```
In [1]:
           import pandas as pd
           import numpy as np
           import matplotlib.pyplot as py
           import seaborn as sb
In [2]:
           # read in the csv file
           data = pd.read_csv("/users/baylife/desktop/csv files/nfl_pass_rush_receive_raw
           type(data)
Out[2]: pandas.core.frame.DataFrame
In [3]:
           # display the columns of the data
           data.columns
Out[3]: Index(['game_id', 'player_id', 'pos', 'player', 'team', 'pass_cmp', 'pass_at
                   'pass_yds', 'pass_td', 'pass_int', 'pass_sacked', 'pass_sacked_yds',
                   'pass_long', 'pass_rating', 'rush_att', 'rush_yds', 'rush_td', 'rush_long', 'targets', 'rec', 'rec_yds', 'rec_td', 'rec_long',
                   'fumbles_lost', 'rush_scrambles', 'designed_rush_att',
                   'comb_pass_rush_play', 'comb_pass_play', 'comb_rush_play',
                   'Team_abbrev', 'Opponent_abbrev', 'two_point_conv', 'total_ret_td', 'offensive_fumble_recovery_td', 'pass_yds_bonus', 'rush_yds_bonus',
                   'rec_yds_bonus', 'Total_DKP', 'Off_DKP', 'Total_FDP', 'Off_FDP',
'Total_SDP', 'Off_SDP', 'pass_target_yds', 'pass_poor_throws',
                   'pass_blitzed', 'pass_hurried', 'rush_yds_before_contact', 'rush_yac',
                   'rush_broken_tackles', 'rec_air_yds', 'rec_yac', 'rec_drops', 'offens
          e',
                  'off_pct', 'vis_team', 'home_team', 'vis_score', 'home_score', 'OT', 'Roof', 'Surface', 'Temperature', 'Humidity', 'Wind_Speed', 'Vegas_Line', 'Vegas_Favorite', 'Over_Under', 'game_date'],
                 dtype='object')
In [4]:
           # rename the a few of the columns, then display the number of rows in the file
           data.rename(columns={'pos':'position','game_date':'date'},inplace=True)
           print("number of row in the data set:", len(data))
          number of row in the data set: 18481
In [5]:
           # sort the dataFrame by position, then by name and lastly the date
           data.sort_values(by=['position','player','date'],inplace = True)
In [6]:
           #print data
           data
```

game\_id player\_id position player team pass\_cmp pass\_att pass\_yds pass Out[6]: Aaron 10939 202012060oti BrewAa01 С TEN 0 0 0 Brewer Garrett **4776** 201912080min BradGa00 MIN 0 0 С 0 Bradbury Garrett 6209 202001110sfo BradGa00 С MIN 0 0 Bradbury Nick **11721** 202012200nyg С NYG 0 0 GateNi00 0 Gates Matt 202011080clt SkurMa01 C/G 9355 BAL 0 0 0 Skura ... ... ... ... ... ... Willie 9216 202011010rav SneaWi00 WR/R BAL 0 0 0 Snead Willie 9352 202011080clt SneaWi00 WR/R  $\mathsf{BAL}$ 0 0 0 Snead Willie **9822** 202011150nwe SneaWi00 0 WR/R BAL 0 0 Snead Willie **10223** 202011220rav SneaWi00 WR/R BAL Snead Willie **12799** 202101160buf SneaWi00 WR/R BAL 0 0 0 Snead 18481 rows × 69 columns In [7]: #check if the column names were changed data.columns 'comb\_pass\_rush\_play', 'comb\_pass\_play', 'comb\_rush\_play', 'Team\_abbrev', 'Opponent\_abbrev', 'two\_point\_conv', 'total\_ret\_td', 'offensive\_fumble\_recovery\_td', 'pass\_yds\_bonus', 'rush\_yds\_bonus', 'rec\_yds\_bonus', 'Total\_DKP', 'Off\_DKP', 'Total\_FDP', 'Off\_FDP', 'Total\_SDP', 'Off\_SDP', 'pass\_target\_yds', 'pass\_poor\_throws', 'pass\_blitzed', 'pass\_hurried', 'rush\_yds\_before\_contact', 'rush\_yac', 'rush\_broken\_tackles', 'rec\_air\_yds', 'rec\_yac', 'rec\_drops', 'offens e', 'off\_pct', 'vis\_team', 'home\_team', 'vis\_score', 'home\_score', 'OT', 'Roof', 'Surface', 'Temperature', 'Humidity', 'Wind\_Speed', 'Vegas\_Line', 'Vegas\_Favorite', 'Over\_Under', 'date'],

```
In [8]: # get the list of players who are QBs
    df = data[data["position"]=='QB']

In [9]: # sort the QB's by name
    df.sort_values(by=['player'])
```

dtype='object')

game\_id player\_id position

Out[9]:

	3724	201911170rav	McCaA.00	QB	A.J. McCarron	HOU	0	1	0		
	5864	201912290htx	McCaA.00	QB	A.J. McCarron	HOU	21	36	225		
	11160	202012130chi	McCaA.00	QB	A.J. McCarron	HOU	0	0	0		
	12474	202101030htx	McCaA.00	QB	A.J. McCarron	HOU	1	1	20		
	10045	202011220clt	RodgAa00	QB	Aaron Rodgers	GNB	27	38	311		
	•••	•••		•••		•••		•••			
	13825	202109260den	WilsZa00	QB	Zach Wilson	NYJ	19	35	160		
	13032	202109120car	WilsZa00	QB	Zach Wilson	NYJ	20	37	258		
	17958	202112120nyj	WilsZa00	QB	Zach Wilson	NYJ	19	42	202		
	14418	202110030nyj	WilsZa00	QB	Zach Wilson	NYJ	21	34	297		
	18248	202112190mia	WilsZa00	QB	Zach Wilson	NYJ	13	23	170		
	1849 ro	ws × 69 columr	S	)							
In [10]:		by the number_downs = df.s			pass_td'])						
In [11]:	<pre># save the columns of interest in an array, create new dataframe for the quate interest = ['position', 'player', 'team', 'pass_cmp',</pre>										
In [12]:	_	the QB who pta.loc[:,"pla	_		games: t	che fre	quence of t	he name	that appe		
Out[12]:	count unique top freq Name:	1849 114 Tom Brady 51 player, dtype	•								

# get the collection of data for the QB who played in the most amount of games

most\_prominent = QB\_data.loc[:,"player"].describe()
brady = QB\_data[QB\_data['player'] == most\_prominent[2]]

player team pass\_cmp pass\_att pass\_yds pas

brady

In [13]:

Out[13]:

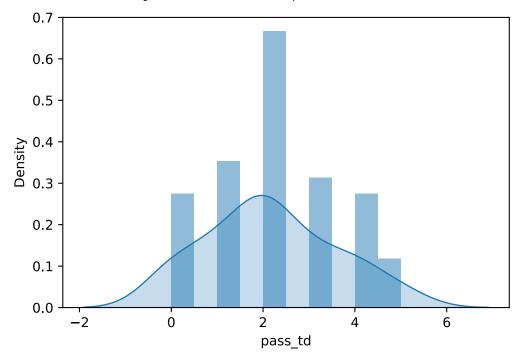
	position	player	team	pass_cmp	pass_att	pass_yds	pass_td	pass_int	pass_rating
210	QB	Tom Brady	NWE	24	36	341	3	0	124.9
555	QB	Tom Brady	NWE	20	28	264	2	0	124.7
977	QB	Tom Brady	NWE	28	42	306	2	0	103.9
1202	QB	Tom Brady	NWE	18	39	150	0	1	45.9
1787	QB	Tom Brady	NWE	28	42	348	3	1	106.1
1846	QB	Tom Brady	NWE	31	41	334	0	1	88.9
2485	QB	Tom Brady	NWE	31	45	249	1	1	80.7
2763	QB	Tom Brady	NWE	20	36	259	2	0	96.9
3100	QB	Tom Brady	NWE	30	46	285	1	1	80.4
3669	QB	Tom Brady	NWE	26	47	216	0	0	67.3
4017	QB	Tom Brady	NWE	17	37	190	1	0	70.8
4370	QB	Tom Brady	NWE	24	47	326	3	1	85.9
4809	QB	Tom Brady	NWE	19	36	169	1	1	63.3
4987	QB	Tom Brady	NWE	15	29	128	2	0	86.6
5324	QB	Tom Brady	NWE	26	33	271	1	0	111.0
5952	QB	Tom Brady	NWE	16	29	221	2	1	88.4
6115	QB	Tom Brady	NWE	20	37	209	0	1	59.4
6547	QB	Tom Brady	TAM	23	36	239	2	2	78.4
7087	QB	Tom Brady	TAM	23	35	217	1	1	80.3
7275	QB	Tom Brady	TAM	25	38	297	3	0	115.8
7809	QB	Tom Brady	TAM	30	46	369	5	1	117.0
7894	QB	Tom Brady	TAM	25	41	253	1	0	86.7
8520	QB	Tom Brady	TAM	17	27	166	2	0	104.9
8830	QB	Tom Brady	TAM	33	45	369	4	0	127.0
9249	QB	Tom Brady	TAM	28	40	279	2	0	106.1
9561	QB	Tom Brady	TAM	22	38	209	0	3	40.4
9657	QB	Tom Brady	TAM	28	39	341	3	0	124.0
10297	QB	Tom Brady	TAM	26	48	216	2	2	62.5
10632	QB	Tom Brady	TAM	27	41	345	3	2	96.1

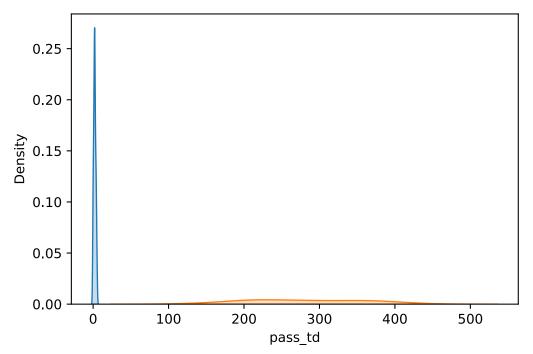
	NFL_quaterbacks								
	position	player	team	pass_cmp	pass_att	pass_yds	pass_td	pass_int	pass_rating
11436	QB	Tom Brady	TAM	15	23	196	2	0	120.9
11545	QB	Tom Brady	TAM	31	45	390	2	0	110.4
11926	QB	Tom Brady	TAM	22	27	348	4	0	158.3
12638	QB	Tom Brady	TAM	26	41	399	4	1	117.8
12699	QB	Tom Brady	TAM	22	40	381	2	0	104.3
12883	QB	Tom Brady	TAM	18	33	199	2	0	92.9
12887	QB	Tom Brady	TAM	20	36	280	3	3	73.8
12953	QB	Tom Brady	TAM	21	29	201	3	0	125.8
12969	QB	Tom Brady	TAM	32	50	379	4	2	97.0
13710	QB	Tom Brady	TAM	24	36	276	5	0	129.2
14088	QB	Tom Brady	TAM	41	55	432	1	0	103.0 r
14390	QB	Tom Brady	TAM	22	43	269	0	0	70.8
14858	QB	Tom Brady	TAM	30	41	411	5	0	144.4
14931	QB	Tom Brady	TAM	34	42	297	2	1	102.1
15559	QB	Tom Brady	TAM	20	36	211	4	0	109.8
15847	QB	Tom Brady	TAM	28	40	375	4	2	112.0
16626	QB	Tom Brady	TAM	23	34	220	2	2	80.5
17027	QB	Tom Brady	TAM	30	46	307	2	1	89.7
17150	QB	Tom Brady	TAM	25	34	226	1	1	88.6
17440	QB	Tom Brady	TAM	38	51	368	4	1	112.2
18034	QB	Tom Brady	TAM	31	46	363	2	0	105.6
18369	QB	Tom Brady	TAM	26	48	214	0	1	57.1

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	pass_cmp	pass_att	pass_yds	pass_td	pass_int	pass_rating
16626	23	34	220	2	2	80.5
17027	30	46	307	2	1	89.7
17150	25	34	226	1	1	88.6
17440	38	51	368	4	1	112.2
18034	31	46	363	2	0	105.6
18369	26	48	214	0	1	57.1

```
In [15]: # histogram of Tom Brady pass touchdowns, Kdeplot for touchdowns
#for col in temp2['pass_cmp']:
    #density unitizes each histogram such that the area under the curver is 1
    #alpha essential allows one to visibly see through each graph
    #kdeplot kernal density plot
#sb.kdeplot(cluster['pass_td'],cluster['pass_yds'],shade=True)
sb.kdeplot(cluster['pass_td'],shade=True)
py.hist(cluster["pass_td"],density = True, alpha = 0.5)
#py.hist(cluster["pass_yds"],density = True, alpha = 0.5)
```





In [17]:

# get the sum of all the rows for each column
print(cluster.sum())

pass\_cmp 1279.0
pass\_att 2000.0
pass\_yds 14308.0
pass\_td 110.0
pass\_int 35.0
pass\_rating 4930.5
dtype: float64

In [18]:

# get a general description of the data cluster from Tom Brady
cluster.describe()

Out[18]:

	pass_cmp	pass_att	pass_yds	pass_td	pass_int	pass_rating
count	51.000000	51.000000	51.000000	51.000000	51.000000	51.000000
mean	25.078431	39.215686	280.549020	2.156863	0.686275	96.676471
std	5.747497	6.862401	76.536348	1.405312	0.836426	24.598484
min	15.000000	23.000000	128.000000	0.000000	0.000000	40.400000
25%	20.500000	36.000000	216.000000	1.000000	0.000000	80.450000
50%	25.000000	40.000000	276.000000	2.000000	0.000000	97.000000
75%	29.000000	45.000000	346.500000	3.000000	1.000000	112.100000
max	41.000000	55.000000	432.000000	5.000000	3.000000	158.300000

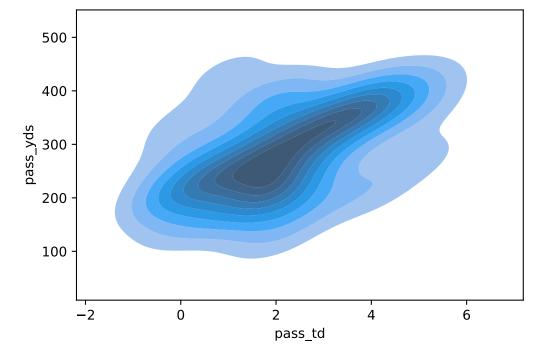
In [19]:

# the cluster shows the corrolation between the passing td and passing yds sb.kdeplot(cluster[1[0]],cluster[1[1]],shade=True)

/Library/Frameworks/Python.framework/Versions/3.9/lib/python3.9/site-packages/seaborn/\_decorators.py:36: FutureWarning: Pass the following variable as a key word arg: y. From version 0.12, the only valid positional argument will be `da ta`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

warnings.warn(

Out[19]: <AxesSubplot:xlabel='pass\_td', ylabel='pass\_yds'>



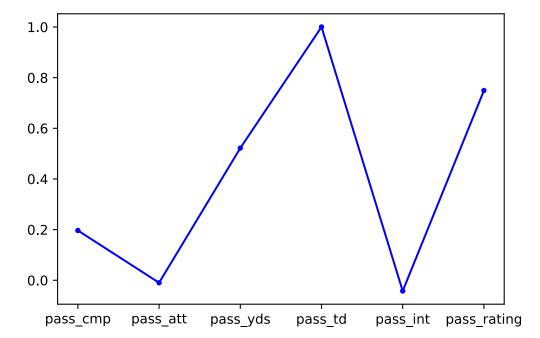
In [20]:
# We will determine the corrolation between each column for Tom Brady to deter
analyze = cluster.corr()

In [21]: # after multiplying the cluster by the Transpose of the cluster we have a symmanalyze

Out[21]:		pass_cmp	pass_att	pass_yds	pass_td	pass_int	pass_rating
	pass_cmp	1.000000	0.797197	0.741627	0.196539	0.063465	0.237020
	pass_att	0.797197	1.000000	0.537903	-0.009800	0.182760	-0.206267
	pass_yds	0.741627	0.537903	1.000000	0.521881	-0.036308	0.566525
	pass_td	0.196539	-0.009800	0.521881	1.000000	-0.042370	0.748998
	pass_int	0.063465	0.182760	-0.036308	-0.042370	1.000000	-0.509630
	pass rating	0.237020	-0.206267	0.566525	0.748998	-0.509630	1.000000

In [23]: # we find that the graph is consistant with the corralations, so we can surmis
#zone = brady.loc[:,'date']
 py.plot(analyze["pass\_td"],marker="." , color = "blue")

Out[23]: [<matplotlib.lines.Line2D at 0x12f3854c0>]



```
In [24]:
          # simple group by
          inser = ['player', 'pass_cmp',
                  'pass_att', 'pass_yds', 'pass_td', 'pass_int', 'pass_rating']
          test = brady.groupby(by=inser[0])[[inser[1],inser[2],inser[3],inser[4],inser[5]
In [25]:
          #cumulative stats for Tom Brady
          test.head()
                    pass_cmp pass_att pass_yds pass_td pass_int pass_rating
Out[25]:
             player
          Tom Brady
                        1279
                                2000
                                        14308
                                                   110
                                                           35
                                                                   4930.5
 In [ ]:
```