

## add\_advanced\_pitching\_stats

April 29, 2020

```
[53]: import pandas as pd
import matplotlib.pyplot as plt
```

```
[54]: df = pd.read_csv('../core/output/pitchers.csv')
df_adv = pd.read_csv('../core/output/advanced_pitching.csv')
```

Adding Advanced Stats

```
[55]: df_adv.sort_values('retroID')
```

```
[55]:
```

	retroID	IP	K/9	BB/9	HR/9	BABIP	LOB%	ERA	FIP	WAR
2866	aardd001	0.062360	9.08	4.89	1.09	0.285	74.5	4.27	4.45	1.1
841	aased001	0.205233	5.20	3.71	0.72	0.282	73.4	3.80	3.85	11.7
3237	abadf001	0.061102	7.62	3.16	1.14	0.281	77.7	3.67	4.24	0.6
949	abbog001	0.237967	3.39	2.46	1.13	0.278	69.3	4.39	4.46	10.2
394	abboj001	0.309765	4.77	3.33	0.83	0.295	70.0	4.25	4.25	22.7
...	...	...	...	...	...	...	...	...	...	...
1030	zolds101	0.171925	2.00	2.91	0.52	0.267	70.7	3.54	3.80	9.3
1934	zubeb101	0.145445	4.39	5.36	0.40	0.283	69.0	4.28	3.96	3.3
2098	zumaj001	0.038711	9.01	4.89	0.77	0.267	78.7	3.00	3.94	2.7
2399	zuveg101	0.118817	3.12	2.84	0.78	0.270	73.2	3.54	3.93	1.9
2808	zycht001	0.013360	9.91	4.21	0.37	0.293	79.1	2.72	3.22	1.1

[8025 rows x 10 columns]

```
[56]: df
```

```
[56]:
```

	retroID	BAOpp	ERA	CG	SHO	IPouts	H	ER	HR	BB	...	WP	\
0	aardd001	0.2574	5.1944	0	0	1011	296	160	41	183	...	12	
1	aased001	0.2508	3.4931	22	5	3328	1085	468	89	457	...	22	
2	abadf001	0.2447	4.0810	0	0	992	309	135	42	116	...	10	
3	abbog001	0.2786	4.3317	37	5	3858	1405	627	162	352	...	18	
4	abboj001	0.2804	4.4964	31	6	5022	1779	791	154	620	...	53	
...	...	...	...	...	...	...	...	...	...	...	...	...	...
8020	zolds101	0.2700	3.6890	30	5	2788	956	366	54	301	...	8	
8021	zubeb101	0.2717	5.3617	23	3	2358	767	374	35	468	...	28	
8022	zumaj001	0.2286	3.4420	0	0	629	169	71	18	114	...	16	

8023	zuveg101	0.2760	4.1280	9	2	1927	660	253	56	203	...	10
8024	zycht001	0.2183	2.8000	0	0	218	57	22	3	34	...	2

	HBP	BK	BFP	GF	R	SH	SF	GIDP	K%
0	16	1	1475	141	169	17	11	21	0.230508
1	7	3	4730	235	503	50	34	106	0.135518
2	12	2	1399	97	143	7	12	25	0.200143
3	32	5	5508	13	707	60	39	111	0.087872
4	32	11	7211	5	880	70	47	200	0.123145
...	...	..	...	...	...	..	..	...	...
8020	3	4	3946	78	423	0	0	0	0.052458
8021	4	1	3476	90	418	0	0	0	0.110184
8022	4	0	911	35	80	6	10	10	0.230516
8023	27	1	2746	139	296	0	0	0	0.081209
8024	8	1	309	14	24	1	3	6	0.258900

[8025 rows x 22 columns]

```
[57]: df = df.drop(columns=['ERA'])
```

```
[58]: df = df.merge(df_adv, on='retroID', how='left')
```

```
[59]: 100 * df.isnull().sum() / len(df)
```

```
[59]: retroID      0.0
      BAOpp      0.0
      CG        0.0
      SHO        0.0
      IPouts     0.0
      H          0.0
      ER         0.0
      HR         0.0
      BB         0.0
      SO         0.0
      IBB        0.0
      WP         0.0
      HBP        0.0
      BK         0.0
      BFP        0.0
      GF         0.0
      R          0.0
      SH         0.0
      SF         0.0
      GIDP       0.0
      K%         0.0
      IP         0.0
      K/9        0.0
```

```

BB/9      0.0
HR/9      0.0
BABIP     0.0
LOB%      0.0
ERA        0.0
FIP        0.0
WAR        0.0
dtype: float64

```

```
[60]: df['Pitching'] = df[['K%', 'ERA', 'FIP', 'WAR']].mean(axis=1).round(3)
```

```
[61]: df['Pitching']
```

```

[61]: 0      2.513
      1      4.871
      2      2.178
      3      4.784
      4      7.831
      ...
      8020    4.173
      8021    2.913
      8022    2.468
      8023    2.363
      8024    1.825
      Name: Pitching, Length: 8025, dtype: float64

```

Finalizing the new Pitching statistic

```
[64]: df['Pitching'].mean()
```

```
[64]: 3.627380436137072
```

```
[65]: df['Pitching'].min()
```

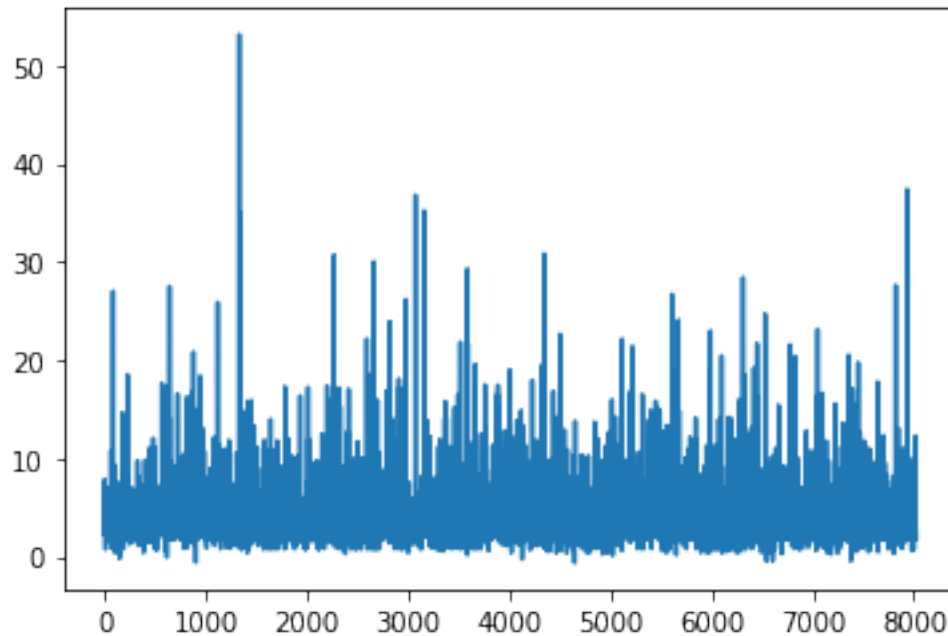
```
[65]: -0.654
```

```
[66]: df['Pitching'].max()
```

```
[66]: 53.138
```

```
[63]: plt.plot(df['Pitching'])
```

```
[63]: [<matplotlib.lines.Line2D at 0x127615250>]
```



```
[67]: df[df['Pitching'] == df['Pitching'].max()]
```

```
[67]:      retroID  BAOpp  CG  SHO  IPouts  H  ER  HR  BB  SO  ...      IP  \
1333  cleaj101   0.83   0    0         1  5   7   0   3   1  ...  0.000019

      K/9  BB/9  HR/9  BABIP  LOB%    ERA    FIP  WAR  Pitching
1333  27.0  81.0   0.0    1.0  12.5  189.0  23.54 -0.1    53.138
```

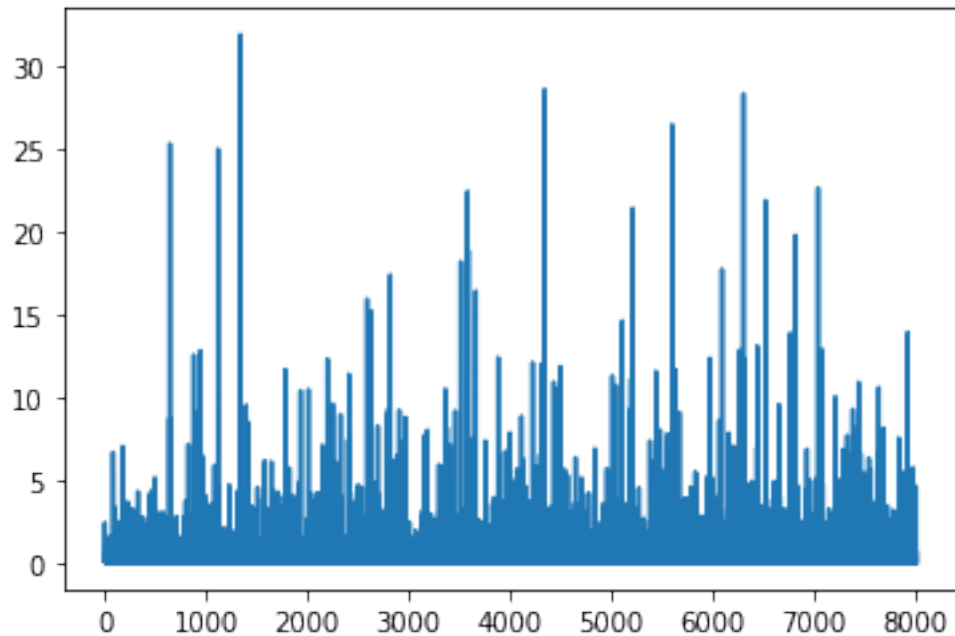
```
[1 rows x 31 columns]
```

This immediately demonstrates an issue with our Pitching stat - a player with very few appearances, but who did well in those appearances, will be skewed too high. We need to also take innings pitched into account.

```
[91]: df['Pitching'] = df[['K%', 'ERA', 'FIP', 'WAR']].mean(axis=1).round(3) * df['IP']
```

```
[92]: plt.plot(df['Pitching'])
```

```
[92]: [<matplotlib.lines.Line2D at 0x12793b850>]
```



```
[93]: df[df['Pitching'] == df['Pitching'].max()]
```

```
[93]:      retroID  BAOpp  CG  SHO  IPouts    H    ER  HR  BB  SO  ...  \
1341  clemr001  0.2308  118   46   14750  4185  1707  363  1580  4672  ...

      IP  K/9  BB/9  HR/9  BABIP  LOB%  ERA  FIP  WAR  Pitching
1341  0.909717  8.55  2.89  0.66  0.284  74.6  3.12  3.09  133.7  31.871935
```

```
[1 rows x 31 columns]
```

This looks like it worked, but let's explore deeper.

```
[94]: df['Pitching'].mean()
```

```
[94]: 0.49352977097414313
```

```
[95]: df['Pitching'].min()
```

```
[95]: -5.0887375e-05
```

```
[96]: df['Pitching'].max()
```

```
[96]: 31.871935094999998
```

```
[97]: df.sort_values('Pitching').tail(10)
```

```
[97]:      retroID  BAOpp  CG  SHO  IPouts      H    ER   HR   BB   SO   ...  \
5219  niekp001  0.2570  245   45   16213  5044  2012  482  1809  3342  ...
6529  seavt001  0.2285  231   61   14348  3971  1521  380  1390  3640  ...
3588  johnr005  0.2252  100   37   12406  3346  1513  411  1497  4875  ...
7049  suttd001  0.2376  178   58   15847  4692  1914  472  1343  3574  ...
1121  carls001  0.2546  254   55   15652  4672  1864  414  1833  4136  ...
645   blylb001  0.2487  242   60   14910  4632  1830  430  1322  3701  ...
5607  perrg101  0.2540  303   53   16051  4938  1846  399  1379  3534  ...
6311  ryann001  0.2088  222   61   16158  3923  1911  321  2795  5714  ...
4347  maddg002  0.2553  109   35   15025  4726  1756  353   999  3371  ...
1341  clemr001  0.2308  118   46   14750  4185  1707  363  1580  4672  ...
```

```
      IP    K/9  BB/9  HR/9  BABIP  LOB%   ERA   FIP   WAR  Pitching
5219  1.000000   5.57  3.01  0.80  0.270  73.6  3.35  3.62   78.5  21.404000
6529  0.884921   6.85  2.62  0.72  0.259  76.7  2.86  3.04   92.7  21.854894
3588  0.765178  10.61  3.26  0.89  0.291  74.7  3.29  3.19  110.4  22.412829
7049  0.977406   6.09  2.29  0.80  0.261  73.6  3.26  3.24   85.9  22.618152
1121  0.965397   7.13  3.16  0.71  0.279  74.1  3.22  3.15   96.9  24.969993
645   0.919672   6.70  2.39  0.78  0.282  74.1  3.31  3.19  103.3  25.286382
5607  0.990008   5.94  2.32  0.67  0.275  73.3  3.11  3.06  100.5  26.441134
6311  0.996651   9.55  4.67  0.54  0.265  73.1  3.19  2.97  107.2  28.307878
4347  0.926722   6.06  1.80  0.63  0.281  72.3  3.16  3.26  116.7  28.562499
1341  0.909717   8.55  2.89  0.66  0.284  74.6  3.12  3.09  133.7  31.871935
```

[10 rows x 31 columns]

```
[98]: df[df['retroID'] == 'kersc001']
```

```
[98]:      retroID  BAOpp  CG  SHO  IPouts      H    ER   HR   BB   SO   ...  \
3761  kersc001  0.2105  25   15   6824  1715  617  173  577  2464  ...

      IP    K/9  BB/9  HR/9  BABIP  LOB%   ERA   FIP   WAR  Pitching
3761  0.420829   9.75  2.28  0.68   0.27  79.4  2.44  2.74   64.5  7.359878
```

[1 rows x 31 columns]

This looks good, but we intuitively see a problem with the Pitching stat. ERA and FIP are part of the average, but a low ERA/FIP is better than a high one. We need to subtract them rather than add. With this change, I'm going to see how the stat looks without taking IP into account.

```
[132]: df['-ERA'] = 0 - df['ERA']
df['-FIP'] = 0 - df['FIP']
```

```
[133]: df
```

```
[133]:      retroID  BAOpp  CG  SHO  IPouts      H    ER   HR   BB   SO   ...  BB/9  \
0        aardd001  0.2574   0    0   1011   296  160   41  183  340   ...  4.89
```

1	aased001	0.2508	22	5	3328	1085	468	89	457	641	...	3.71
2	abadf001	0.2447	0	0	992	309	135	42	116	280	...	3.16
3	abbog001	0.2786	37	5	3858	1405	627	162	352	484	...	2.46
4	abboj001	0.2804	31	6	5022	1779	791	154	620	888	...	3.33
...	...	...	...	...	...	...	...	...	...	...	...	...
8020	zolds101	0.2700	30	5	2788	956	366	54	301	207	...	2.91
8021	zubeb101	0.2717	23	3	2358	767	374	35	468	383	...	5.36
8022	zumaj001	0.2286	0	0	629	169	71	18	114	210	...	4.89
8023	zuveg101	0.2760	9	2	1927	660	253	56	203	223	...	2.84
8024	zycht001	0.2183	0	0	218	57	22	3	34	80	...	4.21

	HR/9	BABIP	LOB%	ERA	FIP	WAR	Pitching	-ERA	-FIP
0	1.09	0.285	74.5	4.27	4.45	1.1	-1.847	-4.27	-4.45
1	0.72	0.282	73.4	3.80	3.85	11.7	1.046	-3.80	-3.85
2	1.14	0.281	77.7	3.67	4.24	0.6	-1.777	-3.67	-4.24
3	1.13	0.278	69.3	4.39	4.46	10.2	0.359	-4.39	-4.46
4	0.83	0.295	70.0	4.25	4.25	22.7	3.581	-4.25	-4.25
...	...	...	...	...	...	...	...	...	...
8020	0.52	0.267	70.7	3.54	3.80	9.3	0.503	-3.54	-3.80
8021	0.40	0.283	69.0	4.28	3.96	3.3	-1.207	-4.28	-3.96
8022	0.77	0.267	78.7	3.00	3.94	2.7	-1.002	-3.00	-3.94
8023	0.78	0.270	73.2	3.54	3.93	1.9	-1.372	-3.54	-3.93
8024	0.37	0.293	79.1	2.72	3.22	1.1	-1.145	-2.72	-3.22

[8025 rows x 33 columns]

```
[134]: df['Pitching'] = df[['K%', '-ERA', '-FIP', 'WAR']].mean(axis=1).round(3)
```

```
[135]: df.sort_values('Pitching').tail(10)
```

```
[135]:      retroID  BAOpp  CG  SHO  IPouts    H    ER  HR  BB  SO  ... \
7049  suttd001  0.2376  178   58  15847  4692  1914  472  1343  3574  ...
2824  grovl101  0.2535  298   35  11822  3849  1339  162  1187  2266  ...
6529  seavt001  0.2285  231   61  14348  3971  1521  380  1390  3640  ...
1121  carls001  0.2546  254   55  15652  4672  1864  414  1833  4136  ...
5607  perrg101  0.2540  303   53  16051  4938  1846  399  1379  3534  ...
645   blylb001  0.2487  242   60  14910  4632  1830  430  1322  3701  ...
6311  ryann001  0.2088  222   61  16158  3923  1911  321  2795  5714  ...
3588  johnr005  0.2252  100   37  12406  3346  1513  411  1497  4875  ...
4347  maddg002  0.2553  109   35  15025  4726  1756  353  999  3371  ...
1341  clemr001  0.2308  118   46  14750  4185  1707  363  1580  4672  ...
```

	BB/9	HR/9	BABIP	LOB%	ERA	FIP	WAR	Pitching	-ERA	-FIP
7049	2.29	0.80	0.261	73.6	3.26	3.24	85.9	19.891	-3.26	-3.24
2824	2.71	0.37	0.284	71.8	3.06	3.36	88.8	20.629	-3.06	-3.36
6529	2.62	0.72	0.259	76.7	2.86	3.04	92.7	21.747	-2.86	-3.04
1121	3.16	0.71	0.279	74.1	3.22	3.15	96.9	22.680	-3.22	-3.15

5607	2.32	0.67	0.275	73.3	3.11	3.06	100.5	23.623	-3.11	-3.06
645	2.39	0.78	0.282	74.1	3.31	3.19	103.3	24.245	-3.31	-3.19
6311	4.67	0.54	0.265	73.1	3.19	2.97	107.2	25.323	-3.19	-2.97
3588	3.26	0.89	0.291	74.7	3.29	3.19	110.4	26.051	-3.29	-3.19
4347	1.80	0.63	0.281	72.3	3.16	3.26	116.7	27.611	-3.16	-3.26
1341	2.89	0.66	0.284	74.6	3.12	3.09	133.7	31.930	-3.12	-3.09

[10 rows x 33 columns]

```
[136]: df[df['retroID'] == 'kersc001']
```

```
[136]:      retroID  BAOpp  CG  SHO  IPouts    H  ER  HR  BB  SO  ...  BB/9  \
3761  kersc001  0.2105  25   15   6824  1715  617  173  577  2464  ...  2.28

      HR/9  BABIP  LOB%   ERA   FIP   WAR  Pitching  -ERA  -FIP
3761  0.68   0.27  79.4  2.44  2.74  64.5    14.899 -2.44 -2.74
```

[1 rows x 33 columns]

```
[137]: df[df['retroID'] == 'johnr005']
```

```
[137]:      retroID  BAOpp  CG  SHO  IPouts    H  ER  HR  BB  SO  ...  \
3588  johnr005  0.2252  100  37  12406  3346  1513  411  1497  4875  ...

      BB/9  HR/9  BABIP  LOB%   ERA   FIP   WAR  Pitching  -ERA  -FIP
3588  3.26  0.89  0.291  74.7  3.29  3.19  110.4    26.051 -3.29 -3.19
```

[1 rows x 33 columns]

```
[138]: df[df['retroID'] == 'bumgm001']
```

```
[138]:      retroID  BAOpp  CG  SHO  IPouts    H  ER  HR  BB  SO  ...  BB/9  \
942  bumgm001  0.2358  15   6   5538  1622  642  192  428  1794  ...  2.09

      HR/9  BABIP  LOB%   ERA   FIP   WAR  Pitching  -ERA  -FIP
942  0.94  0.284  76.4  3.13  3.32  31.3    6.272 -3.13 -3.32
```

[1 rows x 33 columns]

```
[139]: df[df['retroID'] == 'mahop001']
```

```
[139]:      retroID  BAOpp  CG  SHO  IPouts    H  ER  HR  BB  SO  ...  BB/9  \
4377  mahop001  0.2751   0   0   2127  738  431  116  392  452  ...  4.98

      HR/9  BABIP  LOB%   ERA   FIP  WAR  Pitching  -ERA  -FIP
4377  1.47  0.284  69.5  5.47  5.62 -3.0    -3.487 -5.47 -5.62
```



[1 rows x 33 columns]

```
[140]: df[df['retroID'] == 'coleg001']
```

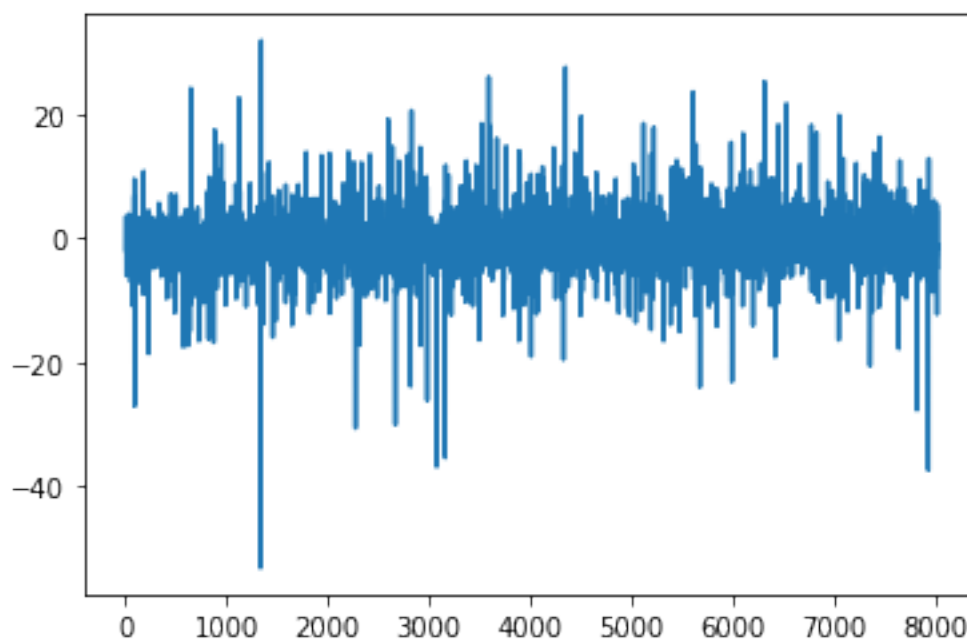
```
[140]:      retroID  BAOpp  CG  SHO  IPouts    H  ER  HR  BB  SO  ...  BB/9  \
1383  coleg001  0.2381  2    1   3585  1034  427  115  315  1336  ...  2.37

      HR/9  BABIP  LOB%   ERA   FIP   WAR  Pitching  -ERA  -FIP
1383  0.87  0.303  75.6  3.22  3.06  28.8    5.699 -3.22 -3.06
```

[1 rows x 33 columns]

```
[141]: plt.plot(df['Pitching'])
```

```
[141]: [<matplotlib.lines.Line2D at 0x12a1537d0>]
```



```
[131]: df.sort_values('Pitching').head(10)
```

```
[131]:      retroID  BAOpp  CG  SHO  IPouts  H  ER  HR  BB  SO  ...  BB/9  HR/9  \
1333  cleaj101   0.83   0    0         1  5   7   0   3   1  ...   81.0   0.0
7932  wurmf101   0.50   0    0         1  1   4   0   5   1  ...  135.0   0.0
3074  heart001   0.75   0    0         1  3   4   0   4   0  ...  108.0   0.0
3159  herne001   0.50   0    0         1  1   3   1   2   0  ...   54.0  27.0
2272  fishf101   0.66   0    0         1  2   4   0   2   1  ...   54.0   0.0
2666  gomec002   0.00   0    0         1  0   3   0   4   0  ...  108.0   0.0
7819  wilst104   0.00   0    0         1  0   3   0   2   0  ...   54.0   0.0
```

88	alexm001	0.50	0	0	2	1	5	1	4	0	...	54.0	13.5
2981	harll101	0.50	0	0	2	2	5	1	4	1	...	54.0	13.5
5674	pickr001	0.60	0	0	2	3	6	0	4	2	...	54.0	0.0

	BABIP	LOB%	ERA	FIP	WAR	Pitching	-ERA	-FIP
1333	1.00	12.5	189.0	23.54	-0.1	-53.132	-189.0	-23.54
7932	1.00	33.3	108.0	41.54	-0.1	-37.374	-108.0	-41.54
3074	0.75	28.6	108.0	39.21	-0.1	-36.828	-108.0	-39.21
3159	0.00	0.0	81.0	60.16	-0.3	-35.365	-81.0	-60.16
2272	1.00	0.0	108.0	14.60	0.0	-30.600	-108.0	-14.60
2666	0.00	25.0	81.0	39.16	-0.1	-30.065	-81.0	-39.16
7819	0.00	0.0	81.0	29.57	-0.1	-27.667	-81.0	-29.57
88	0.00	0.0	67.5	40.67	-0.1	-27.068	-67.5	-40.67
2981	0.50	21.7	67.5	37.08	-0.1	-26.139	-67.5	-37.08
5674	1.00	14.3	81.0	15.14	0.0	-23.979	-81.0	-15.14

[10 rows x 33 columns]

I'm happy with this. The new Pitching stat somewhat reflects its component parts but doesn't immediately align with WAR. We don't just want to recreate WAR so that's a good thing.

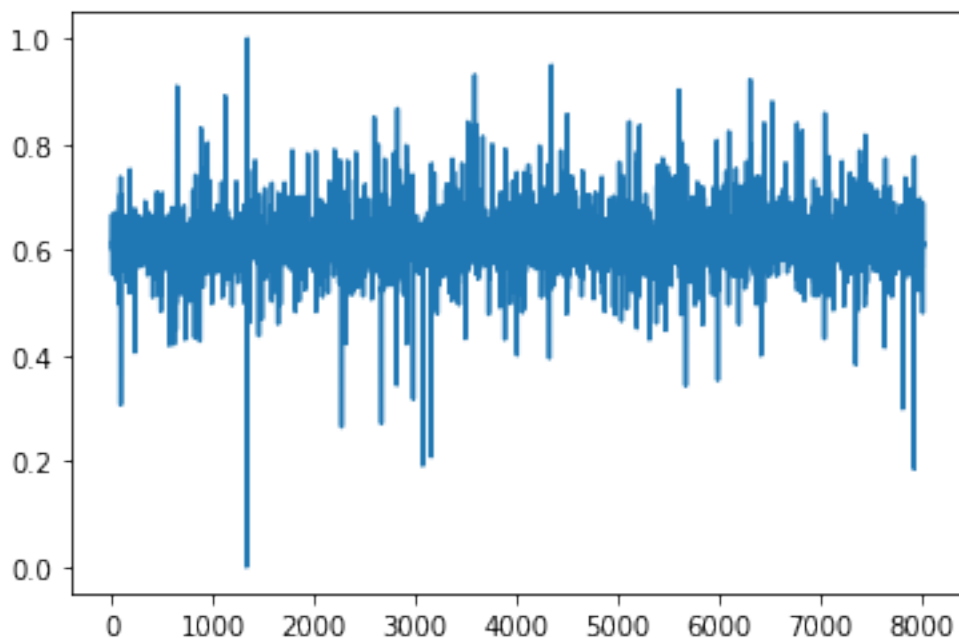
Normalization

```
[142]: from sklearn.preprocessing import MinMaxScaler
```

```
[143]: scaler = MinMaxScaler()
```

```
[144]: plt.plot(scaler.fit_transform(df[['Pitching']]))
```

```
[144]: [<matplotlib.lines.Line2D at 0x1337d3f50>]
```



```
[145]: df['Pitching'] = scaler.fit_transform(df[['Pitching']])
```

```
[147]: df.sort_values('Pitching').head(10)
```

```
[147]:
```

	retroID	BAOpp	CG	SHO	IPouts	H	ER	HR	BB	SO	...	BB/9	HR/9	\
1333	cleaj101	0.83	0	0	1	5	7	0	3	1	...	81.0	0.0	
7932	wurmf101	0.50	0	0	1	1	4	0	5	1	...	135.0	0.0	
3074	heart001	0.75	0	0	1	3	4	0	4	0	...	108.0	0.0	
3159	herne001	0.50	0	0	1	1	3	1	2	0	...	54.0	27.0	
2272	fishf101	0.66	0	0	1	2	4	0	2	1	...	54.0	0.0	
2666	gomec002	0.00	0	0	1	0	3	0	4	0	...	108.0	0.0	
7819	wilst104	0.00	0	0	1	0	3	0	2	0	...	54.0	0.0	
88	alexm001	0.50	0	0	2	1	5	1	4	0	...	54.0	13.5	
2981	harll101	0.50	0	0	2	2	5	1	4	1	...	54.0	13.5	
5674	pickr001	0.60	0	0	2	3	6	0	4	2	...	54.0	0.0	

	BABIP	LOB%	ERA	FIP	WAR	Pitching	-ERA	-FIP
1333	1.00	12.5	189.0	23.54	-0.1	0.000000	-189.0	-23.54
7932	1.00	33.3	108.0	41.54	-0.1	0.185253	-108.0	-41.54
3074	0.75	28.6	108.0	39.21	-0.1	0.191672	-108.0	-39.21
3159	0.00	0.0	81.0	60.16	-0.3	0.208871	-81.0	-60.16
2272	1.00	0.0	108.0	14.60	0.0	0.264889	-108.0	-14.60
2666	0.00	25.0	81.0	39.16	-0.1	0.271179	-81.0	-39.16
7819	0.00	0.0	81.0	29.57	-0.1	0.299370	-81.0	-29.57
88	0.00	0.0	67.5	40.67	-0.1	0.306412	-67.5	-40.67
2981	0.50	21.7	67.5	37.08	-0.1	0.317333	-67.5	-37.08

```
5674    1.00  14.3   81.0  15.14  0.0  0.342726  -81.0 -15.14
```

```
[10 rows x 33 columns]
```

```
[148]: df.sort_values('Pitching').tail(10)
```

```
[148]:      retroID  BAOpp  CG  SHO  IPouts    H    ER  HR  BB  SO  ...  \
7049  suttd001  0.2376  178   58   15847  4692  1914  472  1343  3574  ...
2824  grovl101  0.2535  298   35   11822  3849  1339  162  1187  2266  ...
6529  seavt001  0.2285  231   61   14348  3971  1521  380  1390  3640  ...
1121  carls001  0.2546  254   55   15652  4672  1864  414  1833  4136  ...
5607  perrg101  0.2540  303   53   16051  4938  1846  399  1379  3534  ...
645   blylb001  0.2487  242   60   14910  4632  1830  430  1322  3701  ...
6311  ryann001  0.2088  222   61   16158  3923  1911  321  2795  5714  ...
3588  johnr005  0.2252  100   37   12406  3346  1513  411  1497  4875  ...
4347  maddg002  0.2553  109   35   15025  4726  1756  353   999  3371  ...
1341  clemr001  0.2308  118   46   14750  4185  1707  363  1580  4672  ...
```

```
      BB/9  HR/9  BABIP  LOB%  ERA  FIP  WAR  Pitching  -ERA  -FIP
7049  2.29  0.80  0.261  73.6  3.26  3.24  85.9  0.858468 -3.26 -3.24
2824  2.71  0.37  0.284  71.8  3.06  3.36  88.8  0.867144 -3.06 -3.36
6529  2.62  0.72  0.259  76.7  2.86  3.04  92.7  0.880287 -2.86 -3.04
1121  3.16  0.71  0.279  74.1  3.22  3.15  96.9  0.891256 -3.22 -3.15
5607  2.32  0.67  0.275  73.3  3.11  3.06  100.5  0.902342 -3.11 -3.06
645   2.39  0.78  0.282  74.1  3.31  3.19  103.3  0.909654 -3.31 -3.19
6311  4.67  0.54  0.265  73.1  3.19  2.97  107.2  0.922327 -3.19 -2.97
3588  3.26  0.89  0.291  74.7  3.29  3.19  110.4  0.930886 -3.29 -3.19
4347  1.80  0.63  0.281  72.3  3.16  3.26  116.7  0.949225 -3.16 -3.26
1341  2.89  0.66  0.284  74.6  3.12  3.09  133.7  1.000000 -3.12 -3.09
```

```
[10 rows x 33 columns]
```

Finally, we should get rid of the -ERA and -FIP columns.

```
[149]: df = df.drop(columns=['-ERA', '-FIP'])
```

```
[150]: df
```

```
[150]:      retroID  BAOpp  CG  SHO  IPouts    H    ER  HR  BB  SO  ...  \
0      aardd001  0.2574   0   0    1011   296   160   41  183  340  ...
1      aased001  0.2508  22   5    3328  1085   468   89  457  641  ...
2      abadf001  0.2447   0   0     992   309   135   42  116  280  ...
3      abbog001  0.2786  37   5    3858  1405   627  162  352  484  ...
4      abboj001  0.2804  31   6    5022  1779   791  154  620  888  ...
...      ...      ...  ..  ...      ...      ...      ...      ...      ...      ...
8020  zolds101  0.2700  30   5    2788   956   366   54  301  207  ...
8021  zubeb101  0.2717  23   3    2358   767   374   35  468  383  ...
```

8022	zumaj001	0.2286	0	0	629	169	71	18	114	210	...
8023	zuveg101	0.2760	9	2	1927	660	253	56	203	223	...
8024	zycht001	0.2183	0	0	218	57	22	3	34	80	...

	IP	K/9	BB/9	HR/9	BABIP	LOB%	ERA	FIP	WAR	Pitching
0	0.062360	9.08	4.89	1.09	0.285	74.5	4.27	4.45	1.1	0.602913
1	0.205233	5.20	3.71	0.72	0.282	73.4	3.80	3.85	11.7	0.636924
2	0.061102	7.62	3.16	1.14	0.281	77.7	3.67	4.24	0.6	0.603736
3	0.237967	3.39	2.46	1.13	0.278	69.3	4.39	4.46	10.2	0.628847
4	0.309765	4.77	3.33	0.83	0.295	70.0	4.25	4.25	22.7	0.666725
...	...	...	...	...	...	...	...	...	...	...
8020	0.171925	2.00	2.91	0.52	0.267	70.7	3.54	3.80	9.3	0.630540
8021	0.145445	4.39	5.36	0.40	0.283	69.0	4.28	3.96	3.3	0.610437
8022	0.038711	9.01	4.89	0.77	0.267	78.7	3.00	3.94	2.7	0.612847
8023	0.118817	3.12	2.84	0.78	0.270	73.2	3.54	3.93	1.9	0.608497
8024	0.013360	9.91	4.21	0.37	0.293	79.1	2.72	3.22	1.1	0.611166

[8025 rows x 31 columns]

[ ]: