$$ln(Salary) = \beta_0 + \beta_1 * WAR + \beta_2 * NonWhite + \beta_3 * ServiceTime + \sum_{i} \beta_{i+3} * Trad_i + Team F.E.$$

older front-office employees still weigh traditional statistics heavier than they do sahermetrics statistics. In Salary) = $\beta_0 + \beta_1 *$ WAR + $\beta_2 *$ NonWhite + $\beta_3 *$ ServiceTime + $\sum_i \beta_{i+2} *$ Trad_i + Team F.E. White Statistics rather than traditional statistics in these passes for this the R^2 value increases to 0.6005 and the comparing tearninates these The sixth model explores the team fixed effects that are present in these regressions and then also controls for sabermetric statistics rather than traditional statistics. As you can see, when controlling for this the R² value increases to 0.050 and the effect of race increases to -15.8%. Implying that when comparing teammates there is an increased discrimination against players of color.

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The seventh model is similar to the sixth, however, it includes controlling for traditional stats as well. This only slightly increases the R^2 to 0.6103 adm increases the effect race has on salary to -16.9%

The severalt model is similar to the sixth, however, it includes controlling for traditional stats as well. This only slighly increases the R^2 to 0.6103 ada increases the effect race has on salary to -16.9% with the player is non-white.

In (Salary) = $\beta_0 + \beta_1 \cdot \text{WAR} + \beta_2 \cdot \text{Non-White} + \beta_4 \cdot \text{Non-White} + \beta_4 \cdot \text{Non-White}$

 $\ln(\mathrm{Salary}) = \beta_2 + \beta_1 * \mathrm{WAR} + \beta_2 * \mathrm{NonWhite} + \beta_3 * \mathrm{ServiceTime} + \sum \beta_{t+3} * \mathrm{Tind}_t + \sum \beta_{t+3+2} * \mathrm{Sab}\tau_t + \mathrm{Team} \; \mathrm{F.E.}$

	onWhite on !	

	(1) In.salary	(2) ln_salary	(3) In salary	(4) Insalary	(5) la_salary	(6) In salary	ln_salary	
WM	(5.19)	0.0688***	0.0360° (2.41)	(2.39)	0.0389 (1.92)	(2.85)	0.0358 (1.72)	
age	0.141*** (15.22)							
nonwhite	-0.00272 (-0.05)	-0.188*** (-3.81)	-0.176*** (-3.43)	-0.176*** (-3.43)	-0.150** (-2.77)	-0.158** (-2.95)	-0.169** (-3.11)	structured
year	0.0220 (1.84)	0.0322** (3.10)		0.00557 (0.45)				SUMME
servicetime		0.180*** (22.56)	(21.46)	0.175*** (21.38)	0.173*** (19.24)	(20.46)	0.174*** (19.28)	
Constant	-32.99 (-1.36)	-50.43* (-2.41)	10.36*** (6.21)	-0.836 (-0.03)	14.13*** (29.53)	9.500*** (5.32)	(5.42)	
Traditional Stats	No	No	No	No	Yes	No	Yes	
Sabermetric Stats	No	No	Yes	Yes	No	Yes	Yes	
Team Fixed Effects	No	No	No	No	Yes	Yes	Yes	
N R ²	587 0.2872	587 0.4685	587 0.5068	587 0.5069	535 0.5794	535 0.6005	535 0.6103	

From curiosity and completeness, I explored further and asked the same question but this time wondered if there was pay descrimination based on birth country. Substituting the international vari-able for nonwhite, I run the same regressions again which can be seen in Table 4. Similar results were