

Lab 7: Probability Distributions

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[48]: import numpy as np
      %matplotlib inline
      import matplotlib.pyplot as plt
```

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[56]: #7.2 Simulating the Binomial Process

nexp = 1000
ntry = 10
eps = .25

# nexp = int(input("Enter number of experiments: "))
# ntry = int(input("Enter number of trials: "))
# eps = float(input("Enter a value epsilon () < 1: "))

def th_bi(nexp, ntry, eps):
    x = np.random.uniform(size=(nexp, ntry))
    xresult = np.array(x<eps, dtype = int)
    s = np.sum(xresult, axis = 1)
    return s

v = th_bi(nexp, ntry, eps)

print("mean expected: ", ntry*eps)
print("mean simulated: ", np.mean(v))
print("var expected: ", ntry*eps*(1.0 - eps))
print("var simulated: ", np.var(v))
```

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mean expected:  2.5
mean simulated: 2.442
var expected:  1.875
var simulated:  1.890636
```

```
[57]: #7.3 Histogram for the Binomial Process

counts, edges = np.histogram(v, bins = 11, range = (0,11))
print("Counts: ", counts)
print("Total:  ", np.sum(counts))
```

```

print("Bin Edges: ", edges)

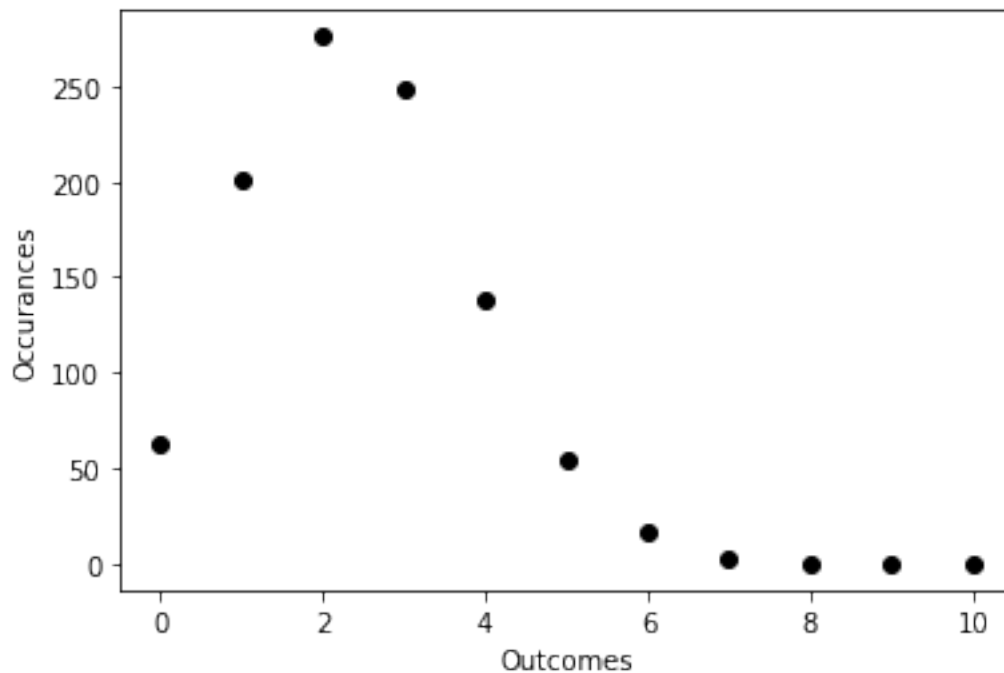
plt.plot(edges[:-1], counts, "ko")
plt.xlabel("Outcomes")
plt.ylabel("Occurances")
print("edges[:-1]: ", edges[:-1])

```

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Counts: [ 63 201 276 248 138  54  17   3   0   0   0]
Total:  1000
Bin Edges: [ 0.  1.  2.  3.  4.  5.  6.  7.  8.  9. 10. 11.]
edges[:-1]: [ 0.  1.  2.  3.  4.  5.  6.  7.  8.  9. 10.]

```



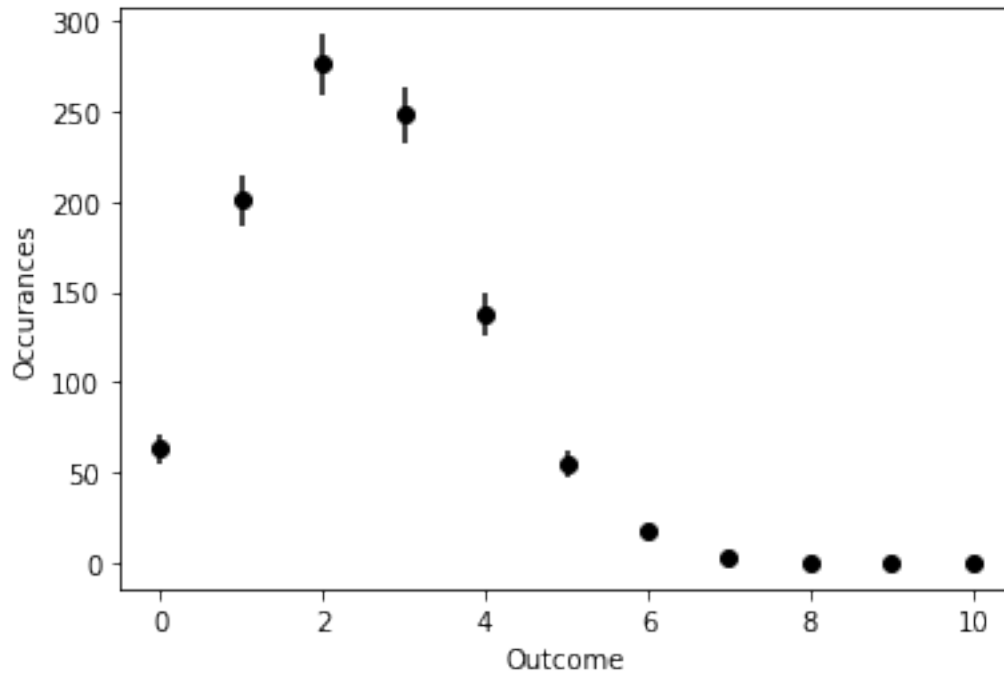
[58]: #7.4 Error Bars

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errs = counts*0.5
plt.errorbar(edges[:-1], counts, yerr = errs, fmt = "ko")
plt.xlabel("Outcome")
plt.ylabel("Occurances")

```

[58]: Text(0, 0.5, 'Occurances')



```
[59]: #7.5 Comparison with Binomial PDF

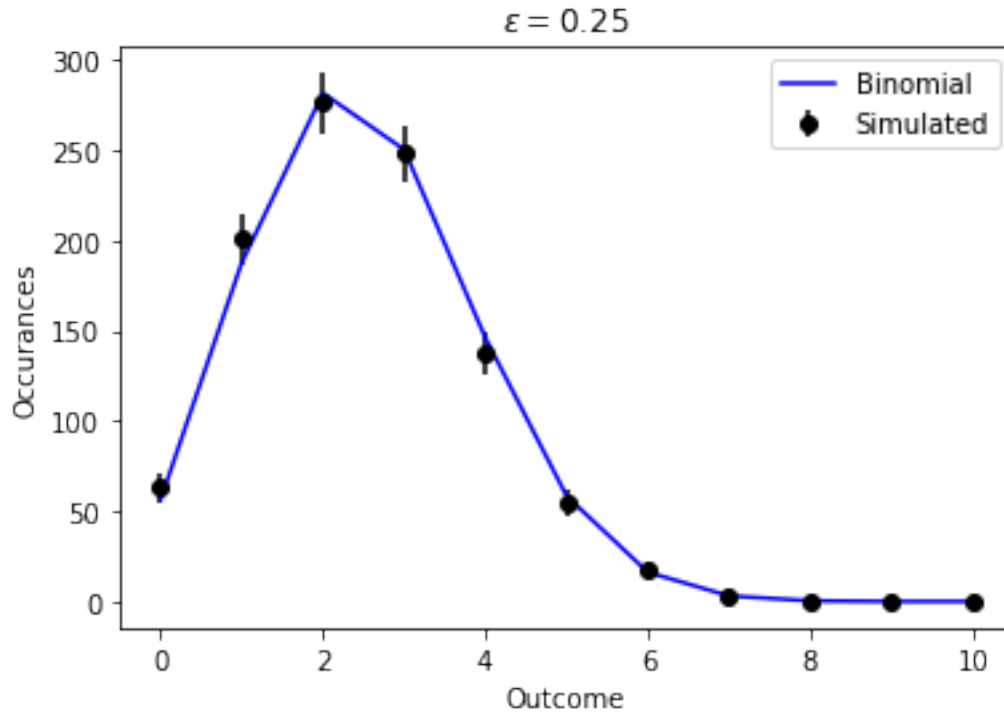
from scipy.stats import binom

# MCBBD = input("Compare Monte Carlo with The Binomial Distribution? 'yes' or 'no': ")
# yes
# if MCBBD == string

errs = counts**0.5
plt.errorbar(edges[:-1], counts, yerr = errs, fmt = "ko", label = "Simulated")

plt.xlabel("Outcome")
plt.ylabel("Occurrences")
xpred = edges[:-1]
ypred = nexp * binom.pmf(xpred, ntry, eps)
plt.plot(xpred, ypred, "b-", label = "Binomial")
plt.legend()
plt.title("$\\epsilon = 0.25$")
```

```
[59]: Text(0.5, 1.0, '$\\epsilon = 0.25$')
```



```
[60]: #7.6 The Poisson Limit
      #note

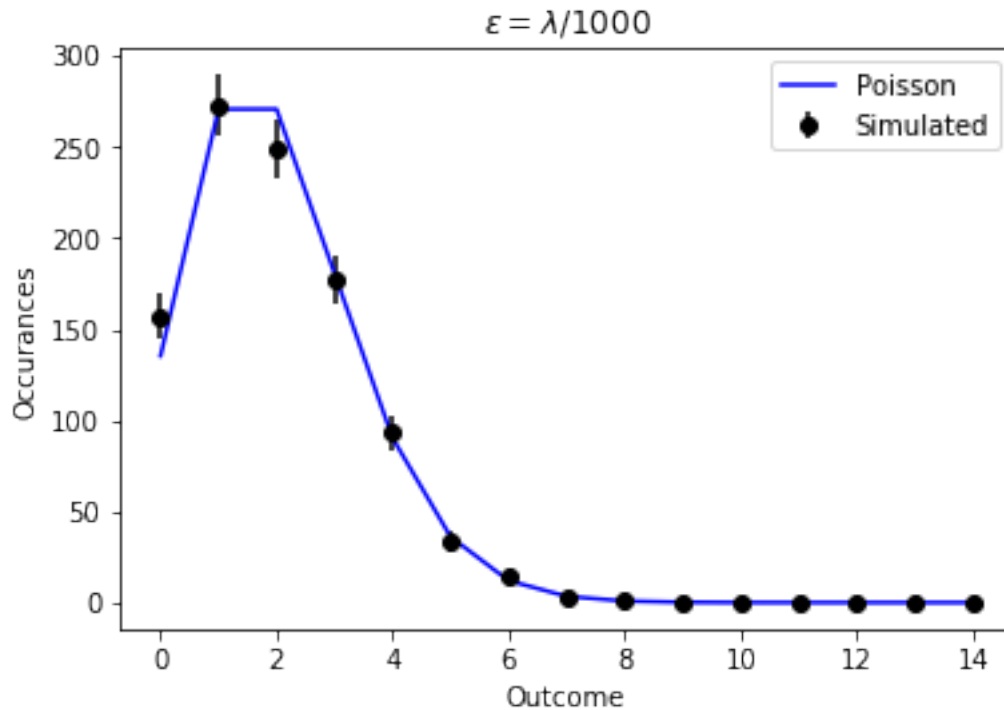
      from scipy.stats import poisson

      nexp1 = 1000
      ntry1 = 1000
      lamb1 = 2
      eps1 = lamb1/ntry1
      v1 = th_bi(nexp1, ntry1, eps1)
      counts, edges = np.histogram(v1, bins = 15, range = (0,15))

      errs = counts**0.5
      plt.errorbar(edges[:-1], counts, yerr = errs, fmt = "ko", label = "Simulated")

      plt.xlabel("Outcome")
      plt.ylabel("Occurrences")
      xpred = edges[:-1]
      ypred = nexp * poisson.pmf(xpred, lamb1)
      plt.plot(xpred, ypred, "b-", label = "Poisson")
      plt.legend()
      plt.title("$\epsilon$ = /1000$")
```

```
[60]: Text(0.5, 1.0, '$\\epsilon = /1000$')
```



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[63]: #7.7 The Gaussian Limit

from scipy.stats import norm

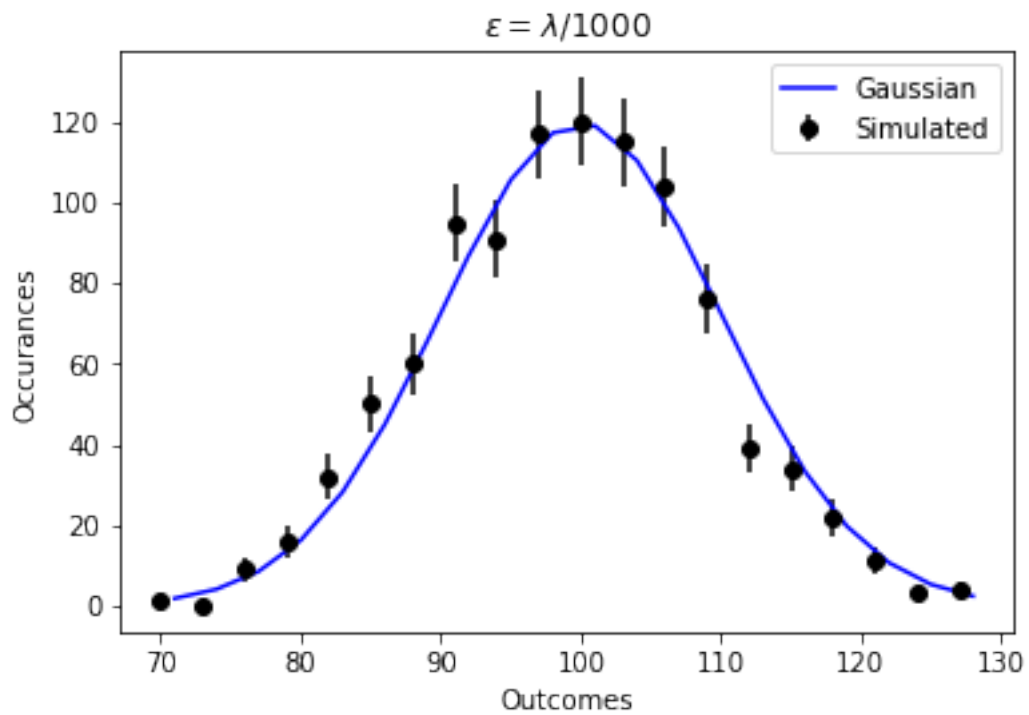
lamb2 = 100
eps2 = lamb2/ntry1
v2 = th_bi(nexp1, ntry1, eps2)
counts, edges = np.histogram(v2, bins = 20, range = (70,130))
w = edges[1] - edges[0]

errs = counts**0.5
plt.errorbar(edges[:-1], counts, yerr = errs, fmt = "ko", label = "Simulated")
print(counts)
cbins = (edges[:-1] + edges[1:] - 1)/2
plt.xlabel("Outcomes")
plt.ylabel("Occurances")
xpred = cbins
ypred = nexp * w * norm.pdf(xpred, loc = lamb2, scale = lamb2**0.5)
plt.plot(xpred, ypred, "b-", label = "Gaussian")
plt.legend()
```

```
plt.title("$\epsilon = /1000$")
```

```
[ 1  0  9 16 32 50 60 95 91 117 120 115 104 76 39 34 22 11
 3  4]
```

```
[63]: Text(0.5, 1.0, '$\epsilon = /1000$')
```



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[ ]:
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