## A Test of Replacement Level

I was browsing twitter and I saw a series of tweets made by Tom Tango and Mitchel Lichtman about realistic manifestations of replacement level. In one of the tweets, Tango linked to an ESPN page and stated, "Here are the players who signed a minor league free agent contract last season. I'd love for someone to total up their stats, including WAR, PA, and IP, to see how well rWAR and fWAR held up." I decided to test this.

I went to ESPN's record of free agent signings, which goes back to the 2006 offseason, and compiled all the minor league free agent signings from the website. I then collected from FanGraphs and Baseball-Reference, respectively, the pitcher and hitter data from the 2007-2017 seasons that contained the relevant data, namely WAR, IP, and PA. For all the minor league free agents that played in MLB the following season, I recorded for that season their fWAR and bWAR (pitchers' offense included) as well as the IP for pitchers and PA for both pitchers (if applicable) and hitters.

Now for the results: I calculated the sum of PA, IP and fWAR and bWAR for each player in each of those seasons. There were 328 players in the sample that compiled 25,096 PA and 7541 IP. The sums of fWAR and bWAR were around 51 and 55 WAR, respectively. Taking the fWAR and bWAR per player, the mean is about .16 WAR per player, with closer to .11 for position players and .2 for pitchers. About 13-14% of players compiled 1 WAR or greater.

What I expected was that the sums of each WAR construction would total closer to 0 WAR. Considering that the mean WAR per player is close to replacement level, the result is not wholly dissatisfying, but overall still may leave an explanation to be desired. One explanation is that there remains the possibility of some selection bias in the data. Obviously, the minor league free agents that were picked to play meaningful MLB innings were ones deemed by MLB teams to be the more promising of the crop, and any players who did not perform to teams' expected baseline were removed quickly and were never given an extended chance (or any chance at all). Furthermore, some of them may be players who were originally undervalued by the industry and forced to sign minor league deals, possibly because of poor performance the previous year or two due to random variation or injuries. A comparison of these players with their preseason projections would therefore be in order for clarification. Additionally, using two years of data instead of just the year afterward or a larger pool of players should also lead further away from potential bias. Another more unlikely explanation is that there is a flaw with the established replacement level.

Further research – Lichtman proposed comparing Steamer or ZIPS projections of these players to their actual performance. For the reasons that I alluded to earlier, since some players' projections

were well above replacement level, and they performed as such, teams may have undervalued them, and if some players' projections were well below replacement level, then teams may have overvalued them.

The method: I downloaded csv files from FG and BB-Ref for the seasonal data, and I copied and pasted the tables from the ESPN website (tedious but faster than figuring out how to scrape the data by my estimation). I used python pandas with jupyter notebook. Here are some sample screenshots. The github repo is located at the following URL: https://github.com/rontsung/replevel-test

```
In [1]: # dependencies
          import pandas as pd
          import json
          import datetime
In [164]: data = {}
          for i in range(2006,2017,1):
              # filter out the FA that signed minor league deals
              fa = pd.read csv(str(i)+"FA.csv")
              fa = fa[fa["PLAYER"]!="PLAYER"]
              fa = fa[fa["DOLLARS"]=="Minor Lg"]
              # compile all csv files into one dict - FA, FG and bbref batters and pitchers
              fbat = pd.read_csv(str(i+1)+" bleaders.csv")
              fpitch = pd.read_csv(str(i+1)+" pleaders.csv")
              bbat = pd.read_csv(str(i+1)+" bWAR batters.csv")
              bpitch = pd.read_csv(str(i+1)+" bWAR pitchers.csv")
              data[i] = [fa,{"FG":[fbat,fpitch], "BBREF":[bbat, bpitch]}]
          data
Out[164]: {2006: [
                               PLAYER POS AGE STATUS 2015 TEAM NEW TEAM YRS RK
                                                                                        DOLLARS
            0 Sandy Alomar Jr. C 51 Signed White Sox
                                                                 Mets NaN NR Minor Lg
                                                  Cardinals Cardinals NaN NR Minor Lg
                    Rick Ankiel CF 38 Signed
            1
                Ronnie Belliard 2B 42 Signed
                                                  Cardinals Nationals NaN NR Minor Lg
            4
            8
                     Bruce Chen SP 40 Signed
                                                    Orioles Rangers NaN NR Minor Lg
                                                               Rockies NaN NR Minor Lg
            9
                    Mike DeJean RP 47 Signed
                                                    Rockies
                 Einar Diaz C 45 Signed
Mike DiFelice C 48 Signed
            10
                                                    Dodgers Pirates NaN NR Minor Lg
            11
                                                      Mets
                                                                 Mets NaN NR Minor Lg
            12
                    Shawn Estes SP 45 Signed
                                                     Padres
                                                                Padres NaN NR Minor Lg
In [135]: # create dict with all players for each year - both their bWAR and fWAR as well as their PA and IP if applicable
          a = datetime.datetime.now()
          players = {}
          for e in data:
              cur = e+1
              players[cur] = [{}]
              fadata = data[e][0]
              seadata = data[e][1]
              # run through each free agent
              for index1, rowl in fadata.iterrows():
                  fa = row1["PLAYER"]
                  for site in seadata:
                     k = 0
                      # run through the seasonal data
                      for f in seadata[site]:
                          for index2, row2 in f.iterrows():
                              # identify the free agents that played in MLB that year
                              if fa[:fa.find(" ")] in row2["Name"] and fa[fa.find(" ")+1:] in row2["Name"] or fa == row2["Name"]:
                                 if fa not in players[cur][0]:
                                    players[cur][0][fa] = {"fWAR":0.0, "bWAR":0.0}
                                 war = row2["WAR"]
                                  # store the data in the dict
                                 if site == "FG":
                                     players[cur][0][fa]["fWAR"] += war
                                         players[cur][0][fa]["IP"] = row2["IP"]
                                     else:
                                        players[cur][0][fa]["PA"] = row2["PA"]
                                     if row2["WAR"] == row2["WAR"]:
                                         players[cur][0][fa]["bWAR"] += war
                         k+=1
          b = datetime.datetime.now()
          print(b-a)
          0:05:32.115728
```

```
In [137]: players
Out[137]: {2007: [{'Aaron Sele': {'IP': 53.2, 'PA': 6, 'bWAR': -0.3, 'fWAR': 0.0},
                              | 107: [('Aaron Sele': {'IP': 53.2, 'PA': 6, 'bWAR': -0.3, 'fWAR': 0.0] |
| Brian Moehler': {'IP': 59.2, 'PA': 5, 'bWAR': 0.6, 'fWAR': 0.0] |
| Bruce Chen': {'IP': 10.0, 'PA': 0, 'bWAR': -0.3, 'fWAR': -0.2], |
| Dan Kolb': {'IP': 3.0, 'PA': 0, 'bWAR': -0.1, 'fWAR': -0.1], 'David Newhan': {'PA': 83, 'bWaR': -0.4, 'fWAR': 0.0}, 'Eddie Guardado': {'IP': 13.2, 'PA': 0, 'bWAR': -0.3, 'fWAR': 0.1], 'Jamey Wright': {'IP': 77.0, 'PA': 2, 'bWAR': 1.4, 'fWAR': 0.4}, 'John Mabry': {'PA': 39, 'bWAR': -0.3, 'fWAR': -0.3}, 'Jon Knott': {'PA': 19, 'bWAR': 0.0, 'fWAR': 0.0}, 'Mark Mark Mark': 0'TP': 41.1.
                                'Mark Redman': {'IP': 41.1,
                                 'PA': 14,
                                 'bWAR': -0.89999999999999999999,
'fWAR': 0.19999999999999998},
                                'Matt Herges': {'IP': 48.2,
                                 'PA': 6,
                                 'bWAR': 1.0999999999999999
                                  'fwar': 0.300000000000000004}
                                'Matt Stairs': {'PA': 405, 'bWAR': 2.4, 'fWAR': 1.4},
                               'Mike Difelice': {'PA': 47, 'bWAR': 0.0, 'fWAR': 0.1},
In [153]: # for each year, add all the IP, PA, bWAR, and fWAR and store in dict
                        totals = {}
                        for e in players:
                                totals[e] = {"Total IP": 0, "Total PA": 0, "Total fWAR": 0, "Total bWAR": 0}
                                 for p in players[e][0]:
                                          totals[e]["Total fwar"] += players[e][0][p]["fwar"]
totals[e]["Total bwar"] += players[e][0][p]["bwar"]
                                          if "PA" in players[e][0][p]:
                                                  totals[e]["Total PA"] += players[e][0][p]["PA"]
                                          if "IP" in players[e][0][p]:
                                                  IP = players[e][0][p]["IP"]
if ".1" in str(IP):
                                                   IP += .23
elif ".2" in str(IP):
                                                          IP += .47
                                                   totals[e]["Total IP"] += IP
                       totals
In [154]: # for all the years, add the fWAR, bWAR, PA, and IP and store in dict
                       totalfwar = 0
                       totalbwar = 0
                       totalpa = 0
                       totalip = 0
                       for e in totals:
                               totalfwar += totals[e]["Total fWAR"]
totalbwar += totals[e]["Total bWAR"]
                               totalpa += totals[e]["Total PA"]
totalip += totals[e]["Total IP"]
                       fulltotal = {"Total fWAR": totalfwar, "Total bWAR": totalbwar, "Total PA": totalpa, "Total IP": totalip}
                      print(fulltotal)
                      {'Total fWAR': 50.80000000000000, 'Total bWAR': 55.000000000001, 'Total PA': 25096, 'Total IP': 7541.03}
In [177]: # calculate the WAR/player in several varieties, e.g. fWAR per player
                      fWARbatters = 0
                       fWARpitchers = 0
                      bWARpitchers = 0
                      bWARpitchers = 0
                       batters = 0
                      pitchers = 0
                       totalplayers = 0
                      for e in players:
                               for f in players[e]:
                                        totalplayers += len(f)
                       for e in players:
                               for p in players[e][0]:
    if "IP" in players[e][0][p]:
                                                 pitchers += 1
                                                  try:
                                                           fWARpitchers += players[e][0][p]["fWAR"]
bWARpitchers += players[e][0][p]["bWAR"]
                                                  except:
                                                           print("")
                      batters = totalplayers - pitchers
fWARbatters = totalfwar - fWARpitchers
bWARbatters = totalbwar - bWARpitchers
                      print (total fwar/total players, \ total bwar/total players, \ fWAR batters/batters, \ bwar batters/batters, \ fwar batters/batters, \ fwar batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/batters/
                      print(fWARbatters*600/totalpa, bWARbatters*600/totalpa)
                      0.49729040484539366 0.4112209116990755
```