By the Numbers

The Newsletter of the Statistical Analysis Committee of the Society for American Baseball Research

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Greetings

Welcome to all old and new committee members. I hope that everyone is having a pleasant winter, and that you are able to keep up with players' comings and goings better than I. We in the San Francisco Bay Area are still smarting over the departure of Will Clark (though I don't want to sound like one of those people who goes around saying "in the good old days ...").

This issue of the newsletter contains four interesting articles. Tony Darkin kicks us off with a piece on one possible cause of high/low pitchers' run support.

Willie Runquist follows-up a previous article with a piece on the variability in pitching statistics.

Michael Wolverton introduces a new method to evaluate the true quality of a pitcher's start.

Don Coffin writes on whether good pitching stops good hitting in a short series.

Although we have not depleted our backlog of material, we always encourage submissions for the newsletter.

Please send material, comments, etc., to my address: Rob Wood, 2101 California St. #224, Mountain View, CA 94040. My home number is (415) 961-6574, and my daytime number is (415) 854-7101.

Notes from committee members

In this section of the newsletter, I will pass along news/ideas/information I receive from committee members.

David Stephan has made available two sets of materials which he presented at the national convention in San Diego. The first, a paper entitled "Allan Roth's True Discovery of Sabermetrics Revealed," details the many pioneering contributions of the late Allan Roth to the field which only later became known as sabermetrics. The second is a comprehensive paper entitled "Batting Barrages: Streaks, Tears, & Rampages." If you'd like a copy of this material, contact me or David at 1248 South La Jolla Avenue, Los Angeles, CA 90035-2645.

Art Schott (Louisiana's Official Baseball Historian) has forwarded the findings of some of his historical research to make them available to other interested members. First is a home run comparison of Hank Aaron and Babe Ruth. Second is research into home runs hit in the 1953 Northern League (Roger Maris' first season as a professional in which he hit 9 HR). Third is Art's newspaper column describing Lou Gehrig's last major league home run hit in an exhibition game in April 1939. Finally, Art sends a game by game description of Mickey Mantle's and the 1951 New York Yankees' exhibition season. If you'd like a copy of this material, contact me.

Ben McGrath is a new SABR member and seeks data on intentional walks. He is looking for lists of the hitters (preferably more than just the league leader) who received the most intentional walks for each season they have been recorded. If you have any information, contact me or Ben at 213 E. Allendale Ave., Allendale, NJ 07401.

James Graham is another new SABR member who is interested in any statistical studies which have tested the theory of Ted Williams suggesting that a batter should routinely take the first pitch of his first at bat in a game to refamiliarize himself with the pitcher and learn what the pitcher has that day. If you know of any studies, contact me or James at 4601 Katy Court, La Mesa, CA 91941.

Notes from other committees

Andy McCue, chairman of the Bibliography
Committee, writes to solicit our help in SABR's
"Baseball Online" project. The project seeks to create a
comprehensive electronic bibliography of Englishlanguage materials of interest to baseball researchers.
This research tool is of great potential use to us in the
Statistical Analysis Committee.

Andy is looking for volunteers to help in the project in several areas. One area would be organizing the project as it affects statistical research. Another area would be contacting committee members and finding information about materials that should be included in the bibliography. Another area, of course, would be cataloguing the material once it is found and assembled.

If you have any questions on the project, or want to express interest in it, please contact Andy at 4025 Beechwood Place, Riverside, CA 92506; phone: (909) 787-4954.

Eduardo Valero, chairman of SABR's new Latin America Committee, writes to say hello. If there is anything that his committee can do for us, let him know. If you're doing a study pertaining to Latin Americans (e.g., why isn't Orlando Cepeda in the Hall of Fame), Eduardo can help. Of course, the Statistical Analysis Committee is happy to reciprocate in the offer. Eduardo can be reached at 670 Ponce de Leon, Apt. 516, Santurce, Puerto Rico, 00907; phone: (809) 724-7762.

PITCHER RUN SUPPORT

by Tony Darkin

In 1992 Jack Morris won 21 games for the Blue Jays with an ERA of 4.04. Obviously not many pitchers win 20 games with an ERA as high as that. Possibly the main reason for Morris' high number of wins was the number of runs his team scored for him.

Whilst preparing for last year's fantasy baseball draft, Jack Morris' name came up. I mentioned to a friend that he would win 20 games if his team scored 5.98 runs per game for him. The matter of Run Support was discussed for a while, and my friend suggested that the reason for Jack getting so many runs was that he was a veteran pitcher well past his best and the opposition

would not feel that they would have to put their best starter out against Toronto to beat them.

To the best of my knowledge, major league teams do not think this way. Pitching rotations are pretty well set up for their starting pitchers to pitch every 5th day barring injuries, call-ups from the minors, etc.

However my friend's thought struck a chord and I thought it would be interesting to study the quality of opposition starters that Morris faced in 1992. To widen the study, I looked at an additional three pitchers who also had extreme run support that season.

First of all, let's try to define what makes a quality starting pitcher. In his 1986 Baseball Abstract (page 238), Bill James did a small study on the number of front-line starters faced by teams in the National League West.

James' definition of a quality pitcher in that study was any starter who pitches 162 innings and has (1) a winning record or (2) an ERA better than the league. Liberal though this definition is, it is the one that we shall use here. In 1992, there were 33 pitchers in the American League who met these criteria, just over 2 per team. The pitchers whose run support is studied here are all on the list.

What we are now looking for is whether facing quality opposition pitchers has any effect on a starter's Run Support. That is, was Jack Morris' 5.98 runs per game of offensive support due to the fact that he faced mainly weak pitching from the other team? Note that the average starter in the majors had 4.32 runs per game support in 1992.

Jack Morris was 21-6 with a 4.04 ERA and 5.98 runs support per nine innings pitched. Jack is the only pitcher that has been mentioned so far since the support he received was phenomenal. So let's have a look at the data.

Morris started 34 games in 1992, 13 against frontline starters and 21 against non-front liners. In the 21 games against poorer pitchers, Morris pitched 147.1 innings with a won-loss record of 8-6, an ERA of 4.64, and runs support of 4.58 runs per 9IP. In the 13 games against front-line starters, he pitched 93.1 innings with a won-loss record of 13-0, an ERA of 3.09, and runs support of 8.20 runs per 9IP!

So much for facing weak opposition. Morris appears to have turned that around straight away. Although Morris had massive run support against front-line

starters, more remarkable to me was his 3.09 ERA. You or I would almost certainly go 13-0 if our teammates scored 8.20 runs for us, but would we have that low ERA? Looking at this data, the only way to beat Morris in 1992 was to put your weakest starter out against him. So far, it does not appear that facing good pitching reduces your run support. However we have other pitchers to study.

Jim Abbott was 7-15 with a 2.77 ERA and only 2.65 runs support per 9IP in 1992. Abbott was chosen for the study because he was the exact opposite to Jack Morris, with his low ERA, poor won-loss record, and very low run support.

Abbott started 29 games, 17 against front-line starters and 12 against non-front liners. In the 17 starts against quality pitching, he pitched 114.1 innings with a won-loss record of 4-11, an ERA of 3.23, and run support of 2.20 runs per 9IP. In his 12 starts against "bad" pitching, Abbott pitched 96.2 innings with a won-loss record of 3-4, an ERA of 2.23, and run support of 3.17 runs per 9IP.

Abbott's run support was increased when he went up against inferior opposition. Almost a full run per game, although the jump from 2 runs per game to 3 is hardly earth shattering. The bottom line here is probably that the 1992 Angels did not score many runs against anybody!

So having looked at two pitchers we cannot yet see any real pattern emerging, for one pitcher (Morris) run support increased against front-line starting pitcher, while for the other (Abbott) it went down.

Let's have a look at two pitchers from the same team who had varying run support to see if there is any trend there. Melido Perez had a 13-16 record with a 2.87 ERA for the Yankees in 1992, but his run support was 4.07 -- not exceptionally low, but 1/4 of a run below the major league average. On the other hand, Scott Sanderson had a 12-11 record but with a 4.94 ERA and run support of 6.24 runs per game.

Perez obviously was a superior pitcher to Sanderson given that his ERA was more than 2 runs per game lower, but his run support was also more than 2 runs per game less. You don't need to be Einstein to figure out that the discrepancy in won-loss records was due to run support. But did Perez come up against tougher opposition than Sanderson?

Sanderson started 33 games in 1992, 14 against front-line starting pitching, 19 against non-front liners. In the 14 games against quality pitching, he pitched 80.1 innings with a won-loss record of 5-6, an ERA of 5.60, and run support of 5.04 runs per 9IP. In the 19 games against the retreads, he pitched 113 innings with a won-loss record of 7-5, an ERA of 4.46, while run support increased to 7.09 runs per 9IP.

This appears to be in line with Jim Abbott's performance: higher ERA and lower run support against front-line starters; lower ERA, better won-loss record, and higher run support against non-front line starters. Perhaps there is a trend here after all, but we need to look at Sanderson's hard luck teammate.

Melido Perez started 33 games in 1992, 17 against quality starters, 16 against non-quality starters. In the 17 starts against front-liners, he pitched 129 innings with a 5-11 won-loss record, an ERA of 3.28, and support of 4.40 runs per 9IP. In the 16 starts against non-front liners, he pitched 118.2 innings with a won-loss record of 8-5, an ERA of 2.43, and support of 3.72 runs per 9IP.

Perez seems to match up a little with everybody else in the study. His run support went up against front-line starters as did Jack Morris', while his ERA and won-loss record became worse as did Abbott's and Sanderson's. But there is nothing to suggest that his run support was low because he pitched substantially more innings against the best starters in the league.

Sanderson and Morris were the only pitchers who had any large differential in innings pitched against non-front line starters (Morris +54, Sanderson +32.2), but Morris' run support went down when pitching against guys in this category.

In conclusion, after hours of wading through box scores and scribbling notes on hundreds of sheets of paper, I can find no real evidence to support the hypothesis that pitcher run support is heavily decided by the quality of the opposition pitching. Indeed, in this study the average run support went up against front-line starters (4.77 vs. 4.66). Admittedly, the pitchers here have been extremes and perhaps a larger study would reveal evidence as to the extent of the run support. I would gladly welcome any comments, criticisms, or questions on this study.

Tony Darkin, 46 Adelaide Road, Edgeley, Stockport, England SK3 9LP

SEASONAL VARIATION IN PITCHING by Willie Runquist

In a previous article the average deviation of a seasonal statistic from the average of several seasons (V*) was used to measure a batter's variability from season to season. Here we present similar values for pitching statistics. The data sample consists of all 37 pitchers who pitched more than 100 innings in every season from 1987 to 1991. V* is the mean of the absolute deviations of the pitcher's statistic in each season from the overall mean of that pitcher's statistic. The results are presented for seven statistics: earned run average, won/lost percentage, number of wins, and hits, on-base, walks, and strikeouts per nine innings. The following table presents the mean and standard deviation among the V*'s of the 37 pitchers in the sample.

Pitching	Average	Std.Dev.
Category	<u>Y*</u>	<u>V*</u>
ERA	0.53	0.18
W-L Pct.	0.09	0.04
Wins	3.05	1.27
Hits/Game	0.64	0.35
On-Base/Game	0.75	0.37
Walks/Game	0.36	0.17
Strikeouts/Game	0.59	0.22

Values of V* do not necessarily indicate the quality of pitching, but only its consistency. In fact, there is some regression, particularly in strikeouts, in that pitchers who seldom strike out batters cannot vary much from season to season.

Although we are not directly concerned with predicting one season's performance from another, we will report some correlations between successive seasons. The correlations were quite variable, but are nevertheless positive and in some cases predictive accuracy was at least as good as that for batting statistics. These correlations are provided in the following table.

We carried out a similar analysis on a larger sample of 100 pitchers that had 100 innings pitched in both 1979

Year	Wins	ERA	Hits	OB	BB	SO
1987-88	.10	.14	.08	.24	.63	.61
1988-89	.39	.20	.28	.43	.71	.72
1989-90	.05	.49	.53	.53	.62	.85
1990-91	.50	.38	.56	.58	.76	.81

and 1980. For games won, the mean absolute change from 1979 to 1980 was 4.46 wins with a standard deviation of .73. The correlation between 1979 and 1980 wins was .28. For earned run average, the mean absolute change was .73 earned runs, with a standard deviation of .62. The correlation between 1979 and 1980 ERA was only .18.

As in the case of batters, the problem with prediction from one year to the next results from the relative homogeneity of the pitchers. About two-thirds of the pitchers have earned run averages within .7 of the overall average of all pitchers. The variation between seasons for the same pitcher is therefore likely to be as great as the difference between pitchers. Pitchers do seem to be more consistent with respect to walks and strikeouts, however.

Willie Runquist, The Union Bay Oyster Chucker, P.O. Box 289, Union Bay, B.C. VOR-3BO CANADA

"SUPPORT-NEUTRAL" STATISTICS: A METHOD OF EVALUATING THE TRUE QUALITY OF A PITCHER'S START

by Michael Wolverton

MOTIVATION

In recent years, we've seen the development and growing use of two measurements designed to evaluate starting pitchers on a game-by-game basis: Quality Starts and Game Score. Both measures are attempting in some way to look at the quality of each outing the starter has, rather than looking at the average or cumulative performance over the course of year like ERA does. But both measures have their weaknesses as total measures of pitching performance.

The arguments against Quality Starts are well known. Detractors point out that the worst qualifying outing—6 innings and 3 earned runs—is not "quality" at all. A related objection is that Quality Starts makes no attempt to quantify the degree of quality a start has—6 innings, 3 runs is the same as 8 innings, 2 runs which is the same as a 9-inning shutout.

Partly in answer to these objections, Bill James developed the Game Score, which combines a starter's box score numbers (IP, H, ER, R, BB, K) using weights, where the weights are assigned such that the league average score is around 50, the best imaginable score is around 100, and the worst imaginable score (by someone outside the state of Colorado) is around 0. Game Score is acknowledged as an interesting measure of "game domination" by a starter, but it has weaknesses as a total measure of starter quality (i.e., his contribution to team victories): it's too dependent on strikeouts, possibly too dependent on hits and walks (after all, the number of runs given up is really the only thing that matters), and it isn't park-adjusted.

Despite the weaknesses of these two measures, looking at a pitcher's starts game-by-game is still a good idea. Looking at each start's contribution to winning, rather than cumulative run-prevention over the course of a year (ERA or Pitching Runs), can help us answer questions like: Given equal ERAs, do some pitchers pitch in a way that will tend to win more games than other pitchers? In particular, is it better for a starter to be flaky—either very good or very bad on a given day—or consistently average? Does the park have a smaller influence on the value of the start when the start is very good or very bad?

So here's what we'd like out of a stat measuring the quality of a start:

- it should depend only on numbers appearing in a box score;
- it should be independent of a pitcher's support, both from his team's offense and from his team's relievers;
- it should be park-adjusted;
- the resulting measurement should be in terms of some kind of meaningful unit, such as games or runs, rather than being a unitless index (and, ideally, it should be obvious to any baseball fan what a good or bad score in those units is);

 most importantly, it should reflect the contribution that a start had toward winning the game.

I've developed a couple of measurements that meet these five requirements. (Actually, the ideal stat would also be very easy to compute, but hey, 5 out of 6 isn't bad, right?) Support-Neutral Wins and Support-Neutral Losses (SNW and SNL) measure the expected number of wins and losses a pitcher would have with his outings, if he got average support from his offense and his bullpen. Support-Neutral Value Added (SNVA) measures the total number of games that an average team would win given the pitcher's starts, over the number of games they'd win with a league average starter. All of these stats are computed using only the number of innings pitched, number of runs given up, and the park the game was pitched in. SNVA may be a slightly more accurate measure of a starter's actual value compared to league average, but the SNW/L record has the advantage of being flexible and more understandable. Both of them, in my opinion, constitute an improvement over Thorn and Palmer's Pitching Runs as a total measure of starter worth.

Support-Neutral Wins and Losses

Support-Neutral Wins is calculated by determining the probability that a pitcher would get the win for each start he has, and then summing up the individual probabilities over all of his starts. The sum gives you the number of wins a pitcher could expect to get for an average team, given his performances. A "performance" here consists only of the number of innings pitched, the number of runs (not earned runs) given up, the park in which the game was played, and whether the pitcher was at home or on the road—SNW assumes that these are the only things which influence whether the pitcher wins or loses.

The rest of this section describes the formulas that are used to calculate SNW; readers who aren't interested in the specific methods of calculation are welcome to skim or skip to the next section.

To calculate the probability that a pitcher wins the game, we just need to look at the definition of a win: A starting pitcher wins the game if his team has the lead when he's taken out of the game, and they never relinquish that lead. So, for a given outing by the starter, the probability that he gets the win is just the probability that his team will take the lead (score more runs than the

starter gives up) by the time he's removed times the probability that they'll hold that lead until the game is over.

To put this into a formula, we just need to determine and add up the probabilities of all the different ways his team can take and hold a lead:

$$SNW(i,r) = \sum_{j=r+1}^{\infty} PScore(i,j) \times PHold(j-r,9-i)$$

where SNW(i, r) is the probability a starter who goes i innings and gives up r runs will get the win, given an average team playing behind him.

PScore(i, r) is the probability that an average team will score r runs in i innings.

PHold(k, i) is the probability that an average team will hold a k-run lead (without ever relinquishing it) for the i remaining innings until the end of the game.

The above formula is actually a simplification of the formula I use in my software to calculate SNW (I'll refer to the formula in my software as the "real" SNW formula). In order to make it easier to explain, I made a few assumptions to get the formula above. First, that formula assumes that the starter comes out of the game after pitching a full inning (i.e., he pitches no extra thirds of an inning). The formula is complicated somewhat when thirds of an inning are taken into account, but the same general idea applies: his team must be leading when he comes out, and his team must hold the lead for the extra thirds in the inning he leaves, plus all the rest of the remaining innings. The real SNW formula does take thirds of an inning into account.

Second, the above formula doesn't explicitly take the park into account. To take park effects into account, we need to make SNW, PScore, and PHold be functions of the park in which the game is played. A hitter's park should inflate the probabilities that an average team will score a high number of runs, and a pitcher's park should do the opposite. The real SNW formula does take park effects into account. I talk a little more about my handling of park effects in the Appendix.

Third, the above formula doesn't take into account whether the starter is pitching at home or on the road. Maybe contrary to intuition, this does make a difference. Consider a starter who leaves after pitching the 7th

inning: if he's at home, he's pitched the top of the 7th, so he gets credit for the runs his team scored in the first 6 innings, plus the runs they score in the bottom of the 7th; if he's on the road, however, he pitched the bottom of the 7th, so he gets credit for the runs his team scored in the first 7 innings, plus the runs they score in the top of the 8th. So, all other things being equal, it's easier for pitchers to get wins (and harder for them to get losses) when they pitch on the road. The formula above is for a pitcher pitching at home, and the road formula is slightly different. The real SNW formula does take home/road status into account.

Finally, the above formula doesn't quite reflect the full definition of a pitcher's win—a starter can't get the win unless he goes 5 innings or more. Presumably, this extra condition was put into the win rule to reduce the number of undeserving starters getting lucky wins. But when you're assigning fractions of a win, rather than 1 win or 0 wins, there's no possibility of getting lucky. So, the real SNWL formula does *not* take the five-inning condition into account, although, for the purposes of comparison, I do calculate an expected win (E(W)) number which is equal to 0 if the pitcher goes less than 5 innings and equal to SNW otherwise.

Let's finish off the formula above. PScore is easy to find recursively, provided you know an average team's single-inning scoring distribution, PInningScore:

$$PScore(i,r) = \sum_{j=0}^{r} PInningScore(j) \times PScore(i-1,r-j), \text{ for } i > 1$$

PScore(1,r) = PInningScore(r)

where PInningScore(r) is the probability that an average team will score r runs in an inning.

PHold is a little more complicated, since you have to see to it that the pitcher's team never relinquishes the lead. Still, it's not too hard to reduce it to the following (below, "tr" stands for the number of runs the pitcher's team scores in an inning, and "or" stands for the number the opposing team scores in an inning):

$$PHold(k,i) = \sum_{r=0}^{\infty} \sum_{or=0}^{k+r-1} PInningScore(tr) \times PInningScore(or)$$

$$\times PHold(k+tr-or,i-1), \text{ for } i>0$$

PHold(k,0) = 1

The only remaining unknown is the single-inning scoring distribution, PInningScore. But that's readily available from linescores of past games. The scoring distribution (separate distributions for each league) I'm using right now was taken from a few weeks of linescores in USA TODAY from late-April and early-May of 1992. I'll probably be able to get a more accurate distribution someday, but I'm sure that this one is close enough. The SNL value for a single start is calculated analogously to SNW.

Support-Neutral Value Added

SNW and SNL gives us a nice way of getting a "fair" W/L record for a starter, which can then be used to compare to his actual W/L record, or a replacement-level winning percentage, etc. (see the Results section). But these numbers calculate how likely it is that the pitcher will win or lose the game—i.e., get the "W" or "L" next to his name in the box score. A related but slightly different notion is the likelihood that the team will win when a pitcher takes the mound. In measuring the starter's contribution to team victories, we'd like to evaluate how much the outing by the starter changes the team's chance of winning from what it was at the beginning of the game (which I'll assume to be 50%). This is what SNVA is designed to measure.

Not surprisingly, the formula for SNVA looks pretty similar to the formula for SNW:

$$SNVA(i,r) = -0.5 + \sum_{i=0}^{\infty} PScore(i,j) \times PATWin(j-r,9-i)$$

where SNVA(i, r) is the difference between an average team's chance of winning after the starter has left after pitching i innings and giving up r runs, and their chance of winning at the beginning of the game (50%).

PScore(i, r) was defined above.

PATWin(r, i) is the chance that an average team will eventually win the game given that there are i innings left and the difference between their score and their opponents' score is r.

Also not surprisingly, PATWin looks a lot like PHold:

$$PATWin(r,i) = \sum_{t=0}^{\infty} \sum_{\sigma=0}^{\infty} PInningScore(tr) \times PInningScore(\sigma r)$$

$$\times PATWin(r+tr-or,i-1)$$
, for $i>1$

PATWin(r,0) = 1, for r > 0

PATWin(0,0) = 0.5

PATWin(r,0) = 0, for r < 0

What SNVA gives us (when summed over all a pitcher's starts) is the number of games in the standings he's worth to his team above the average starter. Of course, this is exactly the same unit (games above the average player) that all of <u>Total Baseball</u>'s [1] measurements are in. So it'll be interesting to compare SNVA to Thorn and Palmer's Adjusted Pitching Runs to see how well they correlate and also where the differences lie.

RESULTS

Best, worst, luckiest, and unluckiest starters of 1992

That's enough of the gory details of the calculation of the stats. Let's look at the fun stuff—what the stats tell us about real pitchers. I tracked all starting pitchers in the majors over the 1992 season, and Tables 1 and 2 show the top pitchers in both leagues for 1992. Each table shows the pitcher's Support-Neutral Wins (SNW), Losses (SNL), and Winning Percentage (SNPct), followed by his actual win-loss record (W, L), his runs allowed per 9 innings (RA), his Adjusted Pitching Runs¹ (APR), and his Support-Neutral Value Added (SNVA). Interestingly, Greg Maddux, with the fabulous year he had pitching in Wrigley, was the only pitcher in either league who came close to "deserving" to win 20 games.

¹ Adjusted Pitching Runs is the basic metric which Thorn and Palmer (the authors of <u>Total Baseball</u>) use to evaluate pitchers. APR is the number of runs prevented by a pitcher that a league average pitcher would've given up. The APR that I'm using in this paper differs from Thorn and Palmer's statistic in two ways: 1) I'm using runs where Thorn and Palmer use earned runs, and 2) the method of park adjustment I use is a simplification of the one used in <u>Total Baseball</u>. It is included here for comparison with SNVA.

Pitcher	Team	SNW	SNL	SNPct	W	L	RA	APR	SNVA
Mussina	BAL	17.2	7.8	.688	18	5	2.61	47.0	4.60
Clemens	BOS	17.5	8.5	.674	18	11	2.92	43.8	4.39
Appier	KCR	15,2	6.6	.698	15	8	2.55	42.6	4.08
Guzman,Ju	TOR	13.4	6.4	.679	16	5	2.79	32.3	3.34
Nagy	CLE	16.3	9.9	.623	17	10	3.25	33.3	3.11
Eldred	MIL	8.2	2.4	.776	11	2	1.88	28.1	2.81
McDowell	СНІ	16.3	10.7	.602	20	10	3.28	30.5	2.53
Smiley	MIN	16.0	10.5	.603	16	9	3,47	28.3	2.75
Navarro	MIL	15.8	10.8	.595	17	- 11	3.59	22.8	2.45
AbbottJ	CAL	13.6	8,6	.612	7	15	3.11	27.7	2.36
Viola	BOS	15.8	11.2	.586	13	12	3.74	21.4	2.35
Fleming	SEA	15.1	10.7	.586	17	10	3.73	19.7	2.10
Perez,M	NYY	14.9	10.5	.586	13	16	3.42	26.3	1.90
Wegman	MIL	15.6	11.5	.576	13	14	3.58	24.4	2.06
Erickson	MIN	13.3	9.8	.574	13	12	3.65	20.8	1.75
Bosio	MIL	14.3	11.1	.563	16	6	3.89	13.6	1.52
Key	TOR	13.3	10,4	.561	13	13	3.66	18.0	1.42
Brown,K	TEX	15.5	12.9	.545	21	11	3.96	14.5	1.18
Welch	OAK	8.0	5.8	.580	- 11	7	3.42	9.7	0.91
Rasmussen	KCR	3.0	0.8	.785			1.67	11.4	1.09

Table 1: Top 20 AL Starters in 1992, ranked by SNW-SNL

Pitcher	Team	SNW	SNL	SNPct	w	L	RA	APR	SNVA
Maddux,G	CHI	19.5	7.4	.724	20	11	2.28	53.9	5.75
Tewksbury	STL	16.1	7.3	.687	15	5	2.45	38.5	4.12
Schilling	PHI	13.9	6.8	.670	12	9	2.59	31.1	3.37
Morgan	CHI	16.3	9.5	.632	16	8	3.00	30.4	3.22
Rijo	CIN	13.9	8.1	.632	15	10	2.86	28.5	2.57
Smoltz	ATL	16.6	11.0	.601	15	12	3.28	25.1	2.67
Glavine	ATL	15.1	9.8	.608	20	8	3.24	23.9	2.71
Martinez,D	MON	14.5	9,1	.613	16	- 11	2.98	24.0	2,50
Swindell	CIN	13.8	8.5	.619	12	7	3.05	24.5	2.56
Swift	SFG	10.4	5.1	.670	9	3	2.36	23.6	2.51
Drabek	PIT	15.9	10.8	.595	15	11	2.95	26.7	2.32
Fernandez,S	NYM	13.6	8.8	.608	14	11	2.81	24.7	2.25
Hill	MON	14.0	10.0	.583	16	9	3.14	19.3	1.93
Leibrandt	ATL	13.5	9.5	.586	15	7	3.68	11.8	2.02
Smith,P	ATL	5.6	2,1	.724	7	0	2.22	16.2	1.69
Wakefield	PIT	6.4	3.3	.656	8	1	2.54	13.8	1.42
Rivera	PHI	6.5	3.7	.639	7	3	2.95	10.8	1,34
Benes	SDP	14.0	11.3	.553	13	14	3.50	10.7	1,22
Portugal	HOU	6.6	4.0	.621	5	3	2.69	12.5	1.18
Avery	ATL	14.0	11.5	549	11	11	3.66	14.8	1.15

Table 2: Top 20 NL Starters in 1992, ranked by SNW-SNL

On the flip-side, Tables 3 and 4 show the worst 10 starting pitchers in 1992 for each league. Not surprisingly, many of these guys showed up in different uniforms in 1993, several on expansion teams. Actually, it's probably inaccurate to use the word "worst" here, since the method of ranking the pitchers — ranking them according to SNW-SNL — sets the baseline for comparison at league average (anyone below .500 gets a negative rating). Of course, it's quite possible for a

below-average pitcher to still be valuable to his team. A better method of producing this list might have been to compare a pitcher's SN record to a lower baseline, e.g., a .450 pitcher. This would have left pitchers like Hershiser and McCaskill, who pitched a lot of innings at somewhat below-league-average performance, off of the lists in favor of other pitchers who pitched fewer innings but at further-below-average performance.

Pitcher	Team	SNW	SNL	SNPct	w	L	RA	APR	SNVA
Armstrong	CLE	5.2	11.5	.313	3	15	6.37	-28.4	-3.08
Milacki	BAL	4.5	9.5	.320	6	8	6.18	-21.8	-2.32
Terrell	DET	2.9	7.5	.280	3	6	6.98	-22.9	-2,26
Slusarski	OAK	2.5	6.9	.265	5	5	6.25	-18.7	-2.05
Sanderson	NYY	9.9	14.0	,414	12	11	5.40	-22.4	-2.02
Aldred	DET	2.4	6.5	.273	2	7	7.63	-21.7	-1.89
McCaskill	CHI	10.2	14.2	417	12	- 13	5.00	-16.1	-1.92
Wells	TOR	3,4	6.9	.332	- 6	7	7.70	-27,7	-1.81
Stieb	TOR	3.4	6.7	.337	3	- 6	5.92	-13.3	-1.50
Otto	CLE	3.9	7.2	.354	5	9	6.75	-19.8	-1.57

Table 3: Bottom 10 AL Starters in 1992, ranked by SNW-SNL

Pitcher	Team	SNW	SNL	SNPct	W	L	RA	APR	SNVA
Bowen	HOU	0.6	6.1	.094	0	7	12.22	-31.3	-2.61
Wilson,T	SFG	7.0	11.3	.384	8	14	4.79	-18.5	-2.03
Abbott,K	PHI	4.5	8.3	.352	1	14	4.92	-11.4	-1.84
Martinez.R	LAD	7.3	11.1	.397	8	11	4.90	-19.1	-1.84
Henry,B	HOU	8.3	11.7	.414	6	9	4.40	-12.4	-1.57
Young,A	NYM	2.8	6,2	.313	- 1	7	5.79	-16.8	-1,63
Black	SFG	8.6	11.9	.420	10	12	4.47	-14.7	-1,54
Hershiser	LAD	10.2	13.3	.434	10	15	4.31	-12.5	-1.60
Hammond	CIN	7.0	10.0	409	7	10	4.61	-7.4	-1.36
Blair	HOU	1.4	4.5	.241	1	5	7.51	-16.8	-1.52

Table 4: Bottom 10 NL Starters in 1992, ranked by SNW-SNL

This method also allows you to evaluate the level of luck a pitcher experienced in his W/L record—i.e. it allows you to look at how much a pitcher's actual W/L record differs from his expected W/L record given the way he pitched. Tables 5 through 8 show the luckiest and unluckiest starters in each league in 1992. No one should be surprised that Jack Morris, who compiled a 21-6 record despite a 4+ ERA, was far and away the luckiest

starter in either league in 1992. SNW/L evaluation shows that you'd expect his 1992 performance to produce a 13-13 mark if he had gotten average support. Equally unsurprising is the result that Jim Abbott was the unluckiest pitcher in either league. The Angels gave him enough support only for a miserable 7-15 record, while his pitching actually merited something closer to 13-9.

Pitcher	Team	E(W)	E(L)	W	L	Diff
Morris	TOR	13.3	13.1	21	6	14.7
Brown,K	TEX	15.5	12.9	21	11	7.4
Moore	OAK	12.1	14.3	17	12	7.3
Bosio	MIL	14.2	11.1	16	6	6.9
Hibbard	CHI	8.1	11.3	10	7	6.2
Darling	OAK	11.5	12.5	15	- 10	6.0
Sanderson	NYY	9.7	14.0	12	- 11	5.3
Wickman	NYY	2.7	2.9	- 6	- 1	5.2
Slusarski	OAK	2.3	6.9	- 5	- 5	4.6
McDowell	CHI	16.2	10.7	20	10	4.5

Table 5: Luckiest 10 AL Starters in 1992, ranked by W-E(W) + E(L)-L

Pitcher	Team	E(W)	E(L)	W	L	Diff
Abbott_J	CAL	13.3	8.6	7	15	-12.7
Perez,M	NYY	14.6	10.5	13	16	-7.0
Hanson	SEA	9.6	12.8	7	17	-6.8
Armstrong	CLE	5.2	11.5	3	. 15	-5.8
Wegman	MIL	15.6	11.5	13	14	-5.1
Valera	CAL	10.4	9.6	7	11	-4.8
Kamienieck	NYY	8.8	12.0	- 6	14	-4.7
Ryan	TEX	9,3	8.7	5	9	-4.6
Chiampann	· TEX	1.5	1.3	0	4	-4.2
Reed	KCR	5.1	6.0	2	7	4.1

Table 6: Unluckiest 10 AL Starters in 1992, ranked by W-E(W) + E(L)-L

Pitcher	Team	E(W)	E(L)	W	L	Diff
Burkett	SFG	9.6	13.0	13	9	7.5
Glavine	ATL	15.0	9.8	20	8	6.8
Seminara	SDP	5.1	6.3	9	4	6.2
Lefferts	SDP	8.4	10.4	13	9	6.0
Tomlin	PIT	11.5	11.3	14	9	4.8
Hurst,B	SDP	12,4	12,1	14	9	4.7
Cone	NYM	11.3	9,9	13	7	4.6
Leibrandt	ATL	13.2	9.5	15	7	4.3
Osborne	STL	9.2	11.4	10	8	4.2
Wakefield	PIT	6.3	3.3	8	1	4.0

Table 7: Luckiest 10 NL Starters in 1992, ranked by W-E(W) + E(L)-L

Pitcher	Team	E(W)	E(L)	W	L	Diff.
Abbott,K	PHI	4,5	8.3	1	14	-9.2
Candiotti	LAD	11.8	10.5	10	15	-6.3
Gross,Ke	LAD	10.9	10.9	8	13	-5.0
Clark,M	STL	5.4	8.0	3	10	-4.4
Schilling	PHI	13.9	6.8	12	9	-4.0
Benes	SDP	13.8	11.3	13	14	-3.5
Carter	SFG	1.5	2.4	- 1	5	-3.1
Boskie	CHI	3.2	7.1	3	10	-3.1
Maddux,(3	СНІ	19.5	7.4	20	11	-3.1
Whitelaurst	NYM	2.3	3.3	- 1	5	-3.0

Table 8: Unluckiest 10 NL Starters in 1992, ranked by W-E(W) + E(L)-L

League total numbers

In theory, the support-neutral record of the entire league should come close to the actual win-loss record of the league, and in fact, in 1992, SNW/L did appear to predict league W/L pretty well. Table 9 shows both the expected and actual W/L totals for each league in 1992.

	E(W)	E(L)	E(Pct.)	w	L	Pct.
NL	660.9	690.3	.489	655	678	.491
AT.	776 1	8467	478	205	927	400

Table 9: Expected and Actual records of all starters in the major leagues

The National League's record corresponded very

well to the record expected by the model, with nodecisions being underpredicted only slightly by SNW/L. The American League is predicted a little less successfully—there were nearly 30 more wins in the league than expected, and nearly 10 more losses than expected.

I believe that part of the discrepancy between expected record and actual record can be explained by the fact that relief pitchers prevented runs better than starters in 1992. Since starters are competing for the (actual) decision primarily with the other starter, it makes sense that starters would get a few more (actual) wins than predicted by a model which has them competing with league average pitching for the decision.

Value of "flaky" and "steady" pitchers

Do the Support-Neutral stats tell us anything that Thorn and Palmer's Adjusted Pitching Runs weren't already telling us? Since both APR and SNVA are trying to measure exactly the same thing (albeit by different methods), we'd expect there to be a pretty strong correlation between them. There is. For most pitchers, SNVA (whose unit is "games above average") is approximately equal to one-tenth of APR (whose unit is "runs above average"). This is what you'd expect given the well-known result that each 10 runs prevented (or gained) leads on average to about 1 extra win in the standings (see, e.g., [2]). However, there are plenty of cases where APR and SNVA give significantly different evaluations. Look at the 1992 records of Charlie Leibrandt and Melido Perez:

	<u>APR</u>	<u>SNVA</u>	
Leibrandt	11.8	2.02	
Perez, M.	26.3	1.90	

APR evaluates Perez as being 14.5 runs—about oneand-a-half games—better than Leibrandt. However, SNVA shows that, when the pitchers' performance is evaluated game-by-game, Leibrandt was actually a little better than Perez.

The key to this discrepancy between the two measurements is found in the amount of consistency the two pitchers exhibited in their starts. Perez was a model of consistency in 1992; he rarely got bombed, but he also was rarely dominating. Leibrandt, on the other hand, was one of the *least* consistent pitchers in the majors.

And that is the most surprising result I've seen so far from these SN stats: run-prevention stats such as ERA and APR tend to undervalue flaky pitchers, and overvalue consistent ones, at least when you consider them pitching for an average team. Tables 10 through 13 show the "flakiest" (most inconsistent) and "steadiest" (most consistent) pitchers in the leagues in 1992, as evaluated by the variance of the SNVA of their individual starts.

You can see from those tables that APR pretty consistently underestimates a pitcher's value when the pitcher is flaky, and pretty consistently overestimates his value when he's steady. 9 of the 10 flakiest pitchers in both the NL and AL were underestimated by APR, and 8 of the 10 steadiest in the NL and 10 of the 10 steadiest in the AL were overestimated by APR. And the pitchers for whom there were really large discrepancies between APR and SNVA—Leibrandt, Kyle Abbott, Gooden, Hammond, Sutcliffe, Perez, Kamieniecki, McDowell—all showed up near the top of the predicted list.

The reason for this undervaluing is that APR counts all runs as equal, while in fact all runs do not contribute an equal amount toward winning/losing a game. In particular, Bill James did a study that showed that runs scored by a team after they've already scored 5 in a game do not contribute the same amount toward the probability of winning than those first 5 runs did [3]. So, pitchers who give up more than 5 runs in a couple of games will be undervalued by ERA and APR, because those really crummy outings probably weren't quite as crummy as ERA and APR would have you believe.

Pitcher	Team	APR	SNVA	SNVA
				Var
Smith,Z	PIT	3.7	0.70	0.088
Smoltz	ATL	25.1	2.67	0.083
Saberhagen	NYM	4.5	0.73	0.082
Leibrandt	ATL	11.8	2.02	0.082
Osborne	STL	-12.6	-0.97	0.079
Glavine	ATL	23.9	2.71	0.076
Hurst,B	SDP	-1.5	0.12	0.075
Cone	NYM	8.5	0.67	0.074
Belcher	CIN	1,9	0.53	0.074
Benes	SDP	10.7	1.22	0.068

Table 10: Flakiest 10 NL Starters in 1992, ranked by variance of SNVA (min 15 starts)

Pitcher	Team	APR	SNVA	SNVA	
				Var	
Sutcliffe	BAL	-8.3	-0.33	0.089	
Smiley	MIN	28.3	2.75	0.078	
Krueger	MIN	-0.1	0.14	0.078	
Johnson,R	SEA	1.7	0.24	0.077	
Gubicza	KCR	7.3	0.78	0.075	
Langston	CAL	5.6	0.77	0.073	
Fleming	SEA	19.7	2.10	0.073	
Viola	BOS	21.4	2.35	0.073	
Rhodes	BAL	6.7	0.77	0.071	
Darling	OAK	-4.8	40.31	0.070	

Table 12: Flakiest 10 AL Starters in 1992, ranked by variance of SNVA (min 15 starts)

As an example of this, consider a David Wells outing from 1992: he gave up 13 runs in 4+ innings. APR just subtracts his 13 runs from the number of runs a league average pitcher would have given up in those same 4 innings (about 2), and concludes that Wells was worth about -11 runs, or -1.1 games, in that start. Did Wells really cost the Blue Jays more than a game in the standings with that awful start? Of course not.

Pitcher	Team	APR	SNVA	SNVA
				Var
Abbott,K	PHI	-11.4	-1.84	0.022
Rijo	CIN	28.5	2.57	0.032
Browning	CIN	-8.7	-1.17	0.035
Gooden	NYM	-6.1	-1.33	0.036
Hammond	CIN	-7.4	-1.36	0.041
Tewksbury	STL	38.5	4.12	0.042
Maddux,G	CHI	53.9	5.75	0.042
Fernandez,S	NYM	24,7	2,25	0.043
Boskie	СНІ	-12.5	+1.47	0.044
Gardner	MON	-9.5	-1.20	0.044

Table 11: Steadiest 10 NL Starters in 1992, ranked by variance of SNVA (min 15 starts)

Pitcher	Team	APR	ŞNVA	ŞNVA
				Var
Armstrong	CLE	-28.4	-3.08	0.034
Darwin	BOS	8.2	0.45	0.036
Milacki	BAL	-21.8	-2.32	0.037
Kamieniecki	NYY	-8.9	-1.57	0.038
Perez,M	NYY	26.3	1.90	0.039
Reed	KCR	-0.3	-0.20	0.040
Appier	KCR	42.6	4.08	0.040
Cook	CLE	-3.6	-0.56	0.042
Hibbard	CHI	-10.7	-1.28	0.045
McDowell	CHI	30.5	2.53	0.045

Table 13: Steadiest 10 AL Starters in 1992, ranked by variance of SNVA (min 15 starts)

He guaranteed them a loss, of course, but they had some chance of losing the game to begin with anyway—about a 50% chance if you make the simplifying assumption that they're an average team. SNVA gives a far more reasonable value for Wells's start: it was worth about -0.5 games. That's as much as a single start can cost you. Wells didn't have the requisite 15 starts to show up in Table 12, but you can see from his record in Table 3 how much he was underestimated by APR.

Effect of the park on win probability

One other question I've been looking at is how the value of starts is influenced by park effects. Figure 1 shows the SNVA for a 9-inning complete game in both Wrigley Field (an extreme hitters' park) and the Astrodome (an extreme pitchers' park). We can see from the figure that the effect of the park on the value of

the start is far less at the two extremes of the start than it is for middle-of-the-road starts. The difference between Wrigley and the Astrodome for the value of a 9-inning, 5-run start is about four times as large as the difference between Wrigley and the Astrodome for the value of a shutout.

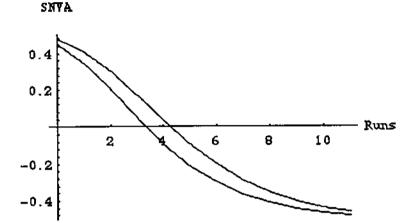


Figure 1: SNVA for Wrigley Field (top line) and the Astrodome (bottom line), given that the starter pitched 9 innings

This would imply that methods of park adjustments which simply multiply a pitcher's "raw" value by a park factor might be over- or underestimating the park's actual effect on his value. Since the park's effect on very good or very bad starts is much less than on average starts, a reasonable hypothesis would be that very good or very bad pitchers deserve less of a boost (or less diminishment) to their rating than current park adjustment methods give them.

However, the preliminary investigation of this hypothesis I have done on real starting pitchers (with 1992 data) has failed to find much support for it. I'd still like to do some more work on this issue.

Weaknesses of the Approach

Here are a few of the problems with the measurements which I have developed:

- They assume that scoring distributions of an inning are independent from the distributions of surrounding innings.
- They (like most other measures of pitching) don't account for situational pitching. A

pitcher who gets a big lead is likely to start throwing all fastballs, and he may give up a few meaningless runs that he wouldn't have given up without the big lead. I'm not too worried about this, because I don't think those big-lead situations are common enough for anybody to make much of a difference.

- They don't account for differences in the ways pitchers are used by their managers. Some pitchers get left in the game to get pounded, some are routinely yanked early, etc. Note however that SN stats do a better job than other methods of mitigating the manager's effect. If Cito Gaston leaves David Wells in the game to give up 13 runs, SNVA produces a rating which is not much different than if Gaston had yanked Wells after giving up "only" 7 or 8 runs.
- They don't account for the defense playing behind the pitcher. Suffice it to say that this is a very hard problem.

Conclusion

I've presented Support-Neutral Wins, Losses, and Value Added, three park- and league-adjusted measurements of the value of individual starts, and of starting pitchers. I feel these are a valuable addition to existing measurement methods, both because they can provide a measurement of pitcher worth in units which are familiar to all baseball fans (pitcher wins and losses) and because they seem to be a slightly more accurate measure of the true value of a start than existing methods.

Special thanks to Greg Spira, whose discussion sparked many of the ideas presented here. Thanks to David Tate and others on the Internet newsgroup rec.sport.baseball, who provided valuable feedback on the method. And thanks to my wife, Cindy, for reading this paper and giving me many useful suggestions.

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Appendix: Park Effects

One possible way of incorporating park effects into these measurements would be to take whatever final value the above formulas produce (SNW, SNL, or SNVA) and multiply it by some park effect constant for the pitcher's home park. This is essentially the approach Thorn and Palmer use in Total Baseball. But the method of calculating the Support-Neutral stats allows a potentially more informative use of park effects. Since park effects (e.g., in Elias) reflect how a park inflates or deflates average scoring ability, it makes sense to have the "average team" playing behind the pitcher effected by the park, and then calculate the likelihood that the pitcher's outing plus this park-adjusted average team will lead to a win. So for any game, the PInningScore (league average scoring) distribution is adjusted to reflect the park's effect on run scoring. The resulting number then reflects the park's effect on winning rather than cumulative run scoring/prevention.

The question then becomes how do you translate a single park effect percentage like the ones in Elias into an adjusted PInningScore distribution? There are an infinite number of ways to do this. The way I'm doing it now is to change the probability of scoring 0 runs by one factor, and change the probability of scoring i runs for i>1 by another factor, such that the total number of expected runs scored in an inning is increased/reduced by the Elias number. For example, if the Astrodome decreases scoring by 10%, I increase PlnningScore(0) for the Astrodome by one factor, and decrease PInningScore(i) for i>1 by another factor, such that the expected single-inning score reflected by PInningScore is reduced by 10% from the park-neutral scoring distribution. If that isn't clear (and I'm sure it isn't), I should say that I don't think it makes much difference the exact method used.

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DOES "GOOD PITCHING"? STOP "GOOD HITTING"? by Don Coffin

Baseball broadcasters, especially during the League Championship Series and the World Series, frequently tell us that "Good pitching stops good hitting, especially in a short series." But does it, really? I decided to find out, by examining the results of League Championship Series and World Series play.

I. "Dominant" Pitching and "Dominant" Hitting Teams

I looked at the results of the LCS and the World Series since 1969 (excluding 1981), identifying teams as "dominant hitting" teams and as "dominant pitching" teams. My criteria were simple—a team was a "dominant hitting" team if it scored more runs per game than the average in its league and had a higher ERA than its league average and scored fewer runs per game than its league average and scored fewer runs per game than its league average.

I identified 12 "dominant pitching" teams that made it into the LCS (four of which made it into the World Series), as shown in Table 1. All 12 of these teams played teams with above average R/G and ERAs below their league averages.

Of the 16 series played by "dominant pitching" teams, these teams won 7 and lost 9; three of them won the World Series (1969 Mets, 1985 Royals, and 1988 Dodgers); the 1973 Mets also made the World Series.

I identified only five "dominant hitting" teams (see Table 2); once again, all of them played teams with above average offensive and defensive performances. These teams won 3 and lost 4 series, with the Reds (in 1976) winning the World Series. The 1975 Red Sox were the only other "dominant hitting" team to reach the Series.

Both types of teams had comparable records. What I found interesting is that there were twice as many "dominant pitching" teams as "dominant hitting" teams.

II. "Stronger" Pitching and "Stronger" Hitting Teams

In most cases, division champions had above average runs per game and ERAs below their league average. Of 92 division champions, 74 were above average in scoring and in preventing runs, 12 were "dominant pitching" teams (as defined above), five were "dominant hitting" teams (also as defined above), and one was below average on both counts (the 1987 Twins). So, I looked at the playoff performance of the 74 teams who scored a lot and gave up only a little. I defined these teams as "stronger pitching" if their ERAs were further below the league average than their runs per game were above the league average (e.g., ERA 15% below the league average, runs per game 7% above the league average). "Stronger hitting" teams had runs per game further above the league average than their ERAs were below the league average.

The "stronger hitting" teams are shown in Table 3. Of these 35 teams, 17 made the World Series (49%), and 7 (20% of the teams and 41% of the World Series participants) won the World Series. Their overall record was 24 playoff series won and 28 lost.

Of the 39 "stronger pitching" teams (see Table 4), 22 (56%) made the World Series, winning it 11 times (28% of the teams, and 50% of the World Series participants). Their overall (series) record was 33-28.

III. Statistical Testing

I have, then, four groups of teams, with records ranging from 33-28 (0.541) to 7-9 (0.438). I also know how they fared in the World Series. I conducted a Chisquared test on the series records, against the null hypothesis that the expected winning percentage for all four groups of teams is 0.500, both overall and in the World Series. None of the records display a statistically significant difference from a 0.500 record. Based on this evidence, we cannot conclude that a team with a particular set of offensive and defensive performance levels will do exceptionally well -- or exceptionally poorly -- in championship play.

IV. Conclusions

My examination of the LCS and World Series results since 1969 suggests that there is no tendency for "good pitching" to stop "good hitting" with any regularity, with one proviso. During this period, there were a total of 51 "dominant" or "stronger" pitching teams, and only 40 "dominant" or "stronger" hitting teams. If we test this outcome against the null hypothesis that "dominant" and "stronger" pitching teams are as likely (but no more likely) to make the championships than "dominant" or "stronger" hitting teams, we calculate a t-statistic [for the difference between the observed frequency (51) and the expected frequency (45.5)] of 1.54. This is significant at about the 12% level (there is about a 12% chance that the observed frequency occurs by chance).

This suggests that "dominant" and "stronger" pitching teams have had a greater chance of winning their divisions during the past 25 years, but that, once they won their divisions, they were no more likely to do well in the LCS and the World Series than "dominant" or "stronger" hitting teams. In short, good pitching beats bad teams more often than good hitting does. However, I find nothing to suggest that we should expect teams with stronger pitching to do exceptionally well against other championship-caliber teams.

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Table 1: "Dominant" Pitching Teams				
Team/Year	Relative R/G	Relative ERA	Series W-L	
Mets/1969	96.2%	120.1%	2-0	
Pirates/1970	99.6%	109.5%	0-1	
Mets/1973	91.0%	112.3%	1-1	
Orioles/1974	99.2%	110.7%	0-1	
Astros/1980	96.9%	116.1%	0-1	
Dodgers/1983	97.8%	117.1%	0-1	
Royals/1984	94.0%	101.8%	0-1	
Dodgers/1985	92.8%	121.3%	0-1	
Royals/1985	93.1%	118.9%	2-0	
Astros/1986	96.6%	118.1%	0-1	
Dodgers/1988	99.9%	116.6%	2-0	
Blue Jays/1991	94.1%	116.9%	0-1	

Relative R/G is the team's R/G divided by its league average. Relative ERA is the league average ERA divided by the team's ERA (so that teams with better-than-average ERAs have relative ERAs in excess of 100%). These definitions also apply to Tables 2, 3, and 4.

Table 2: "Dominant" Hitting Teams			
Team/Year	Relative R/G	Relative ERA	Series W-L
Red Sox/1975	115.7%	95.0%	1-1
Reds/1976	132.9%	99.7%	2-0
Angels/1979	114.6%	97.2%	0-1
Braves/1982	111.6%	94.2%	0-1
Cubs/1984	116.4%	95.7%	0-1

Team/Year	Relative R/G	Relative ERA	Series W-L
Braves/1969	105.2%	101.7%	0-1
Twins/1969	119.2%	111.7%	· · 0-1
Pirates/1971	124.4%	104.8%	2-0
Giants/1971	111.5%	104.5%	0-1
Orioles/1971	121.4%	115.7%	1-1
Reds/1972	117.5%	107.5%	1-1
Reds/1973	110.2%	107.6% .	0-1
Pirates/1974	111.7%	103.7%	0-1
Reds/1975	125.7%	107.4%	2-0
Phillies/1976	119.4%	113.6%	0-1
Yankees/1976	114.5%	110.3%	1-1
Phillies/1977	118.8%	105.4%	0-1
Yankees/1977	113.2%	112.5%	2-0
Phillies/1978	109.6%	107.2%	0-1
Roylas/1978	109.1%	109.0%	0-1
Pirates/1979	112.8%	109.4%	2-0
Reds/1979	107.4%	104.2%	0-1
Phillies/1980	111.4%	105.0%	2-0
Royals/1980	110.8%	105.2%	1-1
Brewers/1982	122.1%	102.3%	1-1
Angels/1982	112.0%	106.5%	0-1
Padres/1984	104.2%	103.2%	1-1
Tigers/1984	115.7%	114.3%	2-0
Red Sox/1986	107.0%	106.4%	1-1
Cardinals/1987	109.1%	104.3%	1-1
Tigers/1987	112.9%	110.9%	0-1
Red Sox/1988	115.2%	100.0%	0-1
Cubs/1989	109.9%	101.7%	0-1
Giants/1989	109.5%	101.7%	1-1
Pirates/1991	118.6%	107.0%	0-1
Braves/1991	112.8%	105.4%	1-1
Pirates/1992	110.3%	104.5%	0-1
Blue Jays/1992	111.4%	100.8%	2-0
Athletics/1992	106.4%	105.6%	0-1

Team/Year	Relative R/G	Relative ERA	Series W-L
Orioles/1969	117.6%	127.9%	1-1
Reds/1970	105.9%	109.8%	1-1
Orioles/1970	117.3%	117.8%	2-0
Twins/1970	110.2%	114.9%	0-1
Athletics/1971	111.0%	113.4%	0-1
Pirates/1972	114.1%	122.8%	0-1
Tigers/1972	103.2%	103.4%	0-1
Athletics/1972	112.4%	118.6%	2-0
Orioles/1973	108.8%	124.4%	0-1
Athletics/1973	109.4%	116.1%	2-0
Dodgers/1974	118.7%	121.9%	1-1
Athletics/1974	103.8%	122.7%	2-0
Pirates/1975	107.2%	120.3%	0-1
Athletics/1975	108.8%	115.6%	0-1
Dodgers/1977	107.9%	121.4%	1-1
Roylas/1977	112.0%	115.3%	0-1
Dodgers/1978	112.6%	114,4%	2-0
Yankees/1978	107.3%	117.9%	1-1
Orioles/1979	102.0%	129.4%	1-1
Yankees/1980	112.3%	112.6%	0-1
Cardinals/1982	103.4%	106.8%	2-0
Phillies/1983	104.1%	108.7%	1-1
Orioles/1983	110.9%	111.8%	2-0
Cardinals/1985	101.7%	115.8%	1-1
Blue Jays/1985	103.5%	125.4%	0-1
Mets/1986	115.7%	119.6%	2-0
Angels/1986	105.3%	108.9%	0-1
Giants/1987	107.0%	110.9%	0-1
Mets/1988	113.2%	118.6%	0-1
Athletics/1988	113.3%	115.4%	1-1
Blue Jays/1989	105.1%	108.4%	0-1
Athletics/1989	102.3%	125.3%	2-0
Pirates/1990	107.6%	111.5%	0-1
Reds/1990	101.7%	111.8%	2-0
Red Sox/1990	100.3%	105.1%	0-1
Athletics/1990	105.2%	117.0%	1-1
Twins/1991	106.8%	110.8%	2-0
Braves/1992	108.6%	111.5%	0-1