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Lag Modeling of the Baseball Trade Deadline

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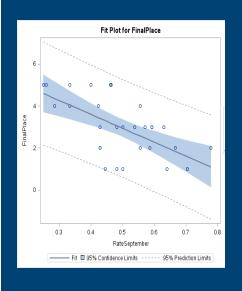
OUTLINE

Baseball's Trade Deadline

Lag Models

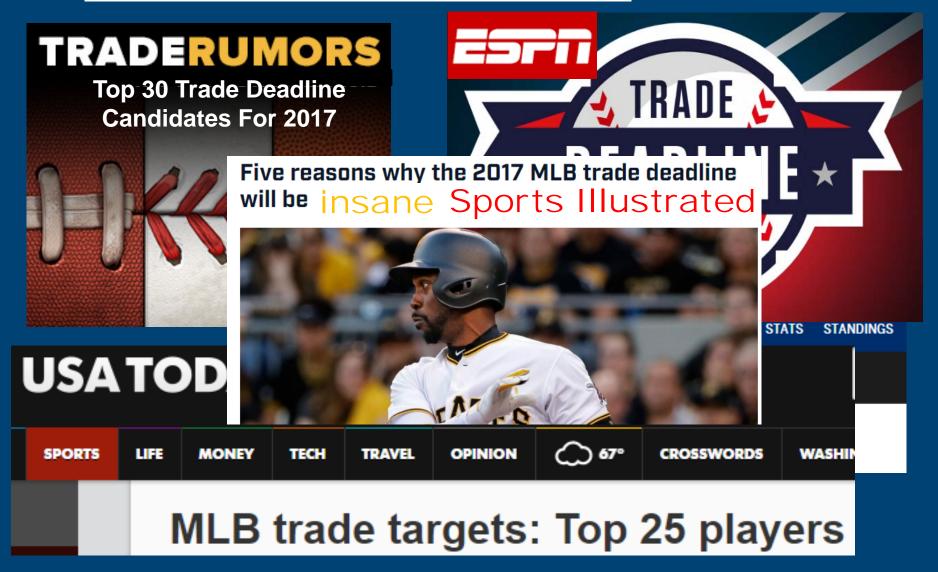


Regressors by Month



Polynomial Distributed Lag
Summary: Trade Up?

Baseball's Trade Deadline



Lag Models – Variables

	Pearson Correlation Coefficients, N = 30								
Prob > r under H0: Rho=0									
	RateApril	RateMay	RateJune	RateJuly	RateAugust	RateSeptember	RateSpring	FinalPlace	
RateApril	1.00000	0.28698	0.15902	0.34627	0.66620	0.35075	0.42484	-0.62485	
		0.1241	0.4013	0.0609	<.0001	0.0574	0.0193		
RateMay	0.28698	1.00000	-0.02411	0.12697	0.27896	0.34268	-0.11647	-0.49392	
	0.1241		0.8994	0.5037	0.1355	0.0638	0.5399		
RateJune	0.15902	-0.02411	1.00000	0.25013	0.06874	0.43940	0.12328	-0.32988	
	0.4013	0.8994		0.1825	0.7181	0.0151	0.5163		
RateJuly	0.34627	0.12697	0.25013	1.00000	0.36815	0.35926	0.07401	-0.60854	
	0.0609	0.5037	0.1825		0.0453	0.0512	0.6975		
RateAugust	0.66620	0.27896	0.06874	0.36815	1.00000	0.17994	-0.01802	-0.68765	
	<.0001	0.1355	0.7181	0.0453		0.3413	0.9247		
RateSeptember	0.35075	0.34268	0.43940	0.35926	0.17994	1.00000	0.14406	-0.62891	
	0.0574	0.0638	0.0151	0.0512	0.3413		0.4476		
RateSpring	0.42484	-0.11647	0.12328	0.07401	-0.01802	0.14406	1.00000	-0.09659	
	0.0193	0.5399	0.5163	0.6975	0.9247	0.4476			
FinalPlace	-0.62485	-0.49392	-0.32988	-0.60854	-0.68765	-0.62891	-0.09659	1.00000	
	0.0002	0.0055	0.0750	0.0004	<.0001	0.0002	0.6116		

Regressors by Month: Just T-1 and T-2

The REG Procedure

Model: MODEL2

Dependent Variable: FinalPlace

Number of Observations Read 30 Number of Observations Used 30

Analysis of Variance								
		Sum of Mean						
Source	DF	Squares	Square	F Value	Pr > F			
Model	2	42.62505	21.31252	37.75	<.0001			
Error	27	15.24162	0.56450					
Corrected Total	29	57.86667						

Root MSE	0.75134	R-Square	0.7366
Dependent Mean	2.93333	Adj R-Sq	0.7171
Coeff Var	25.61369		

Parameter Estimates								
		Parameter	Standard		Pr > t			
Variable	DF	Estimate	Error	t Value	1			
Intercept	1	9.05341	0.71757	12.62	<.0001			
RateSeptember	1	-5.50040	1.05785	-5.20	<.0001			
RateAugust	1	-6.73155	1.13844	-5.91	<.0001			

Regressors by Month: Multicoliniarity

The REG Procedure
Model: MODEL4

Dependent Variable: FinalPlace

Number of Observations Read 30 Number of Observations Used 30

Analysis of Variance								
		Sum of	Mean					
Source	DF	Squares	Square	F Value	Pr > F			
Model	4	45.76889	11.44222	23.65	<.0001			
Error	25	12.09778	0.48391					
Corrected Total	29	57.86667						

 Root MSE
 0.69564
 R-Square
 0.7909

 Dependent Mean
 2.93333
 Adj R-Sq
 0.7575

 Coeff Var
 23.71491
 0.7575

Parameter Estimates							
		Parameter	Standard		Pr > t		
Variable	DF	Estimate	Error	t Value	1		
Intercept	1	9.84333	0.92733	10.61	<.0001		
RateSeptember	1	-4.48195	1.12232	-3.99	0.0005		
RateAugust	1	-5.82807	1.11832	-5.21	<.0001		
RateJuly	1	-2.76288	1.13085	-2.44	0.0220		
RateJune	1	-0.74086	1.73885	-0.43	0.6737		

Regressors by Month: All Months

The REG Procedure Model: MODEL5

Dependent Variable: FinalPlace

Number of Observations Read 30 Number of Observations Used 30

Analysis of Variance								
	Sum of Mean							
Source	DF	Squares	Square	F Value	Pr > F			
Model	7	48.12958	6.87565	15.53	<.0001			
Error	22	9.73709	0.44260					
Corrected Total	29	57.86667						

 Root MSE
 0.66528
 R-Square
 0.8317

 Dependent Mean
 2.93333
 Adj R-Sq
 0.7782

 Coeff Var
 22.67994
 0.7782

Parameter Estimates								
		Parameter	Standard		Pr > t			
Variable	DF	Estimate	Error	t Value	I			
Intercept	1	11.11371	1.22090	9.10	<.0001			
RateSeptember	1	-3.39979	1.18492	-2.87	0.0089			
RateAugust	1	-5.07058	1.52682	-3.32	0.0031			
RateJuly	1	-2.87419	1.08466	-2.65	0.0146			
RateJune	1	-1.47438	1.69895	-0.87	0.3949			
RateMay	1	-2.53199	1.13909	-2.22	0.0368			
RateApril	1	-0.32416	2.00001	-0.16	0.8727			
RateSpring	1	-0.67937	1.74593	-0.39	0.7009			

Finite Distributed Lag: SAS / ETS

Analysis of Variance								
Source	DF	Sum of Squares	Mean Square	F Value	Pr > F			
Model	4	35.97258	8.99315	10.27	<.0001			
Error	25	21.89408	0.87576					
Corrected Total	29	57.86667						

Root MSE	0.93582	R-Square	0.6216
Dependent Mean	2.93333	Adj R-Sq	0.5611
Coeff Var	31.90303		

Parameter Estimates								
Variable	DF	Parameter Estimate	Standard Error	t Value	Pr > t			
Intercept	1	10.54399	1.38856	7.59	<.0001			
RateJuly	1	-4.92295	1.42557	-3.45	0.0020			
RateJune	1	-4.38429	2.21658	-1.98	0.0591			
RateMay	1	-4.68814	1.40730	-3.33	0.0027			
Salary	1	-5.76231E-9	3.994239E-9	-1.44	0.1615			

Polynomial Distributed Lag: SAS / ETS

```
data work.TradeDeadline;
   set tsa.bb_win_rates_2013;
   X = RateJuly;
   X11 = RateJune;
   x12 = RateMay;
   x13 = RateApril;
   X14 = RateSpring;
run;
proc pdlreg
   data=baseball_time_series;
   model finalplace =
   win_rate(3,3) salary / covb;
run;
```

Polynomial Distributed Lag: SAS / ETS

The SAS System

The PDLREG Procedure

Ordinary Least Squares Estimates							
SSE	17.9268035	DFE	19				
MSE	0.94352	Root MSE	0.97135				
SBC	81.9457109	AIC	74.632456				
MAE	0.72912854	AICC	79.2991226				
MAPE	33.4210879	HQC	76.6608421				
Durbin-Watson	2.6689	Total R-Square	0.6013				

Parameter Estimates								
Variable	DF	Estimate	Standard Error	t Value	Approx Pr > t			
Intercept	1	-6.4736	7.4093	-0.87	0.3932			
X**0	1	8.3016	6.0082	1.38	0.1831			
X**1	1	4.8543	1.9866	2.44	0.0245			
X**2	1	-3.6729	1.9454	-1.89	0.0744			
X**3	1	4.0671	1.4191	2.87	0.0099			
Salary	1	-6.761E-9	4.2613E-9	-1.59	0.1291			

Estimate of Lag Distribution								
Variable	Estimate	Standard Error	t Value	Approx Pr > t	-3.031 0 5.8021			
X(0)	-3.031318	2.2262	-1.36	0.1892	[*********			
X(1)	4.171272	2.6509	1.57	0.1321				
X(2)	5.624462	2.9892	1.88	0.0753	*************************************			
X(3)	4.359700	3.1382	1.39	0.1808				
X(4)	3.408434	2.9752	1.15	0.2662				
X(5)	5.802112	2.4950	2.33	0.0313	****************			

Summary: Trade Up...or Say Goodbye?

Lag models use regressors that include previous values of a predictor variables

Distributed lag models incorporate effects over <u>successive</u> time periods

In baseball, the most predictive months occur in different parts of the season

The Trade Deadline is best modeled by current place and the win rate in May and July

Questions

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