# Modeling the Useful Residual Life of Railroad Grease

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### Project Description

- The degradation of grease used to lubricate railroad bearings is believed to occur due two processes:
  - Mechanical processes occurring within the bearing,
  - Oxygen diffusion.
- Appropriate lubrication of the bearings is critical during railroad service operation.
- This study focuses on the development of empirical models that can accurately predict the residual useful life of railroad bearing grease.
- Employed Modeling Techniques:
  - Linear Regression Analysis
  - Regression Trees
  - Split Plots

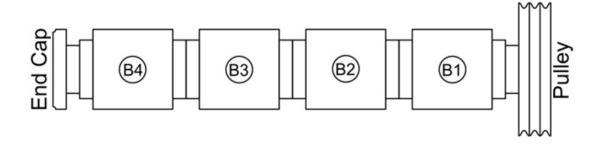
### Project Description (cont.)

 The data set used in the development of the model consists of more than 100 samples of grease taken from the railroad bearings which were observed in a laboratory setting.

### Laboratory Bearing Tester

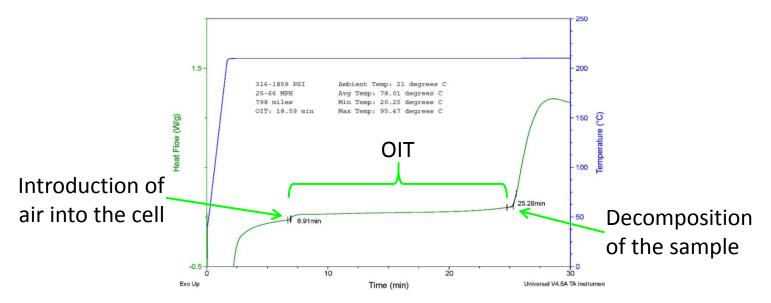
- Four bearings on the axle are subjected to the following experimental variables:
  - Load Conditions
  - Rotational Speed
  - Mileage
  - Temperature





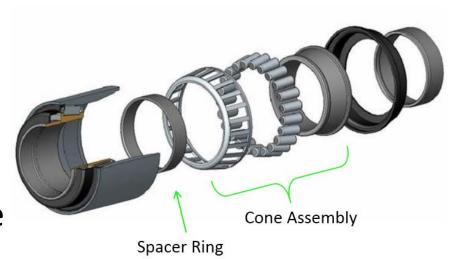
### Oxidation Induction Time

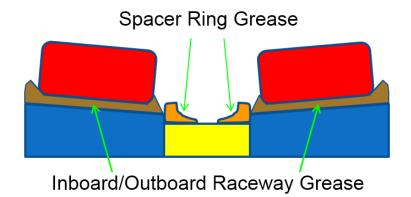
- Oxidation Induction Time (OIT) is a test performed in a Differential Scanning Calorimeter (DSC) which measures the level of thermal stabilizers in the material.
- The DSC produces a graph of heat flow vs time.
- The time elapsed between the introduction of air into the cell and the decomposition of the sample reveals the time to oxidation which is then recorded as OIT.



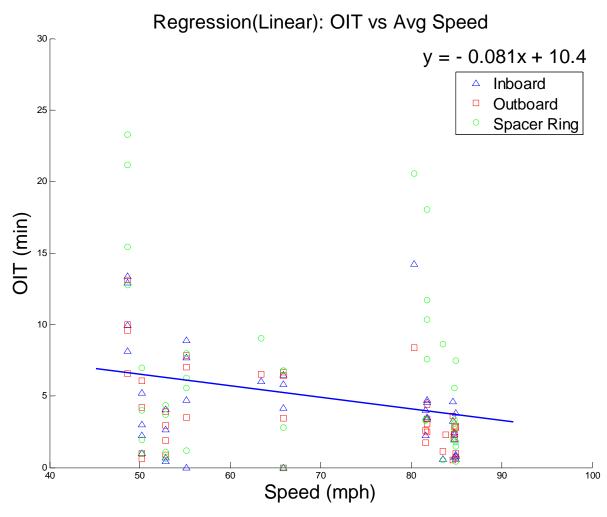
### Bearings

- Three samples come from each bearing, giving a total of twelve possible samples from each axle.
- Grease is sampled from the three critical locations of the bearing:
  - Inboard Cone Assembly Raceway
  - Outboard Cone Assembly Raceway
  - Spacer Ring Area



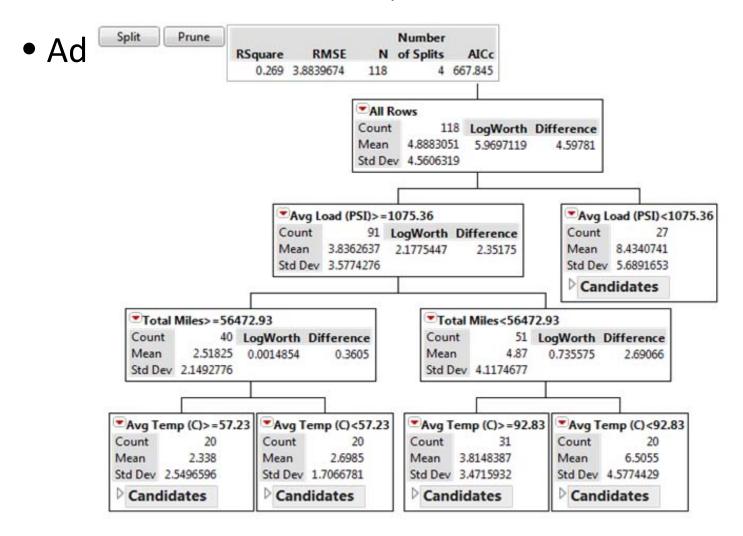


# Linear Regression Plot for OIT vs Speed



### Regression Tree

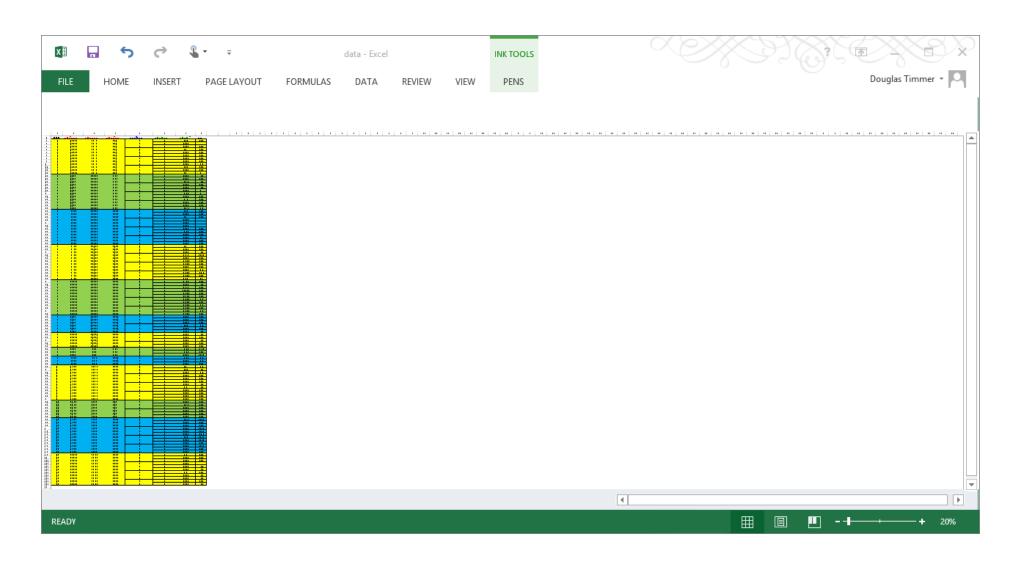
Min size split 20



### Experimental Design

- Split, Split-plot Design
  - Whole plot: axle-setup
  - Sub plot: each bearing on axle
  - Sub, sub plot: sample location within each bearing
- Single replicate
- Unbalanced design

### **Unbalanced Data**



#### Parameter Estimation

- Restricted Maximum Likelihood (REML)
- Implemented in Matlab
- Degrees of Freedom are approximate due to unbalanced data

# Representation of Bearing Location in Regression Model

- The bearing location was recorded as a nominal value (1, 2, 3, 4)
- Modeled using three indicator variables

		<b>Dummy Variables</b>	
Bearing	X4	X5	Х6
1	0	0	0
2	1	0	0
3	0	1	0
4	0	0	1

# Representation of Grease Location in Regression Model

- The grease location was recorded as a nominal value (1, 2, 3)
- Modeled using two indicator variables

	Dummy Variables					
Grease	X7	X8				
1	0	0				
2	1	0				
3	0	1				

### Initial Model

	Term	Coef	se(Coef)	t-statistic	approx p-val			
	Intercept	1.0737	3.7641	0.2853	0.7850		Variance Componer	Estimate
	load	1.5122	7.1931	0.2102	0.8405		wp	15.33
wp terms (approx error df = 6)	mileage	-13.2989	4.3152	-3.0819	0.0216	**	sp	1.87
	speed	1.7393	4.5440	0.3828	0.7151		ssp	5.59
	load*mileage	-9.3852	7.4705	-1.2563	0.2557			
	load*speed	-4.6215	7.1840	-0.6433	0.5438		Variance Ratios	Estimates
	mileage*speed	16.6564	7.5960	2.1928	0.0708	•	eta1	2.742
sp terms	x4	2.8713	0.9978	2.8776	0.0083	**	eta2	0.334
(approx error df =	x5	0.6385	1.0011	0.6378	0.5296			
24)	х6	1.6466	0.9917	1.6604	0.1099			
	х7	2.9221	0.6333	4.6141	0.0000	**		
	x8	-0.3409	0.5432	-0.6276	0.5323			
ssp terms	temperature	-8.7969	3.2085	-2.7417	0.0077	**		
(approx error df =	load*temperature	-3.9283	5.1582	-0.7616	0.4488			
72)	mileage*temperature	-7.8488	3.6546	-2.1476	0.0351	**		
	speed*temperature	2.7801	4.1258	0.6738	0.5026			
Analysis of coded v	ariables							
	Obs	Approx DF	Approx Error	· DF				
WP	13	13	6					
SP	40	27	24					
SSP	118	78	72					

### Model 2

		Term	Coef	se(Coef)		t-statistic	approx p-val		
		Intercept	0.6999		1.5747	0.4445	0.6672	Variance Components	Estimate
wp terms (approx error df = 9)	mileage	-8.0003		3.1156	-2.5678	0.0303 **	wp	13.680	
	(approx error at = 9)	speed	1.6769		1.8711	0.8962	0.3935	sp	2.052
		mileage*speed	6.0180		3.7871	1.5891	0.1465	ssp	5.497
sp terms	(approx error df = 26)	x4	1.9142		0.7417	2.5808	0.0159 **		
		x7	2.9134		0.5294	5.5032	0.0000 **	Variance Ratios	Estimates
ssp terms	(approx error df = 75)	temperature	-6.6029		2.4734	-2.6696	0.0093 **	eta1	2.4885
		mileage*temperature	-6.2692		2.4603	-2.5481	0.0129 **	eta2	0.3733

#### Analysis of coded variables

	Obs	Approx DF	Approx Error DF
WP	13	13	9
SP	40	27	26
SSP	118	78	75

# Model 3

		Term	Coef s	e(Coef)	t-statistic	approx p-val		
wp terms	(approx error df = 11)	Intercept	1.8964	1.3727	1.3815	0.1697	Variance Components	Estimate
		mileage	-3.9511	1.9252	-2.0523	0.0423 **	wp	14.71
sp terms	(approx error df = 26)	x4	1.8173	0.7385	2.4608	0.0208 **	sp	2.08
		x7	2.7981	0.5002	5.5940	0.0000 **	ssp	5.50
sp terms	(approx error df = 75)	temperature	-5.0227	1.9157	-2.6219	0.0106 **		
		mileage*temperature	-3.5885	3.0580	-1.1735	0.2443	Variance Ratios	Estimates
		mileage*temperature	-3.5885	3.0580	-1.1735	0.2443	Variance Ratios eta1	Estimates 2.675
alysis of coc	ded variables	mileage*temperature	-3.5885	3.0580	-1.1735	0.2443		
alysis of cod	ded variables	mileage*temperature  Obs	-3.5885 Approx DF )p			0.2443	eta1	2.675
alysis of coo	ded variables Wi	Obs				0.2443	eta1	2.675
alysis of coo		Obs 13	Approx DF )p	orox Error [		0.2443	eta1	2.675

### Final Model

SSP

118

		Term	Coef	se(Coef)	t-statistic	approx p-val			
wen tarms	( 11)	Intercept	2.3872	1.184	6 2.0152	0.0690		<b>Variance Components</b>	Estimate
wp terms	(approx error df = 11)	mileage	-3.8116	1.728	1 -2.2057	0.0496	**	wp	11.311
sp terms	(approx error df = 26)	x4	1.7551	0.752	1 2.3336	0.0276	**	sp	2.231
		х7	2.7443	0.498	6 5.5040	0.0000	**	ssp	5.599
ssp terms	(approx error df = 76)	temperature	-3.7388	1.548	5 -2.4145	0.0160	**		
								Variance Ratios	Estimates
nalysis of coo	led variables							eta1	2.0202
								eta2	0.3984
		Obs	Approx DF	Approx Error DI					
	W	P 13	13	11					
	SI	P 40	27	26					

76

### Final Model

- $\widehat{OIT} = 2.3872 3.8116 * mileage' + 1.7551 * x_4 + 2.7443 * x_7 3.7388 * temperature'$
- Where
  - $mileage' = \frac{mileage 53396}{45687}$
  - $temperature' = \frac{temperature 80.16}{32.71}$
  - $x_4$  is 1 if bearing 2 location, 0 for other bearing locations
  - $x_7$  is 1 if grease sampling location is the spacer ring and 0 for the inner or outer raceway

### **Future Research**

- Model Diagnostics
  - Residual analysis
  - R^2
  - VIF
- Model Refinement
  - Why is bearing 2 statistically different?
  - Is temperature a covariate (function of load, mileage and speed)?
  - Developing second response variable related to length of grease molecule
  - Alternative Model: neural network or ensemble of neural networks

### Acknowledgements

- University Transportation Center for Railway Safety (<a href="http://www.utrgv.edu/railwaysafety">http://www.utrgv.edu/railwaysafety</a>) for their support of this research
- The Matlab code was provided by Dr. Marcus Perry, Associate Professor of Statistics, Culverhouse College of Commerce, University of Alabama