## **Stochastic Optimization with CPLEX**

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#### Table of content

- 1. New-vendor Problem
  - · problem description
  - · case 1: the original probelm
  - · case 2: maximizing the wrost case
  - case 3: more than 80% the profit is more than 4000 (chance contraints)

```
In [1]: import numpy as np
   import matplotlib.pyplot as plt
   import random
   random.seed(a=600)
```

#### **New-vendor Problem**

#### original problem:

$$\max_{x} E[f(x)]$$

$$st:$$

$$f(x) = c_{s}\hat{s} + c_{d}\hat{s}' - c_{p}x$$

$$0 \le \hat{s} \le \min\{x, \hat{d}\}$$

$$0 \le \hat{s}' \le \max\{0, x - \hat{d}\}$$

where f(x) is the profit of purchasing x units;  $c_s$  is the unit selling price;  $c_d$  is the unit discounted selling price;  $c_p$  is the unit purchasing cost;  $\hat{s}$  is sales;  $\hat{s}'$  is discounted sales;

#### issue:

$$E[f(x)] = \int_0^\infty x f(x) dx \implies$$
 hard to solve

#### relaxation:

Assume the uncertainty set U has a finite number of elements. And the problem can be reformulated into a big LP problem:

$$\begin{aligned} \max_{x} & \sum_{u \in U} p_{u} \left[ c_{s} s_{u} + c_{d} s'_{u} - c_{p} x \right] \\ st: \\ & 0 \leq s_{u} \leq \min\{x, d_{u}\}, \forall u \in U \\ & 0 \leq s'_{u} \leq \max\{0, x - d_{u}\}, \forall u \in U \end{aligned}$$

reformulate the problem:

$$\begin{aligned} \max_{x} \; & \sum_{u \in U} p_{u} \; [c_{s}s_{u} + c_{d}s'_{u} - c_{p}x] \\ st: \\ & s_{u} + s'_{u} = x \;, \forall u \in U \\ & 0 \leq s_{u} \leq \alpha_{u} \;, \forall u \in U \\ & \alpha_{u} \leq x \;, \forall u \in U \\ & \alpha_{u} \leq d_{u} \;, \forall u \in U \\ & 0 \leq s'_{u} \leq \beta_{u} \;, \forall u \in U \\ & \beta_{u} \geq 0 \;, \forall u \in U \\ & \beta_{u} \geq x - d_{u} \;, \forall u \in U \end{aligned}$$

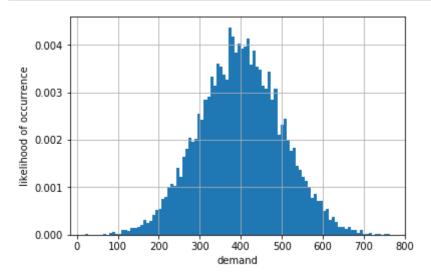
### **Demands**

generating random demands from a normal distribution

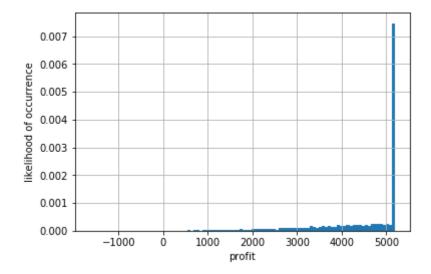
```
In [2]: sigma = 100
    mu = 400
    samples = 10000
    demand = [max(random.normalvariate(mu,sigma),0) for i in range(samples)]

Cs=15
Cd=-3
Cp=2

plt.hist(demand,bins=100,density=True)
plt.grid(True)
plt.ylabel("likelihood of occurrence")
plt.xlabel("demand")
plt.show()
```



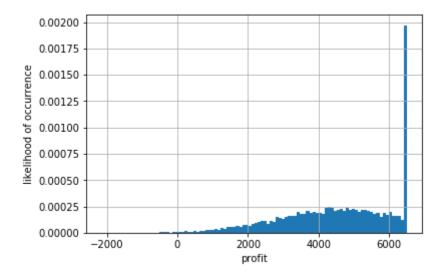
# what if $x = \mu$



order quantity:400.00 | mean profit:4472.32 | std:1053.03 | worst case:-1641.30 | 25.77 % of cases profit is less than 4000

## what if $\mathbf{x} = \boldsymbol{\mu} + \boldsymbol{\sigma}$

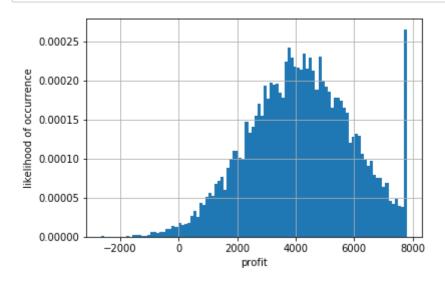
```
In [4]: x=mu+sigma
    profit=[Cs*min(x,demand[u])+Cd*max(0,x-demand[u])-Cp*x for u in range(sample plt.hist(profit,bins=100,density=True)
    plt.grid(True)
    plt.ylabel("likelihood of occurrence")
    plt.xlabel("profit")
    plt.show()
    print("order quantity:{0:.2f}|mean profit:{1:.2f}|std:{2:.2f}| worst case:{
        profit_array=np.array(profit)
        print(100*len(profit_array[profit_array<=4000])/len(profit_array),"% of case</pre>
```



order quantity:500.00 | mean profit:4535.67 | std:1562.59 | worst case:-2141.30 | 35.25 % of cases profit is less than 4000

## what if $x=\mu + 2\sigma$

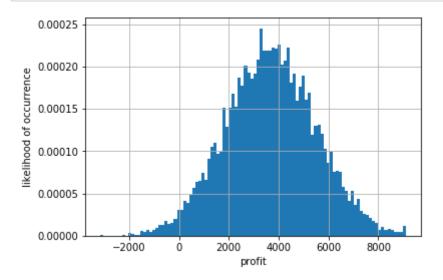
```
In [5]: x=mu+2*sigma
    profit=[Cs*min(x,demand[u])+Cd*max(0,x-demand[u])-Cp*x for u in range(sample plt.hist(profit,bins=100,density=True)
    plt.grid(True)
    plt.ylabel("likelihood of occurrence")
    plt.xlabel("profit")
    plt.show()
    print("order quantity:{0:.2f}|mean profit:{1:.2f}|std:{2:.2f}| worst case:{
        profit_array=np.array(profit)
        print(100*len(profit_array[profit_array<=4000])/len(profit_array),"% of case</pre>
```



order quantity:600.00 | mean profit:4172.72 | std:1771.04 | worst case:-2641.30 46.0 % of cases profit is less than 4000

# what if $x=\mu + 3\sigma$

```
In [6]: x=mu+3*sigma
    profit=[Cs*min(x,demand[u])+Cd*max(0,x-demand[u])-Cp*x for u in range(sample
    plt.hist(profit,bins=100,density=True)
    plt.grid(True)
    plt.ylabel("likelihood of occurrence")
    plt.xlabel("profit")
    plt.show()
    print("order quantity:{0:.2f}|mean profit:{1:.2f}|std:{2:.2f}| worst case:{
        profit_array=np.array(profit)
        print(100*len(profit_array[profit_array<=4000])/len(profit_array),"% of case</pre>
```



order quantity:700.00 | mean profit:3687.02 | std:1804.15 | worst case:-3141.30 | 57.09 % of cases profit is less than 4000

## Case I: the original problem

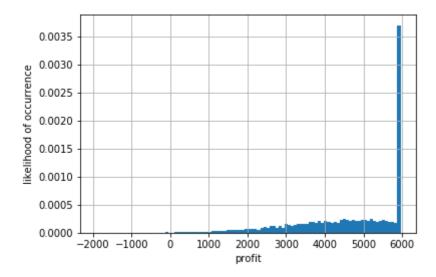
$$\begin{aligned} \max_{x} \; \sum_{u \in U} p_{u} \; [c_{s}s_{u} + c_{d}s'_{u} - c_{p}x] \\ st: \\ s_{u} + s'_{u} = x \;, \forall u \in U \\ 0 \leq s_{u} \leq \alpha_{u} \;, \forall u \in U \\ \alpha_{u} \leq x \;, \forall u \in U \\ \alpha_{u} \leq d_{u} \;, \forall u \in U \\ 0 \leq s'_{u} \leq \beta_{u} \;, \forall u \in U \\ \beta_{u} \geq 0 \;, \forall u \in U \\ \beta_{u} \geq x - d_{u} \;, \forall u \in U \end{aligned}$$

In [7]: from docplex.mp.model import Model

```
In [8]:
                              md1 = Model(name='news-vendor case 1')
                               x = md1.continuous_var(name="x",lb=0)
                               s = \{(u): md1.continuous\_var(name="s_{0}".format(u), lb=0) for u in range(sample of the continuous\_var(name="s_{0}".format(u), lb=0) for u in range(sample of the continuous\_var(name="s_{0}".form
                               sp = {(u):md1.continuous_var(name="sp {0}".format(u),lb=0) for u in range(se
                               alpha = {(u):md1.continuous_var(name="alpha_{0}".format(u),lb=0) for u in ra
                               beta = {(u):mdl.continuous_var(name="beta_{0}".format(u),lb=0) for u in rand
                               profit=(1.0/samples)*md1.sum(Cs*s[u]+Cd*sp[u]-Cp*x for u in range(samples))
                               md1.maximize(profit)
                                for u in range(samples):
                                              md1.add_constraint(s[u]+sp[u]==x)
                                              md1.add_constraint(s[u]<=alpha[u])</pre>
                                              md1.add_constraint(alpha[u]<=x)</pre>
                                              md1.add_constraint(alpha[u]<=demand[u])</pre>
                                              md1.add_constraint(sp[u]<=beta[u])</pre>
                                              md1.add_constraint(beta[u]>=x-demand[u])
                               md1.print_information()
```

```
Model: news-vendor case 1
- number of variables: 40001
- binary=0, integer=0, continuous=40001
- number of constraints: 60000
- linear=60000
- parameters: defaults
```

```
In [9]: md1.solve()
    x=md1.solution.as_dict()['x']
    profit=[Cs*min(x,demand[u])+Cd*max(0,x-demand[u])-Cp*x for u in range(sample plt.hist(profit,bins=100,density=True)
    plt.grid(True)
    plt.ylabel("likelihood of occurrence")
    plt.xlabel("profit")
    plt.show()
    print("order quantity:{0:.2f}|mean profit:{1:.2f}|std:{2:.2f}| worst case:
    profit_array=np.array(profit)
    print(100*len(profit_array[profit_array<=4000])/len(profit_array),"% of case</pre>
```



order quantity:458.49 mean profit:4582.56 std:1384.24 worst case:-1933. 75 31.25 % of cases profit is less than 4000

solution from the optimization model has a high mean and lower std.

### Case II: maximize the worst case

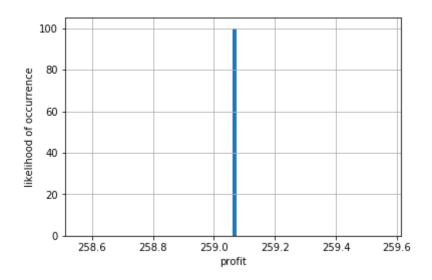
```
\begin{aligned} \max_{x} w \\ st: \\ w &\leq c_{s}s_{u} + c_{d}s'_{u} - c_{p}x \,, \forall u \in U \\ s_{u} + s'_{u} &= x \,, \forall u \in U \\ 0 &\leq s_{u} \leq \alpha_{u} \,, \forall u \in U \\ \alpha_{u} &\leq x \,, \forall u \in U \\ \alpha_{u} &\leq d_{u} \,, \forall u \in U \\ 0 &\leq s'_{u} \leq \beta_{u} \,, \forall u \in U \\ \beta_{u} &\geq 0 \,, \forall u \in U \\ \beta_{u} &\geq x - d_{u} \,, \forall u \in U \end{aligned}
```

```
In [10]: | md2 = Model(name='news-vendor case 2')
                                  w = md2.continuous_var(name="w",lb=-md2.infinity)
                                  x = md2.continuous_var(name="x",1b=0)
                                  s = {(u):md2.continuous_var(name="s_{0}".format(u),lb=0) for u in range(same
                                  sp = \{(u): md2.continuous\_var(name="sp_{0}".format(u), lb=0) for u in range(same="sp_{0}".format(u), lb=0) for u
                                  alpha = {(u):md2.continuous_var(name="alpha_{0}".format(u),lb=0) for u in ra
                                  beta = {(u):md2.continuous_var(name="beta_{0}".format(u),lb=0) for u in range
                                  md2.maximize(w)
                                   for u in range(samples):
                                                 md2.add_constraint(w<=Cs*s[u]+Cd*sp[u]-Cp*x)</pre>
                                                 md2.add_constraint(s[u]+sp[u]==x)
                                                 md2.add_constraint(s[u]<=alpha[u])</pre>
                                                 md2.add_constraint(alpha[u]<=x)</pre>
                                                 md2.add_constraint(alpha[u]<=demand[u])</pre>
                                                 md2.add_constraint(sp[u]<=beta[u])</pre>
                                                 md2.add_constraint(beta[u]>=x-demand[u])
                                  md2.print_information()
```

```
Model: news-vendor case 2
- number of variables: 40002
- binary=0, integer=0, continuous=40002
- number of constraints: 70000
- linear=70000
- parameters: defaults
```

```
In [11]: md2.solve()
    md2.report()
    x=md2.solution.as_dict()['x']
    profit=[Cs*min(x,demand[u])+Cd*max(0,x-demand[u])-Cp*x for u in range(sample plt.hist(profit,bins=100,density=True)
    plt.grid(True)
    plt.ylabel("likelihood of occurrence")
    plt.xlabel("profit")
    plt.show()
    print("order quantity:{0:.2f}|mean profit:{1:.2f}|std:{2:.2f}| worst case:{profit_array=np.array(profit)}
    print(100*len(profit_array[profit_array<=4000])/len(profit_array),"% of case</pre>
```

\* model news-vendor case 2 solved with objective = 259.063



order quantity:19.93 mean profit:259.06 std:0.00 worst case:259.06 loo.0 % of cases profit is less than 4000

### Case III: chance constraints

### 80% of time has a profit more than 4000

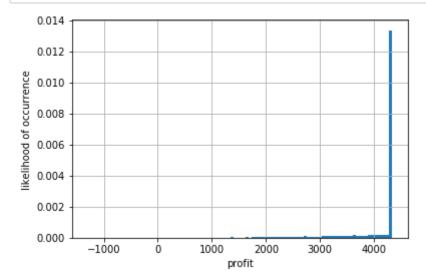
$$\begin{aligned} \max_{x} \ \sum_{u \in U} p_{u} \ [c_{s}s_{u} + c_{d}s'_{u} - c_{p}x] \\ st: \\ \sum_{u \in U} \theta_{u} \geq 0.80 * |U|, \forall u \in U \\ c_{s}s_{u} + c_{d}s'_{u} - c_{p}x \geq 4000 * \theta_{u} - M * (1 - \theta_{u}), \forall u \in U \\ s_{u} + s'_{u} = x, \forall u \in U \\ 0 \leq s_{u} \leq \alpha_{u}, \forall u \in U \\ \alpha_{u} \leq x, \forall u \in U \\ \alpha_{u} \leq d_{u}, \forall u \in U \\ 0 \leq s'_{u} \leq \beta_{u}, \forall u \in U \\ \beta_{u} \geq 0, \forall u \in U \\ \beta_{u} \geq x - d_{u}, \forall u \in U \end{aligned}$$

```
In [12]: md3 = Model(name='news-vendor chance constraint')
                          x = md3.continuous var(name="x", lb=0)
                          s = {(u):md3.continuous_var(name="s_{0}".format(u),lb=0) for u in range(same
                          sp = {(u):md3.continuous_var(name="sp {0}".format(u),lb=0) for u in range(se
                           alpha = {(u):md3.continuous_var(name="alpha_{0}".format(u),lb=0) for u in re
                          beta = {(u):md3.continuous_var(name="beta_{0}".format(u),lb=0) for u in rand
                          theta= {(u):md3.binary var(name="theta {0}".format(u)) for u in range(sample
                          profit=(1.0/samples)*md3.sum(Cs*s[u]+Cd*sp[u]-Cp*x for u in range(samples))
                          md3.maximize(profit)
                          md3.add constraint(md3.sum(theta[u] for u in range(samples))>=0.80*samples)
                           for u in range(samples):
                                     md3.add\_constraint(Cs*s[u]+Cd*sp[u]-Cp*x>=4000*theta[u]-(9999)*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(1-theta[u]-(9999))*(
                                      md3.add_constraint(s[u]+sp[u]==x)
                                      md3.add_constraint(s[u]<=alpha[u])</pre>
                                      md3.add constraint(alpha[u]<=x)</pre>
                                      md3.add_constraint(alpha[u]<=demand[u])</pre>
                                      md3.add_constraint(sp[u]<=beta[u])</pre>
                                      md3.add_constraint(beta[u]>=x-demand[u])
                          md3.print_information()
                          Model: news-vendor chance constraint
                             - number of variables: 50001
                                  - binary=10000, integer=0, continuous=40001
```

- number of constraints: 70001
  - linear=70001
- parameters: defaults

```
In [13]: md3.solve()
         md3.report()
```

\* model news-vendor chance constraint solved with objective = 4058.269



order quantity:333.51|mean profit:4058.27|std:646.31 | worst case:-1308.8 4 20.01 % of cases profit is less than 4000

In [ ]: