, particularly with the slight increase in residual variance at higher predicted values.

Output



Figure 13: Fig-1

Interpretation:

Here's an interpretation of each section in the provided output:

Fit Statistics

These statistics suggest that this model is a reasonable fit and can be compared with other models if needed to find the best balance of fit and simplicity.

Fit Statistics for Conditional Distribution

- -2 log L(StarchNscTissue | r. effects): 812.56 A measure of the fit of the conditional model, where lower values suggest better fit.
- Pearson Chi-Square: 108.96
- Pearson Chi-Square / DF: 0.27 Values near 1 indicate a good fit. A value of 0.27 suggests possible overdispersion (less variation in residuals than expected under the model).

Covariance Parameter Estimates

- **campagne**: Variance component of 0.008996, suggesting low variability attributed to differences between locations (campagne).
- **sample**: Variance component of 0.1116, indicating moderate variability between samples.
- **chamber**: Variance component of 0.02969, indicating minor variability between chambers.
- **Residual**: Variance component of 0.2664, representing the unexplained variability after accounting for the fixed effects and random effects.

The random effects sample and chamber show some variability, with sample contributing the most, whereas campagne has minimal variance. The residual variance is relatively small.

Summary of Findings

- 1. **Significant Effects**: Tissue type (tissu) has a strong effect on starch content, with a significant interaction between tissue type and treatment, meaning that the effect of treatment varies depending on the tissue type.
- 2. **Non-Significant Effects**: Treatment alone, day period, and most interactions involving day period do not significantly affect starch content.
- 3. Random Effects: The sample-level variance is notable, while location (campagne) and chamber-level variances are relatively small. The residual variance is moderate.

In summary, **tissue type is the primary factor** influencing starch content, with a significant interaction indicating that **treatment effects depend on the tissue type**. Day period and interactions involving day period are not significant in this model.

						treatm	ent Lea	st Square	s Means					
			treatmen	t Estimate		ndard Error	DF	t Value	Pr > t	Alpha	Lower	Uppe	er	
			Control	0.4398	(0.2033	386	2.16	0.0312	0.05	0.03987	0.839	4	
			Drought	0.2154	(0.2034	386	1.06	0.2902	0.05	-0.1845	0.615	3	
									Squares ons: Tuk		er			
treatment	_treatme	ent E	stimate	Standard E	rror	DF	t Value	Pr > It	Adj P	Alpha	Lowe	r Upp	er Adj Low	er Adj Upper
Control	Drought		0.2242	0.1	2715	386	0.83	0.4095	0.4095	0.0	-0.309	7 0.75	81 -0.30	97 0.7581
						dayPer	iod Lea	st Squar	es Means					
		d	layPeriod	I Estimate		ndard Error	DF	t Value	Pr > t	Alpha	Lowe	r Upp	er	
)ay	0.3012	0).1521	386	1.98	0.0483	0.05	0.00219	8 0.60	02	
		N	light	0.3539	0).1557	386	2.27	0.0236	0.05	0.0476	8 0.66	00	
				п	ifferen	nces of	davPe	riod Least	t Squares	Means				
				Adju		nt for M	lultiple	Comparis	ons: Tuk	ey-Kram				
	_dayPeri		stimate	Adju Standard E	rror	DF	t Value	Comparis Pr > t	ons: Tuke	ey-Kram Alpha	Lowe		per Adj Lo	
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•				Adju Standard E 0.09	5553	DF 388	t Value	Comparis Pr > t	ons: Tuke Adj P 0.3436	ey-Kram Alpha	Lowe			
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			0.05266 tissu END	Standard E 0.09 Estimate -0.3593	Standa En 0.15	tissi ard ror 588 3	t Value -0.95 u Least DF t \(\)	Comparis Pr > It 0.3436 Squares Value P -2.26 02.70 0.	Adj P 0.3436 Means r > t A 0242 0073	Alpha	Lower 0.6715 -	8 0.05 Upper 0.04710		
			tissu END	Adju Standard E 0.09 Estimate -0.3593 -0.4292	Standa En 0.15	tissi ard ror 588 3 591 3	t Value -0.95 u Least DF t 86	Comparis Pr > It 0.3436 Squares Value P -2.26 02.70 0. 6.98 <	Adj P 0.3436 Means r > t A 00242 00073 00001	ey-Kram 2 Alpha 3 0.05 4 0.05 -4 0.05 -4 0.05 -4 0.05	Lower 0.6715 - 0.7420	Upper 0.04710 -0.1164		
			tissu END IT	Adju Standard E 0.00 Estimate -0.3593 -0.4292 1.1093	Standa En 0.15 0.15	tissi ard ror 588 3 591 3	t Value -0.95 u Least DF t1 86 86	Comparis Pr > It 0.3436 Squares Value P -2.26 02.70 0. 6.98 <	Adj P 0.3436 Means r > t A 00242 00073 00001	ey-Kram 2 Alpha 3 0.05 4 0.05 -4 0.05 -4 0.05 -4 0.05	Lower 0.8715 - 0.7989	Upper 0.04710 -0.1164 1.4216		
			tissu END IT	Adju Standard E 0.00 Estimate -0.3593 -0.4292 1.1093 0.9894	Standa En 0.15 0.15 0.15	tissand ror	ultiple t Value -0.95 u Least DF t1 886 86 86	Comparis	Adj P 0.3436 Means r > It Al 0.242 0.001	Alpha 0.05	Lower 0.6715 - 0.7420 0.7969 0.6770	Upper 0.04710 -0.1164 1.4216		
ау	Night	-	tissu END IT LM UM	Adju Standard E 0.00 Estimate -0.3593 -0.4292 1.1093 0.9894	Standa En 0.15 0.15 0.15	tissuard ror	ultiple t Value -0.95 u Least DF t 1 86 86 86 86 ultiple	Comparis	Adj P 0.3436 Means r > t Al 0242 00073 00001 00001 00001 00001 00001 00001 00001 00001 00001 00001 00001 00001 00001 00001 00001 0000001 0000001 00000001 00000000	Alpha	a Lower 5 -0.161 Lower 0.8715 - 0.7420 0.8770	Upper 0.04710 -0.1164 1.4216 1.3018	352 -0.1	818 0.0565
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END	_tissu IT LM	Estima 0.089	tissu END IT LM UM Stan Stan 92 86 87 85	Adju Standard E 0.00 Estimate -0.3593 -0.4292 1.1093 0.9894 Adju dard Error 0.07885 0.07889	Standa En 0.18 0.18 0.18 0.18 0.18 0.18	tissississississississississississississ	-0.95 t Value -0.95 valu	Pr > tt Squares	Adj P Algorithms	Alpha	Lower 0.6715 - 0.7420 0.6770 0.6770 0.8472 1.6233 - 1.5037 - 1.6952 - 0.855	Upper 0.04710 -0.1164 1.4216 1.3018 Upper 0.2246 1.3139	Adj Lower -0.1330 -1.8718	Adj Upper 0.2729 -1.2858

Interpretation:

Summary of Findings

- 1. **Treatment**: There is no significant difference between Control and Drought treatments on starch content, although the Control group alone shows a significant mean effect.
- 2. **DayPeriod**: Both Day and Night periods individually have significant effects, but there is no significant difference between them.

3. **Tissue (Tissu)**: LM and UM tissues have significantly higher starch content compared to END and IT. However, there is no significant difference between END vs. IT or between LM vs. UM.

In summary, **tissue type** is the primary factor influencing starch content, with **LM and UM showing higher values**. The **DayPeriod** and **Treatment** effects are individually significant, but the comparisons between levels do not show substantial differences.

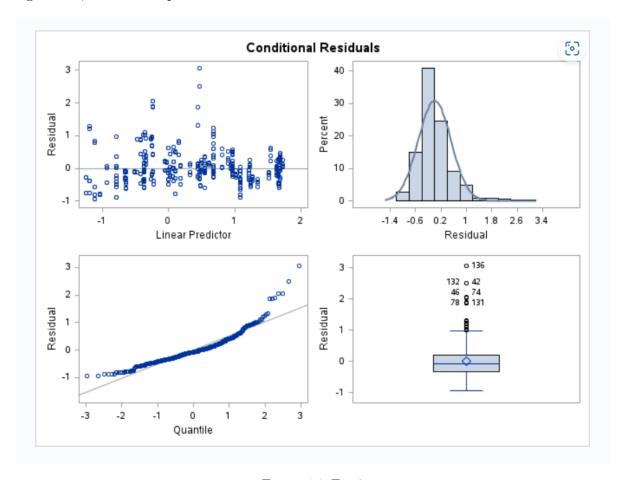


Figure 14: Fig-3

###Interpretation:

The **Conditional Residuals** plot provides diagnostic checks to evaluate the model's assumptions.

1. Residuals vs. Linear Predictor (Top Left)

This plot displays residuals against the linear predictor (fitted values). Ideally, residuals should be randomly scattered around zero with no discernible patterns.

• In this case, the residuals appear fairly well-scattered, but there is some minor clustering around zero, suggesting that the residuals are mostly unbiased but may have slight deviations. No obvious pattern indicates that the assumption of homoscedasticity (constant variance) is mostly met.

2. Histogram of Residuals (Top Right)

This histogram shows the distribution of residuals with an overlaid normal curve.

• The residuals appear approximately normally distributed, though there is some slight skewness, particularly on the right tail. This indicates that the normality assumption is reasonably met, but there may be a few outliers affecting the distribution.

3. Q-Q Plot of Residuals (Bottom Left)

The Q-Q plot compares the residuals to a theoretical normal distribution. Points should ideally lie along the straight line if the residuals are normally distributed.

• Most points fall along the line, indicating approximate normality, although there are deviations at the upper tail. This suggests that while most residuals are normally distributed, a few larger values deviate from normality, indicating possible outliers.

4. Boxplot of Residuals (Bottom Right)

The boxplot provides a summary of the residuals, showing the median, quartiles, and potential outliers.

A few outliers are labeled and extend beyond the upper whisker. While the bulk of the
residuals fall within a reasonable range, these outliers indicate that some data points do
not fit the model as well as others.

Summary

- **Normality**: The residuals are approximately normally distributed, as indicated by the histogram and Q-Q plot, though there are minor deviations in the upper tail.
- **Homoscedasticity**: The residuals vs. linear predictor plot does not show any strong patterns, suggesting that the assumption of constant variance is reasonably met.
- Outliers: The boxplot and Q-Q plot show a few outliers, which may slightly affect the model fit but do not indicate severe violations of assumptions.

Overall, the model diagnostics suggest that the assumptions of normality and homoscedasticity are mostly met, with minor deviations due to a few outliers. The model appears to fit the data reasonably well, although addressing or investigating the outliers could further improve model performance.

##Best Model to fit Based on the above discussion, we would like to fit the **Mixed Model** for this data set. As we can see from the fit statistics and diagnostic result, the mixed model gives us better fitting than the Nested and GLMM models. In the Hierarchical Nested Model, the AIC, BIC, and AICC are comparatively a little bit higher than the Mixed Model, and residual plots remain the same for both plots. Although the AIC, BICC, and AICC are lower in GLMM than in the Mixed Model, the assumptions hold better in the Mixed Model. So, it would be better to fit a **Mixed Model** to ignore unnecessary complexity in the model structure.

Model 3: Nested Model for DayPeriod and Time Effects In this model, dayPeriod is used as a broader time effect, with time nested within dayPeriod.

This model also includes campagne, sample, and chamber as random effects.

Notes for US CHECK

Few notes- 1. Can we keep the model output from SAS but plots fro R? what do you think? 2. As tissue type and treatment are significant and we exactly know which one is mostly significant, can we make some plots or do anything else for this? 3. Can we do anything else for exploratory Analysis? 4. As Dayperiod is not significant wholly, we didn't do anything about time variable. what's your opinion on this?

Yet to be done- 1. Summary 2. Recommendation 3. References

Conclusion

GitHub page found here.

References

Astley, Rick. 1987. "Never Gonna GIve You Up." 1987. https://r.mtdv.me/videos/6QMWR9vBma.

Appendix A - R Code

```
## Prints code without running it
library(knitr)
data <- read.csv("data.csv")
knitr::kable(head(data), format = 'markdown')</pre>
```

Appendix B - SAS Code

```
/* Reading in csv file */
FILENAME REFFILE '<enter your file path';
PROC IMPORT DATAFILE=REFFILE
    DBMS=CSV
    OUT=data;
    GETNAMES=YES;
RUN;
/* Mixed Model*/
proc mixed data=data method=reml plots=(residualpanel);
    class treatment tissu dayPeriod campagne chamber;
    model StarchNscTissue = treatment|tissu|dayPeriod;
    lsmeans treatment dayPeriod tissu / pdiff=all cl adjust=tukey;
    random campagne chamber sample;
run;
/* Hierarchial Nested Model*/
proc mixed data=data method=reml plots=(residualpanel);
    class treatment tissu dayPeriod campagne chamber sample;
    model StarchNscTissue = treatment | tissu | dayPeriod;
    random campagne chamber(campagne) sample(chamber*campagne);
    lsmeans treatment tissu dayPeriod / pdiff=all cl adjust=tukey;
run;
/* GLMM Model */
proc glimmix data=data method=laplace plots=(residualpanel);
    class tissu treatment dayPeriod campagne sample chamber;
    model StarchNscTissue = tissu|treatment|dayPeriod / dist=gamma;
    random campagne sample chamber;
    lsmeans treatment dayPeriod tissu / pdiff=all cl adjust=tukey;
run;
```