# **Statistical Analysis of Auto Dilution Vs Manual Dilution Process in Inductively Coupled Plasma Spectrometer Tests**

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Abstract: Gregory Poole Equipment Company is the Caterpillar Dealer in eastern North Carolina which sells and services anything that Caterpillar Incorporated manufactures. One value added service provided by this company is fluid analysis. This service consists of 22 separate tests on engine oil, power train lubricants, hydraulic oils, coolants and fuels. One particular test performed in this suite of tests is a plasma spectrometer test which analyzes wear metals in the oil. This test is specifically used to help technicians troubleshoot problems or to assist in scheduling planned maintenance in construction and mining equipment fleets. This particular test is currently the bottleneck in the oil lab and more specifically the dilution sub-process in prepping the oil sample for the plasma spectrometer. To overcome this bottleneck an auto diluter was installed to complement the manual dilution process that is also still in process today. The lab manager questioned if the auto-diluter process is generating the same results as the manual process with regards to the quality control check performed on three different certified known value calibration solutions. This paper analyzes control data results from the two populations of different dilution methods to determine if they are producing the same results.

Keywords: Plasma Spectrometer, Analysis of Variance, Sub-Process, Standard, Oil Analysis, Fluid

### Introduction

Gregory Poole Equipment Company is the Caterpillar dealership in eastern North Carolina providing sales and service to the mining, marine engine, electric power and construction machinery industries. This company promotes a comprehensive portfolio of product support offerings that help customer maintain their equipment fleets. One such service is scheduled fluid analysis. Customers provide 6 fluid ounces of lubricating oil, diesel fuel or coolant and the Fluid Analysis Laboratory will perform 22 diagnostic tests on the particular fluid. Some laboratories perform up to 40 tests (Van Rensselar, 2016). The report generated from these tests are very valuable in troubleshooting machine issues and to help plan and maximize maintenance costs. Figure 1 is a copy of the completed report that the equipment owner receives. It shows test results and builds trends from previous tests shown in the two bottom sections. In addition there are specific recommendations given from a highly trained and certified interpreter that reviews the test data and directs a course of action based on the results (Turis and Kucera, 2016). If the report shows a possible issue, the recommendation may be to drain the oil, change the crankcase filter and resample at 100 hours of operation. If there is a major issue, the recommendation could be to shut the machine down and troubleshoot the area that contains the metal that is elevated. If for example the metal molybdenum is trending high in an engine sample, that particular metal is only found in the turbocharger



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bearings. Therefore the technician may inspect bearing end play or perform maintenance on the turbocharger before it fails catastrophically. Data accuracy and proper interpretation of the test results is extremely important. Having inaccurate readings from the instruments could bias the data creating inaccurate measurements. The inaccurate measurements will confound the interpretation of the data resulting in inaccurate recommendations that could cost the customer thousands of dollars.

One of the most important tests performed is the Plasma Spectrometer Test (PST) which analyzes the amount of wear metal in suspension in the oil (Macian et al., 2012). When an engine is operating and under load, very small particles of bearing material or piston cylinder ring material will wear during the normal engine wear process (Kucera et al., 2016; Chen and Tsui, 2013). As the engine wears over time, the parts per million (ppm) of wear metal in the oil should increase and be roughly the same values at the same service interval of say 250 h of engine operation (Cao et al., 2014). This should remain constant unless there is a change in the operation conditions of the machine. If trends begin to be elevated at the same service meter hour reading there may be a bearing failing or dirt entering the engine causing accelerated wear to the engine components (Bartelmus, 2014). If this information is known early enough, relatively inexpensive maintenance or corrective actions can be taken before a catastrophic failure occurs. It is far less expensive to repair an engine before failure than after the bearing completely

fails ruining the crankshaft or other components (Degeratu *et al.*, 2015; Van Rensselar, 2012). Figure 1 shows these trends over time building a history of wear rates on this particular machine the sample was taken from.

The PST is one of the most important and most complex tests performed in generating this report. Its precision and accuracy is critical in making the proper diagnosis for our customers in guiding them through some very large dollar valued decisions. A very important part of the PST is the dilution sub-process (McDowell, 2015). The oil sample being tested must be diluted with kerosene with a one part sample oil to four parts kerosene ratio before the sample trays can be loaded into the PST machine as shown in Fig. 2. This sub process of diluting the sample with kerosene shown by the process maps in Fig. 3 and 4 is the target of study for this paper.

It was found the PST has the longest cycle time of any of the 22 tests performed in the lab and the dilution process was the sub-process that was creating this issue. With customer sample volume growing at a rate of 15% per year, this issue had to be addressed quickly. For years the lab diluted the oil manually using calibrated pipettes to meter the kerosene into the oil being tested shown in Fig. 5. To speed up this subprocess within the PST an auto-diluter was added to the process to complement the manual process shown in Fig. 6. Now the lab technicians can have the autodiluter working concurrently while they are manually diluting another tray for the PST machine. There are two sample dilution processes.

Wear Netals (ppm)	C2	Fe	Cr	AI	Pb	3n	si	Na	к	Мо	Ni	Ca	Mg	Zn	P	Ba
D180-46097-0201	1	4	0	2	0	0	2	2	0	61	0	1707	330	1054	967	0
D180-46015-0303	٥	1	0	0	0	0	2	2	0	58	0	1683	348	1137	1043	0
D180-45295-0118	٥	3	0	0	0	0	2	2	0	65	0	1921	415	1277	1090	0
D180-45225-0116	0	4	0	0	0	0	2	1	٥	66	0	1888	436	1244	1093	0
Oil Condition / Partiole Count (otimi)	ат	OXI	NIT	SUL	w		A	F	PFC	V100						
Oil Condition / Partiole Count (atiml) D180-46097-0201	9T 13	0XI 12	NIT S	SUL 18	W N		A N	F	PFC 0.37	V100 12.5						
Oil Condition / Parfole Count (stiml) D180-46097-0201 D180-46015-0303	9T 13 9	0XI 12 12	NIT G	<b>SUL</b> 18 17	W N		A N N	F N N	PFC 0.37 0.60	V100 12.5 13.2						
Oil Condition /           Particle Count (ctim)           D180-46097-0201           D180-46015-0303           D180-45295-0118	9T 13 9 11	0XI 12 12 12	NIT 6 6 6	SUL 18 17 18	W N N		A N N	F N N	PFC 0.37 0.60 0.54	V100 12.5 13.2 13.6						

Fig. 1. Sample of finished sample report



Fig. 2. Inductively coupled plasma spectrometer with sample tray loaded



Fig. 3. High level auto dilution process







Fig. 5. Manual process of samples being diluted with kerosene



Fig. 6. Auto-dilution process showing samples being diluted with kerosene

# **Problem Statement**

The lab manager was very concerned that having two dilution processes could bring bias or some unforeseen variation into the results of the PST. It was requested that a formal investigation be initiated to determine if indeed there was differences in the results of the manual and automated dilution sub-processes. An additional concern that was uncovered from the initial analysis was to determine if there is variation being generated by the different laboratory technicians performing the manual dilution process.

#### **Research Questions**

- Are there differences in the mean values between the manual dilution process sample and the autodilution sample for each of the three metal known quality control standards?
- Are there differences in the mean values between the lab technician results in manual dilution process for each of the three metal known quality control standards?

### Hypotheses

The null hypothesis for research question (RQ) 1 states:

- Ho: μauto-dilute = μmanual-dilute
- Ha:  $\mu$ auto-dilute  $\neq \mu$ manual-dilute

The null hypothesis for RQ 2 states:

Ho: μtechnician 1 = μtechnician 2 = μtechnician 3 = μtechnician 4

Ha: μtechnician 1 ≠ μtechnician 2 ≠ μtechnician 3 ≠ μtechnician 4

## Method

To test these hypotheses, quality control data were used from the daily work of the lab technicians. Every day there are on average 8 trays of samples of 48 samples per tray prepared and ran through the PST process. Figure 6 shows the auto-dilution process and how the samples are placed in the trays and diluted with kerosene while Fig. 5 shows the manual version of the same sub-process. In each tray there are three control samples containing a known standard of oil with 100 PPM each of iron (Fe), sodium (Na) and zinc (Zn). These three control standard samples are introduced into the PST instrument as shown in Fig. 2 and the output recorded in ppm. Table 1 shows a sample of this output data in order to understand the file layout. For each tray the method of dilution is noted in the comment column allowing differentiation between the two dilution sub-process methods used. The output data of the standard output measurement is noted in the three results columns highlighted in green. The data collection process took about 1.5 months to accumulate with multiple technicians preforming the tests due to operational necessity. For the technician comparison, the run by column will identify the technician that diluted the samples.

Figure 7-9 are the residual plots from the data for each of the metal standards being tested. The three probability plots show three data sets that appear to be normal for the residuals, except for a couple of outliers, adhere rather closely to the probability line with little deviation.



Fig. 9. Probability plot of Zn residuals



### Boxplot of Fe Results (ppm)













Fig. 12. Boxplot of Zn results

Table 1. Sa	ample of data set						
208		b150	92	95	98	mkn	Auto-dilute
209		b152	96	95	99	mkn	Auto-dilute
210		gp7,23	94	94	98	mkn	Manual-dilute
211		g4	96	95	98	rjb	Auto-dilute
212		g5	81	79	95	rjb	Auto-dilute
213		ct36	92	89	97	cmw	Auto-dilute
214		gp6	100	109	93	WWW	Auto-dilute
215		bl53,35b	117	96	104	mkn	Auto-dilute
216	3/18/2016	gp2	98	99	98	mkn	Manual-dilute
217		gp1	93	94	94	mkn	Manual-dilute
218		ct30,40	95	94	101	mkn	Auto-dilute
219		ct31	102	103	107	mkn	Auto-dilute
220		ct32	111	110	113	mkn	Auto-dilute

Table 2. One way ANOVA for	r Fe Vs dilution meth	od				
One-way ANOVA: Fe Results	(ppm) versus dilution	n method				
Method						
Null hypothesis	All means are equal					
Alternative hypothesis	At least one mean is different					
Significance level	$\alpha = 0.05$					
Equal variances were assumed	for the analysis.					
Factor Information						
Factor	Levels	Values				
Dilution method	2 Auto-dilute, Manual-dilute					
Analysis of Variance						
Source	DF	Adj SS	Adj MS	F-Value	P-Value	
Dilution method	1	6.7	6.702	0.16	0.694	
Error	277	11944.7	43.122			
Total	278	11951.4				
Model Summary						
S	R-sq	R-sq(adj)	R-sq(pred)			
6.56672	0.06%	0.00%	0.00%			
Means						
Dilution method	N	Mean	StDev	95% CI		
Auto-dilute	150	98.100	6.361	(97.045, 99.155)		
Manual-dilute	129	98.411	6.798	(97.273, 99.549)		
Pooled StDev = $6.56672$						
Fisher Pairwise Comparisons						
Grouping Information using th	e fisher LSD method	and 95% confidence				
Dilution method	N	Mean	Grouping			
Manual-dilute	129	98.411	А			
Auto-dilute	150	98.100	А			

The heart of this analysis is the analysis of variance test or ANOVA. For each of the three metal quality control runs the population of manually diluted samples is compared to the population of the auto-diluted samples. Table 2-4 show the outputs of these tests with the P-values and F-statistics needed to make decisions on the acceptance of the hypothesis. Table 5 summarizes the outputs for easier interpretation. In addition these tests are graphically depicted by box plots in Fig. 10-12. To reinforce these results by metal control, all data points were combined into one data set and a two sample t-test was performed. Table 6 shows the results from this test along with the box plot in Fig. 13.

The second part of this analysis is to determine whether different technicians impact the output variation of the manual dilution process. It should be noted that four different laboratory technicians perform this work due to operational constraints and work scheduling. The initials of the technician that performed the quality control check are noted in the run by column of Table 1. ANOVA was performed for each of the three metals quality control data to determine if the lab technician was driving any variation between the sampling results. Table 7 to 9 reflect the ANOVA output of this analysis with Table 10 summarizing it into one table for easier interpretation. The graphical representation using boxplots of the variation between technicians is shown Fig. 14.

Table 3. One way ANOVA	A for Na Vs dilutio	on method					
One-way ANOVA: Na res	ults (ppm) versus	dilution method					
Method							
Null hypothesis	thesis All means are equal						
Alternative hypothesis	At least one r	nean is different					
Significance level	$\alpha = 0.05$						
Equal variances were assur-	med for the analys	sis.					
Factor Information							
Factor	Levels	Values					
Dilution method	2	Auto-dilute, Mar	ual-dilute				
Analysis of Variance							
Source	DF	Adj SS	Adj MS	F-Value	P-Value		
Dilution Method	1	82.3	82.27	1.74	0.189		
Error	277	13132.1	47.41				
Total	278	13214.3					
Model Summary							
S	R-sq	R-sq(adj)	R-sq(pred)				
6.88536	0.62%	0.26%	0.00%				
Means							
Dilution Method	Ν	Mean	StDev	95% CI			
Auto-dilute	150	98.167	7.419	(97.060, 99.273)			
Manual-dilute	129	97.078	6.207	(95.884, 98.271)			
Pooled StDev = $6.88536$							
Fisher Pairwise Compariso	ons						
Grouping Information usin	ng the fisher LSD i	method and 95% confide	ence				
Dilution method	Ν	Mean	Grouping				
Auto-dilute	150	98.167	А				
Manual-dilute	129	97.078	А				
Means that do not share a	letter are significat	ntly different.					
Table 4. One way ANOVA	A for Zn Vs dilutic	on method					
One-way ANOVA: Zn res	ults (ppm) versus	dilution method					
Method							
Null hypothesis	All means are	e equal					
Alternative hypothesis	At least one r	nean is different					
Significance level	$\alpha = 0.05$						
Equal variances were assur	med for the analys	sis.					
Factor Information	j~						
Factor Levels Values							
Dilution Method 2 Auto-di	ilute Manual-dilu	te					
Analysis of Variance	nato, manuar-unu						
Source	DE	14:55	AdiMC	E Volue	D Volue		
Source	Dr	Auj SS	AUJINIS	r-value	r-value		

DF	Adj SS	Adj MS	F-Value	P-Value
1	4.76	4.761	0.16	0.685
277	8024.84	28.971		
278	8029.60			
R-sq	R-sq(adj)	R-sq(pred)		
0.06%	0.00%	0.00%		
Ν	Mean	StDev	95% CI	
150	101.467	5.209	(100.602, 102.332)	
129	101.729	5.577	(100.796, 102.662)	
ons				
ng the fisher LSD 1	nethod and 95% confide	ence		
Ν	Mean	Grouping		
129	101.729	А		
150	101.467	А		
letter are significant	ntly different.			
	DF 1 277 278 R-sq 0.06% N 150 129 ons ng the fisher LSD r N 129 150 letter are significat	DF Adj SS 1 4.76 277 8024.84 278 8029.60 R-sq R-sq(adj) 0.06% 0.00% N Mean 150 101.467 129 101.729 ons ng the fisher LSD method and 95% confide N Mean 129 101.729 150 101.467 letter are significantly different.	DF         Adj SS         Adj MS           1         4.76         4.761           277         8024.84         28.971           278         8029.60         R-sq(pred) $0.06\%$ $0.00\%$ $0.00\%$ N         Mean         StDev           150         101.467         5.209           129         101.729         5.577           ons         Mean         Grouping           129         101.729         A           150         101.467         A           letter are significantly different.         A	DF       Adj SS       Adj MS       F-Value         1       4.76       4.761       0.16         277       8024.84       28.971       0.16         278       8029.60       R-sq(pred)       0.00%         R-sq       R-sq(adj)       R-sq(pred)       0.00%         N       Mean       StDev       95% CI         150       101.467       5.209       (100.602, 102.332)         129       101.729       5.577       (100.796, 102.662)         ons       Mean       Grouping       129         150       101.467       A       150         150       101.729       A       150         150       101.729       A       150         150       101.467       A       150         150       101.467       A         letter are significantly different.       Hean       Grouping

Table 5. Summary table for	ANOVA dilution met	hod by metal			
Metal Control Standard	Adj SS	Adj MS	F-Value	P-Value	Fisher Pairwise
Fe	6.70	6.702	0.16	0.694	A,A
Na	8.30	82.770	1.74	0.189	A,A
Zn	4.76	4.761	0.16	0.685	A,A
Table 6. Two sample T test	for combined date cor	nparing dilution	method		
Two-sample T-test and CI: (	Combined dilution co	mbined			
Two-sample T for combined					
Dilution combined	N	M	ean	StDev	SE mean
Auto-dilute	450	99	.24	6.57	0.31
Manual-dilute	387	99	.07	6.50	0.33
Difference = $\mu$ (Auto-dilute)	- u (Manual-dilute)				
Estimate for difference: 0.17	2				
95% CI for difference: (-0.7	17. 1.061)				
T-Test of difference = $0$ (Vs	$\neq$ ): T-Value = 0.38 P	-Value = $0.704  \text{D}$	0F = 818		
	.,				
Table 7 One way ANOVA	for <b>Zn</b> by technician				
One-way ANOVA: 7n man	versus tech man				
Method	versus teen man				
Null hypothesis	All means are equi	1			
Alternative hypothesis	All means are equa	u is different			
Significance level	At least one mean $\alpha = 0.05$	is unicicilit			
Equal variances were assume	u = 0.05				
Equal variances were assume Eactor Information	cu for the analysis.				
Factor	Lavals	Values			
Tech mon	A LEVEIS	values	31/11/31/		
Analysis of Variance	7	cinw, inkii, ijo,	~~~~~		
Source	DF	Adi 88	Adi MS	F-Value	P-Value
Tech man	3	08 77	32 74	1.05	0.371
Frror	125	3883.28	31.07	1.05	0.571
Total	123	3981 50	51.07		
Model summary	120	5701.50			
S	R-sa	R-sa(adi)	R-sa(pred)		
5 57371	2 47%	0.13%	0.00%		
Means	2.4770	0.1570	0.0070		
Tech					
man	Ν	Mean	StDev	95% CI	
cmw	2	01 50	2 12	(93.70, 109.30)	
mkn	2 94	101 628	5 406	(100 490 102 7	(65)
rib	26	101.15	6.02	(98.99, 103.32)	
www	20 7	105.29	6.50	(101 12 109 46)	)
Pooled StDev = $5.57371$	,	100.29	0.00	(101.12, 10).10	)
Fisher Pairwise Comparison	\$				
Grouping information using	the fisher LSD Metho	od and 95% Conf	idence		
Tech					
man	Ν	Mean	Grouping		
WWW	7	105.29	A		
mkn	. 94	101.628	A		
cmw	2	101.50	A		
rjb	26	101.15	А		
5	-				

Means that do not share a letter are significantly different.

Table 8. One Way ANOVA for Na by technician One-way ANOVA: Na man versus Tech man Method Null hypothesis All means are equal Alternative hypothesis At least one mean is different Significance level  $\alpha = 0.05$ Equal variances were assumed for the analysis. Factor Information Factor Levels Values Tech man 4 cmw, mkn, rjb, www Analysis of Variance Adj MS F-Value P-Value Source DF Adj SS 111.88 Tech man 3 335.6 3.04 0.031 125 4595.6 36.76 Error 128 4931.2 Total Model Summary R-sq(pred) S R-sq R-sq(adj) 6.06340 6.81% 4.57% 1.26% Means Tech Ν StDev 95% CI man Mean 2 94.50 3.54 (86.01, 102.99) cmw mkn 94 96.500 6.097 (95.262, 97.738) 26 97.65 5.93 (95.30, 100.01) rjb www 7 103.43 6.43 (98.89, 107.96) Pooled StDev = 6.06340Fisher Pairwise Comparisons Grouping Information Using the Fisher LSD Method and 95% Confidence Tech man Ν Mean Grouping 7 www 103.43 А В rjb 26 97.65 94 mkn 96.500 В cmw 2 94.50 ΑB Means that do not share a letter are significantly different.





Fig. 13. Boxplot of combined data by dilution method



# Boxplot of Fe, Na, Zn With Manual Dilution



Table 9. One Way ANOV	A for Fe by techn	ician					
One-way ANOVA: Fe ma	n versus Tech ma	n					
Method							
Null hypothesis	All means are	All means are equal					
Alternative hypothesis	At least one r	At least one mean is different					
Significance level	$\alpha = 0.05$						
Equal variances were assu	med for the analy	sis.					
Factor Information							
Factor	Levels	Values					
Tech man	4	cmw, mkn, rjb, v	VWW				
Analysis of Variance							
Source	DF	Adj SS	Adj MS	F-Value	P-Value		
Tech man	3	97.70	32.57	0.70	0.554		
Error	125	5817.52	46.54				
Total	128	5915.22					
Model Summary							
S	R-sq	R-sq(adj)	R-sq(pred)				
6.82204	1.65%	0.00%	0.00%				
Means							
Tech							
man	Ν	Mean	StDev	95% CI			
cmw	2	94.500	0.707	(84.953, 104.047)			
mkn	94	98.138	7.346	(96.746, 99.531)			
rjb	26	98.96	5.48	(96.31, 101.61)			
WWW	7	101.14	2.7	(96.04, 106.25)			
Pooled StDev = 6.82204							
Fisher Pairwise Comparise	ons						
Grouping Information Usi	ing the Fisher LSE	Method and 95% Conf	idence				
Tech							
man	Ν	Mean	Grouping				
WWW	7	101.14	А				
rjb	26	98.96	А				
mkn	94	98.138	А				
cmw	2	94.500	А				
Means that do not share a	letter are significa	ntly different.					

Table 10. Summary of ANOVA table for manual dilution by technician						
Metal Control Standard	Adj SS	Adj MS	F-Value	P-Value	Fisher Pairwise Comparison	
Fe	97.70	32.57	0.70	0.5540	A,A,A,A	
Na	335.60	111.88	3.04	0.0310	A,B,B,AB	
Zn	98.22	32.76	1.05	0.0371	A,A,A,A	

Table 10. Summary	of ANOVA tab	le for manual di	ilution by technician
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### Results

When looking at the comparison of auto-dilution and manual dilution data analysis it is found that the P-values for each of the three metals control data above the 0.05 alpha. The Sodium (Na) P-value was the lowest at 0.189 with iron at 0.694 and zinc at 0.685. This tells us to accept the null that the means of the three control output data sets are the same. When looking at the box plots it is evident graphically that there is very little difference in the two methods of dilution. Based on the P-values it should make no difference in the outcomes of the oil analysis if either the manual or auto-dilution sub-process is used in performing the PST test.

The two sample T-test of the combined data results are also resounding. The P-value for this test is 0.704 and a corresponding T-value of 0.38 which again shows that the null hypothesis that the mean values of the data from the two dilution methods are the same.

With regards to the difference of the output data means by technician there are some interesting findings. The P-values for zinc is 0.371, iron is 0.554 and most interesting sodium is 0.031. This P-value indicates there is a difference in the sodium test means therefore one must reject the null hypothesis in this case only. There is a low P-value and a difference of the means when looking specifically at the manual dilution process depending on which technician is performing the test. However there is a very low number of data points for two of the lab technicians in making this determination and caution should be used and more data is needed to make a firm decision.

### Discussion

From the results it appears that no matter which metal control is being tested, there is no difference in the mean results when comparing manual and auto dilution methods. It does appear there is enough data points in both categories to make this test valid and currently both process are being utilized daily. As mentioned earlier this sub-process is a bottleneck in the entire PST process for the auto-diluter needs an additional 20% capacity to satisfy demand. As business levels continue to grow so will this capacity deficit. Knowing that there is no difference in the dilution process gives our Lab Manager some confidence that his lab can continue to utilize both methods with confidence. This knowledge will aid the Laboratory Manager with some operational decisions within the oil lab. There are new auto-diluters available in the marketplace that are much faster than our current model at a cost of \$148,000.00. These results will assist the manager in deciding if he will continue to utilize manpower with the manual method or purchase the new faster auto-diluter. The manager now knows either method will work and give consistent results.

When looking at the manual process with different technicians, the means do vary by the technician performing the test. It should be noted however that the technician results that were different had a relatively small number of data points to draw from. If there was one improvement to this analysis it would be to have more data points from technicians www and cmw. Even with the low number of data points, more learning could possibly be gained from additional experimentation into why these two technicians have higher readings for the sodium metal test. Given the small number of data point along with the consistency of the other tests, it would be interesting to find out too why only sodium shows this difference between technicians.

Other suggestion for future experimentation would be to perform this same analysis using a different spectrometer in order to see if the results are reproducible (Levinsohn, 1984) and to determine if the brand of oil being analyzed impacts the results of these tests (Markova et al., 2014).

### Conclusion

It can be decided that one must accept the null hypothesis that the mean values of both dilution processes are the same for all three sets of metal quality control data studied. The lab manager can with confidence use both methods to dilute the oil samples and not worry about the outcomes generated by either sub-process. With this new found information the management can make better decisions regarding the future of capital expenditures for lab growth and how to better manage the processes dealing with the PST.

It was also found, with very limited data, that there may be differences within the manual dilution data subset of sodium between technicians. In this case we must reject the null hypothesis for the metal sodium only and accept the null for iron and zinc. There should be further research in this area and more data points gathered to first validate the results with adequate data. It should also be determined why only one of the three metal data subsets had this result. Is this an anomaly or is there truly a difference in the means of the output by technician.

## **Author's Contributions**

**Frederick Rich:** This work was complete as a requirement for a project in the course COT703 advanced statistical analysis at Indiana State University.

**M. Affan Badar:** Ensured the analysis was performed and interpreted correctly. The project work was revised by both the authors in order to match the style of the journal article.

### Ethics

This article is original and contains unpublished material. The corresponding author confirms that all of the other authors have read and approved the manuscript and no ethical issues involved.

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