

Notebook

February 2, 2019

Local date & time is : 02/02/2019 11:24:50 PST

```
In [4]: def tvd(b, g):  
        return sum(abs(b - g)) / 2
```

```
In [6]: tvd(b, g)
```

```
Out[6]: 0.15
```


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```
In [7]: def prob_matches(k, n):
        x_even = np.arange(0, n - k + 1, 2)
        x_odd = np.arange(1, n - k + 1, 2)
        eventerms = 1 / (special.factorial(x_even))
        oddterms = 1 / (special.factorial(x_odd))
        return (1 / special.factorial(k)) * (sum(eventerms) - sum(oddterms))
```

```
In [8]: prob_matches(99, 100)
```

```
Out[8]: 0.0
```

$$P(M_n = k) = \frac{1}{k!} \cdot \left(1 - \frac{1}{1!} + \frac{1}{2!} - \frac{1}{3!} + \cdots (-1)^{n-k} \frac{1}{(n-k)!}\right)$$

Looking at the formula, for $M_n = 0, 1, \frac{1}{k!}$ both equal to 1. And the latter terms of the equation for $M_n = 0$ differ only by $(-1)^n \frac{1}{(n)!}$ from the equation for $M_n = 1$. For a large n , such term is very small. Hence $P(M_n = 0)$ is very close to $P(M_n = 1)$ when n is large.

```
In [9]: prob_matches(0, 100), prob_matches(1, 100)
```

```
Out[9]: (0.36787944117144233, 0.36787944117144233)
```

```
In [10]: def match_dist(n):
        k = np.arange(n + 1)
        prob = make_array()
        for i in k:
            prob = np.append(prob, prob_matches(i, n))
        return prob
```

```
In [11]: sum(match_dist(10))
```

```
Out[11]: 1.0000000000000002
```

```
In [12]: sum(match_dist(100)[0:11])
```

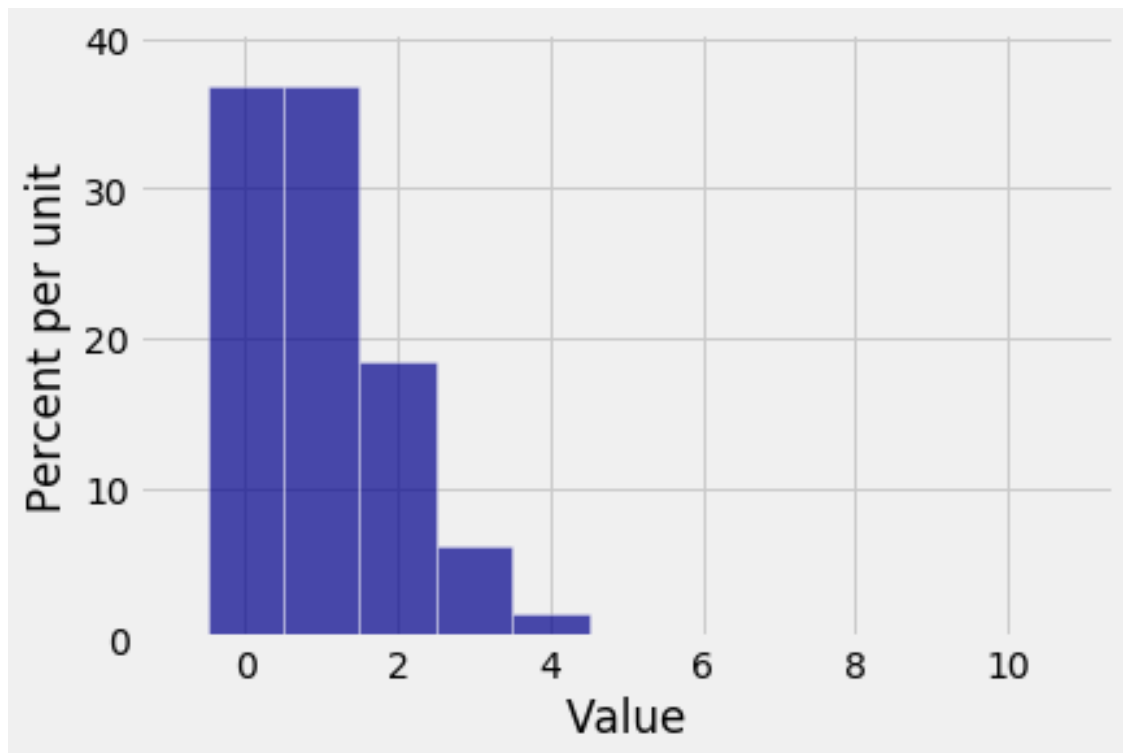
```
Out[12]: 0.9999999899522336
```

The possible values of M_n is from 0 to 100. With the sum of probabilities for $M_n = 0$ to 10 equal to 1, it says that the distribution of M_n lies in between 0 and 10.

```
In [13]: k = np.arange(11)
```

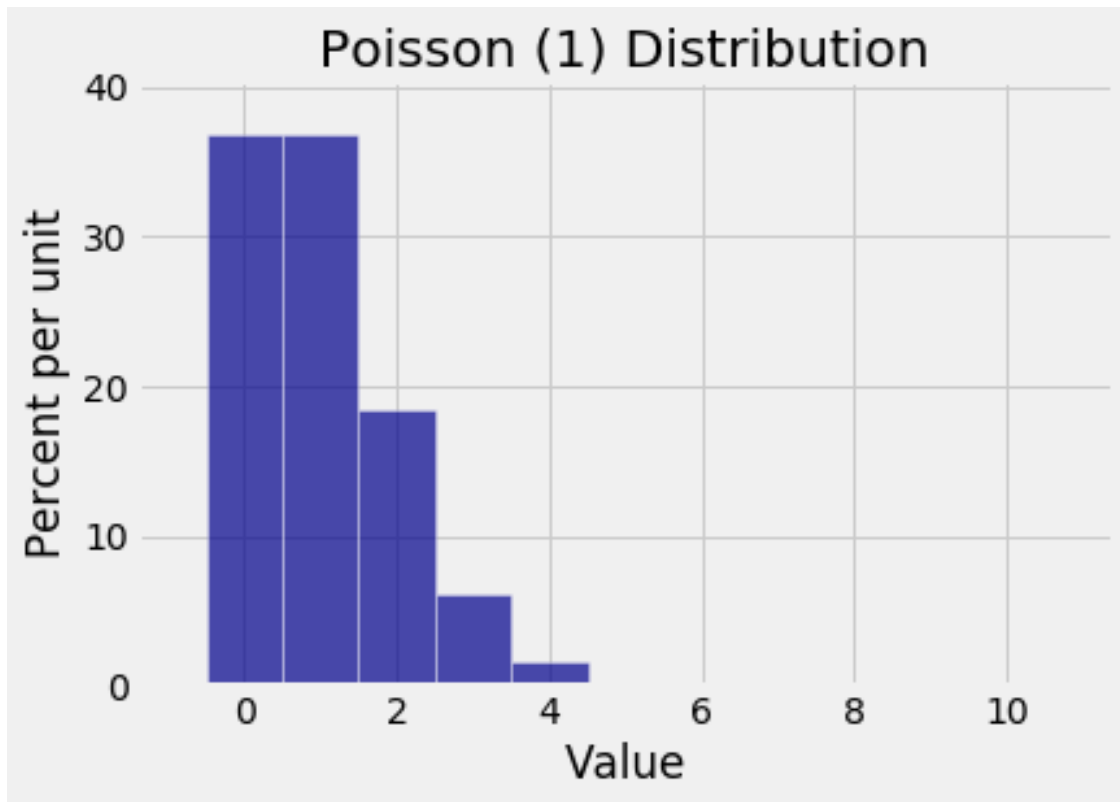
```
prob = match_dist(10)
```

```
matches_100 = Table().values(k).probabilities(prob)
Plot(matches_100)
```



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```
In [14]: k = np.arange(11)           # selected possible values
         poisson_1_probs = stats.poisson.pmf(k, 1)  # array of corresponding Poisson (1) probabilities
         poisson_1_dist = Table().values(k).probabilities(poisson_1_probs)
         Plot(poisson_1_dist)
         plt.title('Poisson (1) Distribution');
```



```
In [15]: def matches_Poisson_tvd(n):
         mn = match_dist(n)

         k = np.arange(n+1)
         poisson = stats.poisson.pmf(k, 1)

         return tvd(mn, poisson)
```

I think `matches_Poisson_tvd(5)` should be larger. The poisson approximation uses e^{-1} is a better approximation for the actual match distribution with higher number of n . Hence the distance away from the actual values will be further away for $n = 5$ compared to $n = 100$

```
In [17]: matches_Poisson_tvd(5), matches_Poisson_tvd(100)
```

```
Out[17]: (0.034111123088143108, 2.0236745963801594e-17)
```

```
In [18]: #matches_Poisson_tvd calculates half of the sum of all the absolute difference
         #between actual distribution and Poisson approximation,
```

```
#Hence, assuming all the differences are positive,  
#the largest possible error is two times the tvd.
```

```
matches_Poisson_tvd(100) * 2
```

```
Out[18]: 4.047349192760319e-17
```

```
In [19]: tvd_table = Table().with_column('n', np.arange(5, 101))
```

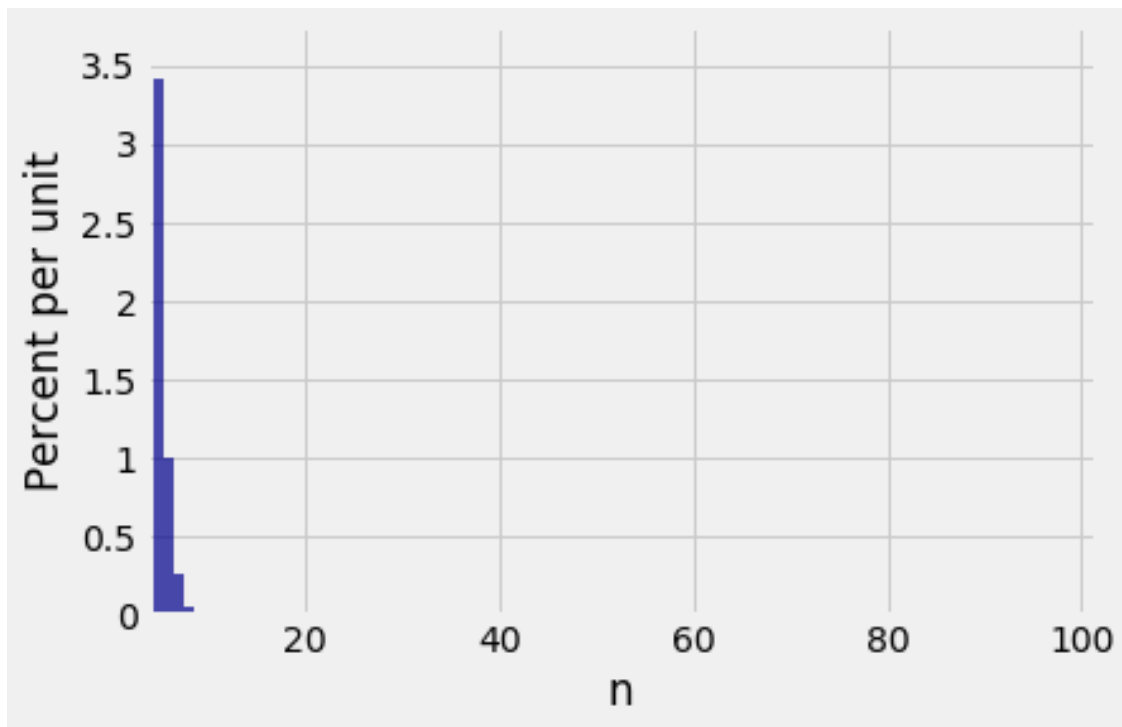
```
matches_tvds = tvd_table.with_column('Matches (n)', tvd_table.apply(matches_Poisson_tvd, 0))
```

```
tvd_table = matches_tvds
```

```
tvd_table
```

```
Out[19]: n      | Matches (n)  
5       | 0.0341112  
6       | 0.0100777  
... Omitting 5 lines ...  
13      | 8.28377e-08  
14      | 1.11286e-08  
... (86 rows omitted)
```

```
In [20]: Plot(tvd_table)
```



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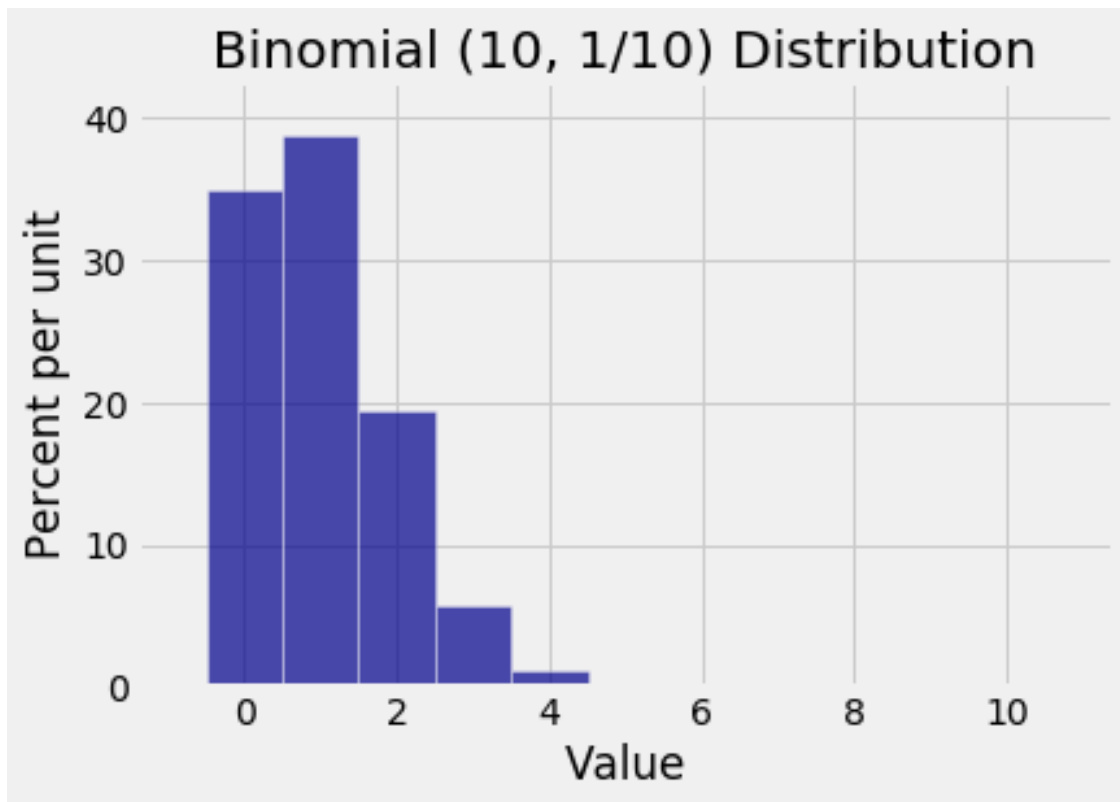
In [21]: *# binomial (n, 1/n) distribution*

```
n = 10                                # number of trials

k = np.arange(11)                     # possible values
binom_probs = stats.binom.pmf(k, n, 1/n) # binomial (n, 1/n) probabilities

binom_dist = Table().values(k).probabilities(binom_probs)

Plot(binom_dist)
plt.title('Binomial (10, 1/10) Distribution');
```



```
In [23]: def binomial_Poisson_tvd(n):
          k = np.arange(n+1)
          binom = stats.binom.pmf(k, n, 1 / n)
          poisson = stats.poisson.pmf(k, 1)
          return tvd(binom, poisson)
```

```
In [24]: binomial_Poisson_tvd(30)
```

```
Out[24]: 0.009379738441886479
```

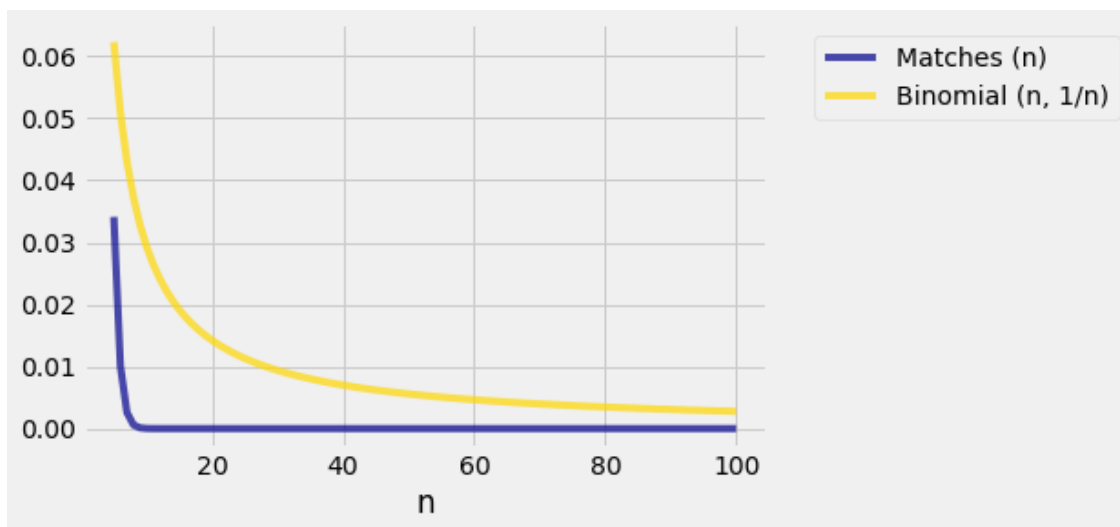
```
In [25]: binom_tvds = tvd_table.apply(binomial_Poisson_tvd, 0)

tvd_table = tvd_table.with_column('Binomial (n, 1/n)', binom_tvds)

tvd_table
```

```
Out[25]: n      | Matches (n) | Binomial (n, 1/n)
5       | 0.0341112   | 0.0622837
6       | 0.0100777   | 0.0509556
... 0mitting 5 lines ...
13      | 8.28377e-08 | 0.0222259
14      | 1.11286e-08 | 0.0205686
... (86 rows omitted)
```

```
In [26]: tvd_table.plot(0)
```



For values of n that are about 60 or more, Poisson(1) approximations to binomial($n, 1/n$) probabilities will be off by at most 0.5%.

```
In [27]: Table().with_columns('n', tvd_table[0], 'diff', abs(tvd_table[1] - tvd_table[2])).show()

<IPython.core.display.HTML object>
```

```
In [ ]:
```