

RESEARCH QUESTION

In Major League Baseball, about 75% of the player pool is composed of players acquired through the amateur draft, held every June. The draft is the single most important amateur talent pipeline to MLB, and the core of each team's talent acquisition strategy, supplementing talent from the Dominican Republic, Cuba, and other places where talent can be signed outside the confines of the draft process. One common perception in the baseball community is that amateur - particularly high school - draftees from "cold weather" places are less likely to develop into major leaguers, due to only being able to play in the spring/summer versus year-round for their counterparts in "warm weather" locations such as Florida, California, Texas, and Arizona. This "cold weather bias" may affect the draft position of prospects from Northeastern and other cold weather states. The most prominent recent example of this phenomenon is Mike Trout of the Los Angeles Angels who was a highly regarded prospect in high school, but dropped all the way to 25th in the 1st round of the 2009 draft due in part to concerns that he played at a cold-weather high school in Millville, NJ. Trout thereafter became a superstar and several commentators pointed that his cold-weather school, in retrospect, contributed to him falling in the draft. The goal of this study is to investigate if there a cold weather effect in the draft positions of prospects, and if cold-weather state prospects are systemically undervalued by teams.

from not having great competition? Klaw (2:02 PM) In the south and west, yes, they'll get on the radar. Might be harder in cold-weather areas

because they start too late and by the time they get traction it's too late for them to get cross checked. But in states like FL, AZ, CA, TX, even the deep south. I don't think those players ca

Ask for an explanation of how Trout could still be available after 23 picks, and what you et sounds like a bunch of excuses:

- The weather is too unpredictable, so it's too hard for scouts to plan trips. • The weather is too cold and wet, so the player's body of work is limited compared
- to players in other parts of the country. • Pitchers from the northeast have historically done OK in pro ball, but position
- players have not. • Due to shorter seasons, some teams don't scout the area with any regularity.
- o a scouting director, those aren't excuses. They're realities of the job.

OF Garrett Whitley, Niskayuna HS (N.Y.)

B Analysis: Was said to be a candidate for the Diamondbacks as the first overall choice. Part of the eason interest in Whitley has been delayed is because he's from New York, and sometimes players from cold weather states aren't as widely scouted. An example: Mike Trout, who went in the first round but not until the 25th overall pick in 2009. Whitley has great bat speed and leg speed that is just as imposing, frequently beating out otherwise routine infield grounders. He's seen as a center fielder in the majors. Whitley batted .356, going 21 for 59, with three home runs, three doubles, a triple and 13

walks during his senior year. He's also hit a ball that was measured by a scout at 497 feet

Above: Examples of common "cold weather bias" lingo in MLB scouting community



121.4

87.8

101.3

81.6

67.1

71.9

71.7

85.8

70.2

94.5

118.5

104.4

64.2

67.2

71.7

60.9

42.4

55.4

57.8

84.9

land using NOAA 1981-2010 Climate Normals Data-



METHODOLOGY NOTES

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- The draft data consisted of 572 records, spanning all 1st rounds of MLB drafts from 2000-2012. Using more recent drafts could not yield reliable data on 3+ WAR players as many are yet to make their debuts. These players may still be prospects, and it is often too early to assign them to one of the "bust" or "3+ WAR" categories definitively.
- The data did NOT include duplicate records of players who were drafted once, did not sign, then were drafted again. ONLY records of players who signed were used.
- Data was limited to first round only; while there are certainly cases of All-Stars and other viable major leaguers being chosen in rounds 2, 3, 12, or later (e.g. Albert Pujols), the overwhelming majority of post-1st rounders do not make the majors, and moreover there is tremendous variability WITHIN the first round. The expected WAR of draft picks plummets after the first handful of picks, and the curve flattens out considerably through the end of Round 1, the supplemental picks, and then the rest of the draft.

DATA NOTES (2)

- that the (perceived) elite talent resides in the warmer climates of Florida, Texas, Southern California, and the Bay Area. In Phase 1, I isolated High School players since a lot of the cold-weather bias commentary concerns Northeastern high schoolers and their alleged lack of practice time, poor competition, and limited seasons. In Phase 2 I decided to include all records, since the data revealed there is a distinct lack of high
- A full 333 out of 572 draft records were in the "Tropical" category; this reflects the scouting truism
- draft picks from large, well-funded cold-weather colleges either. This is partly tautological, as a lot of the elite baseball powerhouse college programs are based in warmer climates.
- E.g., in my database of 1st rounders, there were more draft picks out of the 2 large Los Angeles schools (UCLA, USC) than the universities of Notre Dame, Nebraska, Michigan, Wisconsin, Minnesota, and Washington combined.

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Investigating Major League Baseball Amateur Draft **Patterns Pertaining to Climate** Searching for proof of the "Cold Weather Bias" in the MLB Draft

By Michael Dudkin—MGIS Student, Penn State



Mike Trout Millville, NJ 19th Pick, 2009 Draf 59.6 Career WAR

Should your team be wary of the next? Or excited to unearth the next?



DATA NOTES

- The raster surface created from NOAA data had to have breaks between climatic categories that fit the distribution of draft picks—therefore I created breaks in the surface categories on the lower ends of the "number of snowy days per year" scale. A location with 1 or fewer snowy days was considered "Tropical," 1-6 days categorized as "Nuisance," 6-15 days categorized as "On the Cold Side," 15-25 days as "Snowy," and finally 25+ days as "Igloo." The reason for choosing this metric over others (e.g. the average temperature metric used in Phase
- 1) is that it corresponds more to the traditional "cold weather bias" scouting objections—a snowy location not only indicates the presence of cold and lack of ability to play year round, but also limits potential practice and game opportunities in marginal months such as October and April, when weather could create adverse impacts not found better climates
- Because over half the players came from places with virtually no snow, I also re-grouped them into simple binary "Tropical" and "Non-Tropical" categories and then performed same analysis.

The climate surface analysis in this study was done with the GIS program ArcMap 10.6

The bulk NOAA climate normal datasets and baseball-reference were processed in Excel, and add-

ed as point layers in ArcMap. The draftees' hometowns were converted to lat/long coordinates with

The probabilities associated with a given bin of draft picks becoming a 3+ WAR major league con-

"calculated" probabilities using the 2000-2012 data alone, based on marginal totals, yielded similar

Climate surfaces were based on 6000+ data points from NOAA datasets, using the "Inverted Dis-

tance Weighed" (IDW) interpolation method, using an exponent of 2, to yield a relatively coarse sur-

tributor were taken from research done by the Hardball Times' Matthew Murphy (2014); the

results - 166 vs 172 expected major leaguers out of 572 based on each respective method.

face. The nearest 12 neighbor points were used in the interpolation.

TECHNICAL NOTES

widely available online "batch geocoders."

- Batch Geocoding Tool: http://www.mapdevelopers.com/batch_geocode_tool.php NOAA Data Explainer: https://www1.ncdc.noaa.gov/pub/data/normals/1981-2010/readme.txt Drafting Extrema: The Expectation of Draft Picks (2014). Retrieved from Beyond the Box Score: https:// www.beyondtheboxscore.com/2014/7/31/5950043/mlb-draft-picks-average-expectations-extremes Percentage Points of Chi-Square Distribution Table; From the University of Nebraska Lincoln: http://passel.unl.edu/ Image/Namuth-CovertDeana956176274/chi-sqaure%20distribution%20table.PNG

Chris Lubanski Norristown, PA 5th Pick, 2003 Draft 0 Career WAR

55%

43%

41%

32%

23%

15%

15%

Step 4 - Parse "average snowy days" data in 2 ways one granular by creating 5 different "snowiness" categories; and then binary (tropical vs non-tropical climate); create probabilities chart for "busts" in additional to 3+ WAR players; perform Chi-Square analysis Classification

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No. of players									
Wx_bin	Bin 1	Bin 2			5	Bin	6 Bi	n 7 Gi	and Total
Non-Tropical	39	31	42		72	74	4	13	333
Grand Total	65	59	63	64	125	123	3 7	/3	572
Observed maj	or leaguer	rs							
Wx_bin Tropical	Bin 1	Bin 2	Bin 3	Bin 4	Bin 5	Bin	6 Bi	n 7 Gi 8	and Total
Non-Tropical	16	15	7	8	14	7	•	7	74
Grand Total	37	31	24	20	31	20) 1	15	178
Observed bust	s (no MLE	3 games)							
Wx_bin	Bin 1	Bin 2	Bin 3	Bin 4	Bin 5	Bin	6 Bi	n7Gı	and Total
Non-Tropical	4	4	8 2	8	18	2/	1	5	75
Grand Total	10	10	10		15	51	. 3	3	172
						Í			
xpected busts if H() correct								
Vx_bin		Bin 1	Bin 2			Bin 5	Bin 6	Bin 7	Grand Tot
ropical		5.64	5.64			25.38	28.76	18.61	97.00
lon-Tropical		4.36	4.9			19.62	22.24	14.39	75.00
irand Total		10.00	10.00	Iu		45.00	51.00	33.00	172.00
xpected major lea	guers if HO o	orrect CHI							
Vx_bin		Bin 1	Bin 2	Bin 3	Bin 4	Bin 5	Bin 6	Bin 7	Grand Tot
ropical		0.02	0.25	0.63	0.01	0.07	0.15	0.07	1.19
lon-Tropical		0.02	0.35	0.89	0.01	0.10	0.21	0.09	1.67
irand Total		0.04	0.59	1.52	0.02	0.16	0.36	0.16	2.86
xpected busts if H) correct CH				_				
√x_bin		Bin 1	Bin 2	Bin 3	Bin 4	Bin 5	Bin 6	Bin 7	Grand Tot
ropical		0.02	0.02	0.99	0.74	0.10	0.11	0.02	2.01
an Transford		0.03	0.03	1 28	0.96	0.13	0 14	0.03	2 60

SPECIAL ACKNOWLEDGMENETS

1.70 0.24 0.25

This research is possible thanks to support from my Penn State instructors, advisors and administrators. Special thanks to my advisor Eliza Richardson, Justine Blanford, Beth King, Kary Isett, and my class instructors going back to 2014. Thank you to SA-BR's Scott Fischthal for guiding me in convention planning. A special thank you to my wife Megan and daughter Elise

null hypothesis Null hypothesis so on) within the school has no be ble major league games played)

Goal: Calculate

PHASE1: Drafted players WAR; Murphy p

All Players x2: 8 To be exceeded H0 cannot be r

HS Players only To be exceeded H0 cannot be r

PHASE2: Drafted players WAR; 5 catego All Players **x2**: To be exceede H0 cannot be i

Drafted players WAR; 5 categor All Players x2: 2 To be exceeded H0 cannot be i

Drafted players ML Games; 5 ca All Players x2: 2 To be exceede H0 cannot be re **Empiricall

egories: how lence level; t vation is simp

Drafted players WAR; 2 categor All Players x2: 2 To be exceeded H0 cannot be i

Drafted players ML Games; 2 categories; calculated probabilities) All Players **x2**: 4.60 To be exceeded: 12.59 **-10 cannot be rejected**

No. of players Wx_bin Snowy On the cold si Nuisance Tropical Grand Total Observed busts (no N Wx_bin Igloo Snowy On the Nuisance Tropical Grand To-Expected bust distr Wx_bin Igloo Snowy On the Nuisance Tropical Grand To- 10.00 HI-Square Values (Wx bin Bin 1 0.35 Igloo Snowy 0.81 3.25 On the Nuisance Tropical 0.02 Grand To- 5.01

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RESULTS chi-square statistic with which we are able to reject the t the 95% confidence level hat for any given range of slots (e.g. picks 1-5, 6-10 and first round, the climate associated with a drafted player's aring on the likelihood that that player will become a via- player (3WAR or greater), or be a draft Bust (No MLB
nd average winter temperature climate surface (3+ obabilities) 38 21.02
(2 : 12.11 (3+ WAR; Murphy probabilities) 21.02 jected
nd average number of snow days climate surface (3+ es; Murphy probabilities) 3.80 36.42 jected
nd average number of snow days climate surface (3+ es; calculated probabilities) 0.19 36.42 jected
nd average number of snow days climate surface (Zero egories; calculated probabilities) 0.12 36.42 ected
ere are FEWER busts than expected in the colder cat- er the result only exceeds threshold at 75% confi- refore we CANNOT reject assumption that this obser- a result of chance.**
nd average number of snow days climate surface (3+ es; calculated probabilities) 86 12.59 jected
nd average number of snow days climate surface (Zero

EX	AM	IPL	E

Bin 2	Bin 3	Bin 4	Bin 5	Bin 6	Bin 7	Grand Total
1	3	4	6	4	2	22
2	1	6	3	4	4	23
5	6	3	16	18	8	61
20	11	19	28	23	16	133
31	42	32	72	74	43	333
59	63	64	125	123	73	572
ames)						
Bin 2	Bin 3	Bin 4	Bin 5	Bin 6	Bin 7	Grand Total
0	0	1	2	2	1	6
0	0	3	0	4	7	14
1	1	0 -		6	2	10
1 2	1		11	12	З Д	36
6	8	5	27	27	18	97
U	0	5	27	27	10	57
10	10	13	45	51	33	172
n (if H0 cor-						
Bin 2	Bin 3	Bin 4	Bin 5	Bin 6	Bin 7	Grand Tota
0.35	0.35	0.45	1.57	1.78	1.15	6.00
0.81	0.81	1.06	3.66	4.15	2.69	14.00
				5.60	2.65	40.00
1.10	1.10	1.44	4.97	5.63	3.65	19.00
2.09	2.09	2.72	9.42	10.67	6.91	36.00
5.64	5.64	7.33	25.38	28.76	18.61	97.00
10.00	10.00	13.00	45.00	51.00	33.00	172.00
rees of Free	edom)					
Bin 2	Bin 3	Bin 4	Bin 5	Bin 6	Bin 7	Grand Total
0.35	0.35	0.66	0.12	0.03	0.02	1.87
0.81	0.81	3.56	3.66	0.01	6.93	16.60
0.01	0.01	1.44	0.00	0.02	0.44	4.05
0.01	0.01	1.44	0.00	0.02	0.11	4.85
1 39	0.57	0.60	0.27	0.16	1.22	3.79
0.00		0/4	0.10	0.11	0.02	2.01
0.02	0.99	0.74	0.10		0.02	

• Ruth E Hendricks Photography. PNC Park. Retrieved from https://rutheh.files.wordpress.com/2012/05/pncpark.jpg