



Dealing with Uncertainty

in Decision Making Models

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I. A product mix problem

A formulation

A furniture manufacturer must choose $x_j \geq 0$, how many dressers of type $j = 1, \dots, 4$ to manufacture so as to maximize profit

$$\sum_{j=1}^4 c_j x_j = 12x_1 + 25x_2 + 21x_3 + 40x_4$$

The constraints:

$$t_{c1}x_1 + t_{c2}x_2 + t_{c3}x_3 + t_{c4}x_4 \leq d_c$$

$$t_{f1}x_1 + t_{f2}x_2 + t_{f3}x_3 + t_{f4}x_4 \leq d_f$$

t_{cj} (t_{fj}) carpentry (finishing) man-hours: dresser type j

d_c (d_f) = total time available for carpentry (finishing)

Product mix problem (2)

Solution via linear programming:

$$\max \langle c, x \rangle \text{ so that } Tx \leq d, x \in \mathbb{R}_+^n.$$

With

$$T = \begin{bmatrix} t_{c1} & t_{c2} & t_{c3} & t_{c4} \\ t_{f1} & t_{f2} & t_{f3} & t_{f4} \end{bmatrix} = \begin{bmatrix} 4 & 9 & 7 & 10 \\ 1 & 1 & 3 & 40 \end{bmatrix}, \quad \begin{bmatrix} d_c \\ d_f \end{bmatrix} = \begin{bmatrix} 6000 \\ 4000 \end{bmatrix}$$

Optimal: $x^d = (4000/3, 0, 0, 200/3)$

Value: \$ 18,667.

Product mix problem (3)

But . . . “reality” can’t be ignored!

$$t_{cj} = t_{cj} + \eta_{cj}, \quad t_{fj} = t_{fj} + \eta_{fj}$$

entry	possible values			
$d_c + \zeta_c$:	5,873	5,967	6,033	6,127
$d_f + \zeta_f$:	3,936	3,984	4,016	4,064

10 random variables, say, 4 possible values each

$$L = 1,048,576 \text{ possible pairs } (T^l, d^l)$$

Product mix problem (4)

What if $\sum_{j=1}^4 (t_{cj} + \eta_{cj})x_j > d_c + \zeta_c$? \implies overtime

With $\xi = (\eta_{\{.,.\}}, \zeta_{\{.\}})$, **recourse**: $(y_c(\xi), y_f(\xi))$ @ cost (q_c, q_f) .

$$\begin{array}{llllll} \max & \langle c, x \rangle & -p_1 \langle q, y^1 \rangle & -p_2 \langle q, y^2 \rangle & \cdots - & p_L \langle q, y^L \rangle \\ \text{s.t.} & T^1 x & -y^1 & & & \leq d^1 \\ & T^2 x & & -y^2 & & \leq d^2 \\ & \vdots & & & \ddots & \vdots \\ & T^L x & & & -y^L & \leq d^L \\ & x \geq 0, & y^1 \geq 0, & y^2 \geq 0, & \cdots & y^L \geq 0. \end{array}$$

Structured large scale l.p. ($L \approx 10^6$)

Product mix problem (5)

Define $\Xi = \{\xi = (\eta, \zeta)\}$, $p_\xi = \text{prob} [\xi = \xi]$

$$Q(\xi, x) = \max \{ \langle -q, y \rangle \mid T_\xi x - y \leq d_\xi, y \geq 0 \}$$

$$EQ(x) = E\{Q(\xi, x)\} = \sum_{\xi \in \Xi} p_\xi Q(\xi, x)$$

the **equivalent deterministic program (DEP)**:

$$\max \langle c, x \rangle + EQ(x) \text{ so that } x \in \mathbb{R}_+^n.$$

a *non-smooth* convex optimization problem: EQ concave.

Product mix problem (6)

Solution of **DEP**, or large scale l.p.,:

$$\text{Optimal: } x^* = (257, 0, 665.2, 33.8)$$

expected Profit: \$ 18,051

The solution x^* is *robust*: it considered all $\approx 10^6$ possibilities.

Product mix problem (6)

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$$\text{Recall: } x^d = (1333.33, 0, 0, 66.67)$$

expected “profit” relying on $x^d = \$ 16,942$.

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- x^d is not close to optimal
- x^d isn't pointing in the right direction

Mathematics & Numerics

Stochastic Programming relies on:

- linear, non-linear, mixed-integer programming
- large scale: decomposition methods, structured programs, grid computing
- Variational Analysis: non-smooth, duality, epi-convergence (approximations), etc.
- Probability: stochastic processes, asymptotic laws
- Statistics: estimation, lack of data issues
- Functional Analysis, Combinatorial Geometry, etc.



II. Modeling, modeling & modeling!

Uncertain parameters

Deterministic Optimization problem:

$$\min f_0(x) \text{ so that } x \in S \subset \mathbb{R}^n$$

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Wait-and-see solution ??

$$x(\xi) \in \operatorname{argmin} \{ f_0(\xi, x) \mid x \in S(\xi) \}$$

What's needed: a *here-and-now* solution.

The News Vendor Problem

- $\xi \in \Xi \subset \mathbb{R}_+$ demand for a (perishable) good
e.g., plant capacity, overbooking, etc.
- $x \geq 0$ quantity ordered @ unit cost: $c = 10$
- $y \geq 0$ quantity sold, per unit profit $r = 15$

Total revenue (possibly negative):

$$-cx + (c + r)y \text{ where } 0 \leq x,$$

$$0 \leq y \leq \min \{x, \xi\}$$

Find optimal x^* !

The “deterministic” approach

Pick $\hat{\xi} \in \Xi$ (guessing the future) and solve

$$\min f_0(\hat{\xi}, x) \text{ so that } x \in S(\hat{\xi}) \subset \mathbb{R}^n$$

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News Vendor: $\Xi = [0, 150]$, pick $\hat{\xi} = 75$,

$$\max -cx + (c + r)y$$

$$x \geq 0, \quad 0 \leq y \leq \min\{x, \hat{\xi}\}$$

Solution: $x^o = y^o = \hat{\xi}$, obj. value = $r\hat{\xi} = 1125$

But doesn't tell much about “profit” if $\xi \neq 75$!

Scenario Analysis

Pick ξ^1, \dots, ξ^L (scenarios), and for each ξ^l find:

$$x^l \in \operatorname{argmin} \{ f_0(\xi^l, x) \mid x \in S(\xi^l) \}$$

and “reconcile” the solutions to obtain x^o .

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NewsVendor: pick $\xi^1 = 10, \xi^2 = 20, \dots, \xi^{15} = 150$,

$$(x^l, y^l) \in \operatorname{argmax}_{x \geq 0, y \geq 0} \{ -cx + (c + r)y \mid y \leq \min[\xi^l, x] \}$$

Wait-and-see sol’ns: $x^l = \xi^l$. “Reconciliation”?

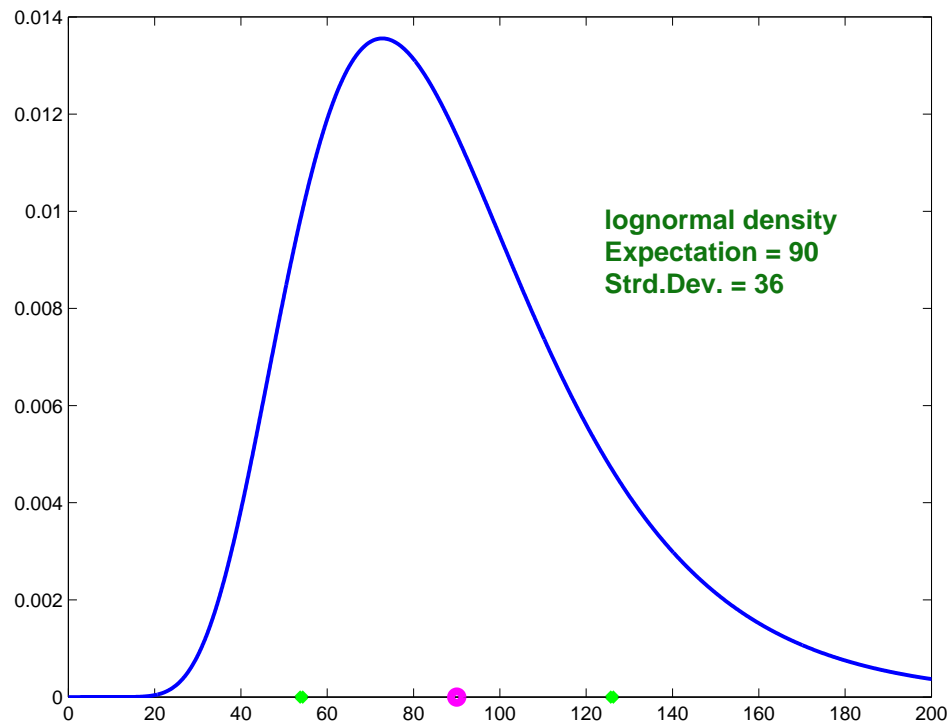
No help in choosing x^o the quantity to order.

ξ : Estimated Density h

ξ log-normal: $h(z) = (z\tau\sqrt{2\pi})^{-1} e^{-\frac{(\ln z - \theta)^2}{2\tau^2}}$

$\theta = 4.43, \tau = 0.38$; $H(z) = \int_0^z h(s) ds$

from data, expert(s), all information available



might affect choice of $\hat{\xi}$, scenarios: ξ^1, \dots

Maximize Expected Return

$$\max -cx + E\{(c+r)y_\xi\}$$

$$\text{so that } x \geq 0, 0 \leq y_\xi \leq \min[\xi, x]$$

The *equivalent deterministic program*:

$$\max_{x \geq 0} -cx + EQ(x), \quad EQ(x) = E\{Q(\xi, x)\}$$

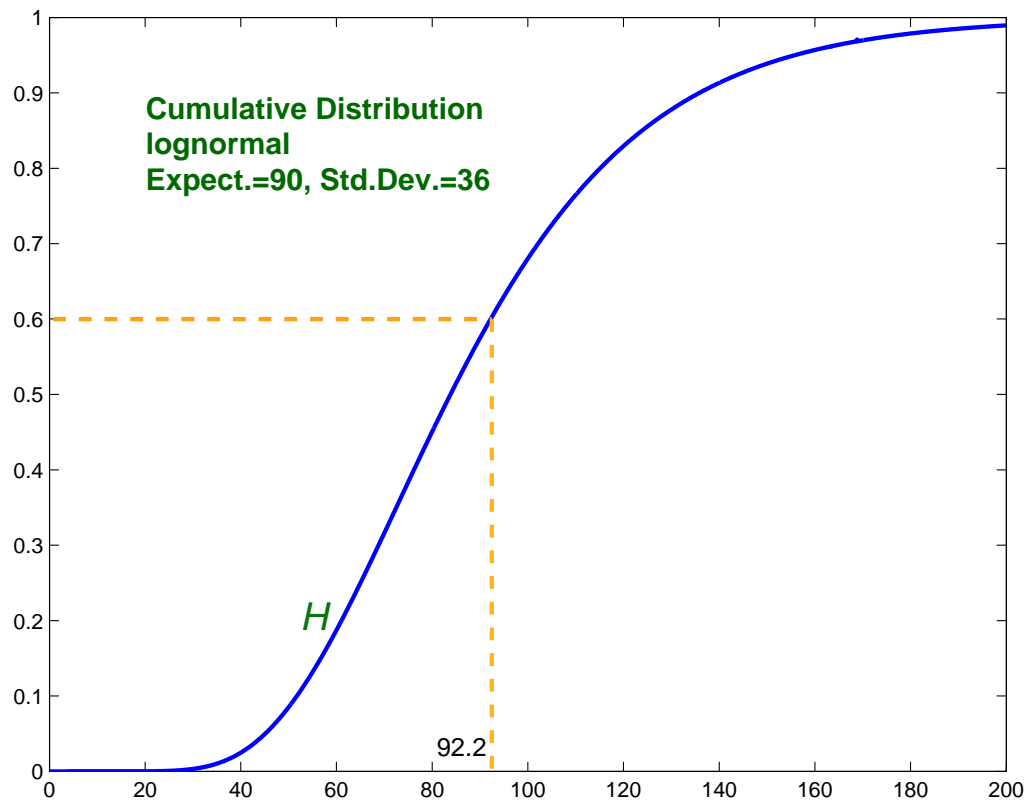
$$\text{where } Q(\xi, x) = \begin{cases} (c+r)\xi & \text{if } \xi \leq x, \\ (c+r)x & \text{if } \xi \geq x \end{cases}$$

$$EQ(x) = (c+r) \left(\int_0^x \xi H(d\xi) + \int_x^\infty x H(d\xi) \right)$$

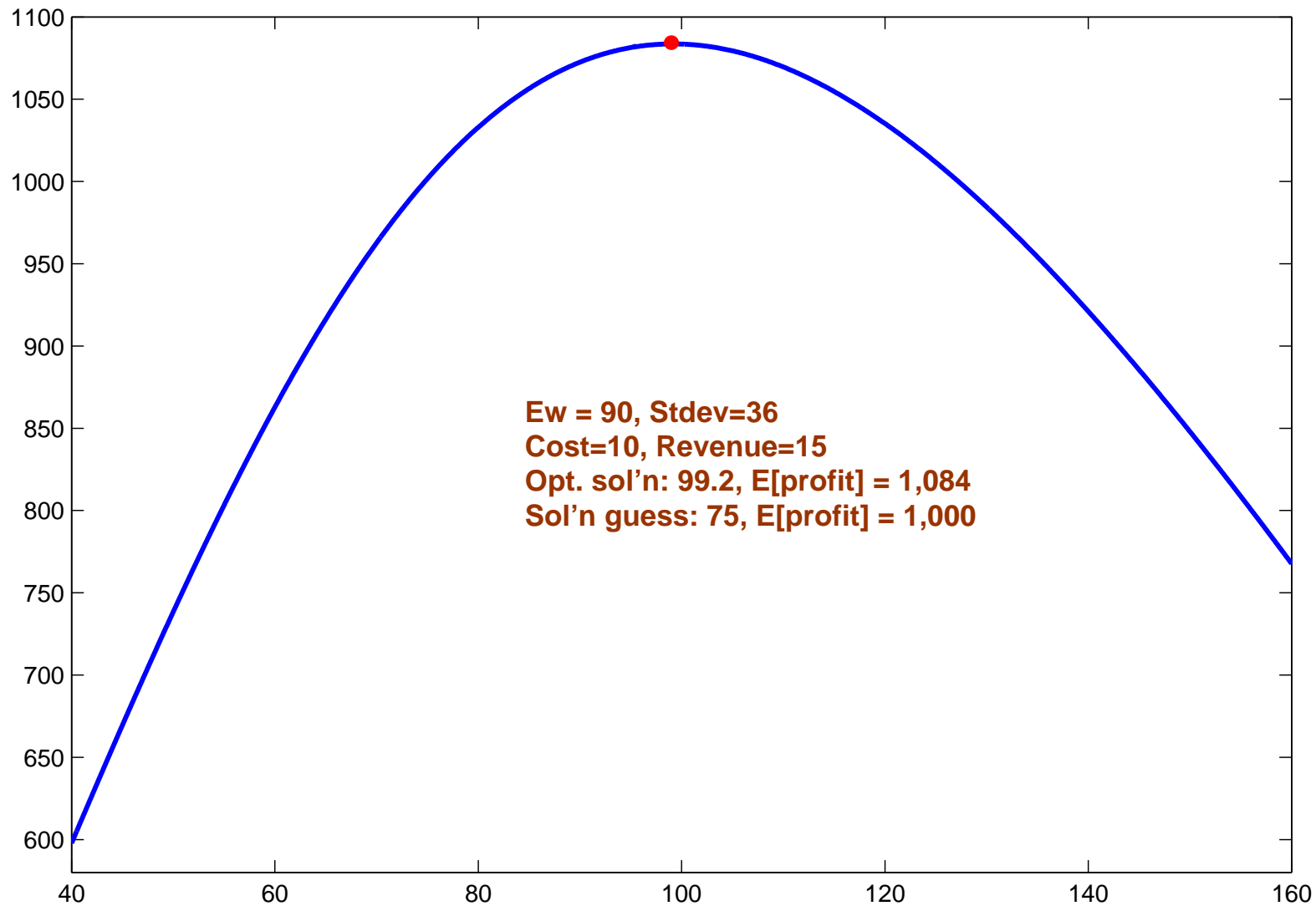
Optimal: Expected Profit


$$x^* = H^{-1}\left(\frac{r}{c+r}\right) = H^{-1}(0.6) = 99.2$$

for $c = 10, r = 15$.



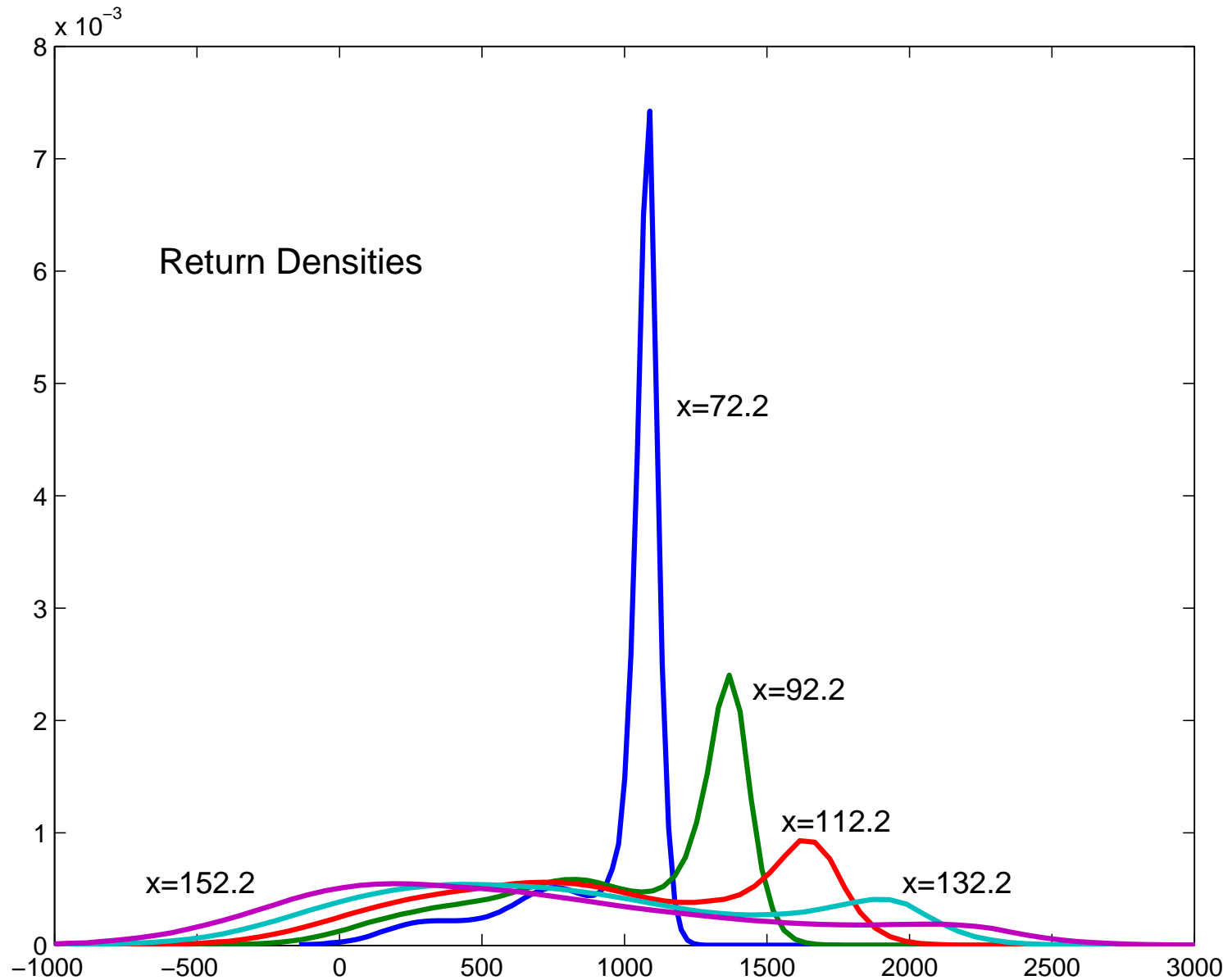
News Vendor's Objective



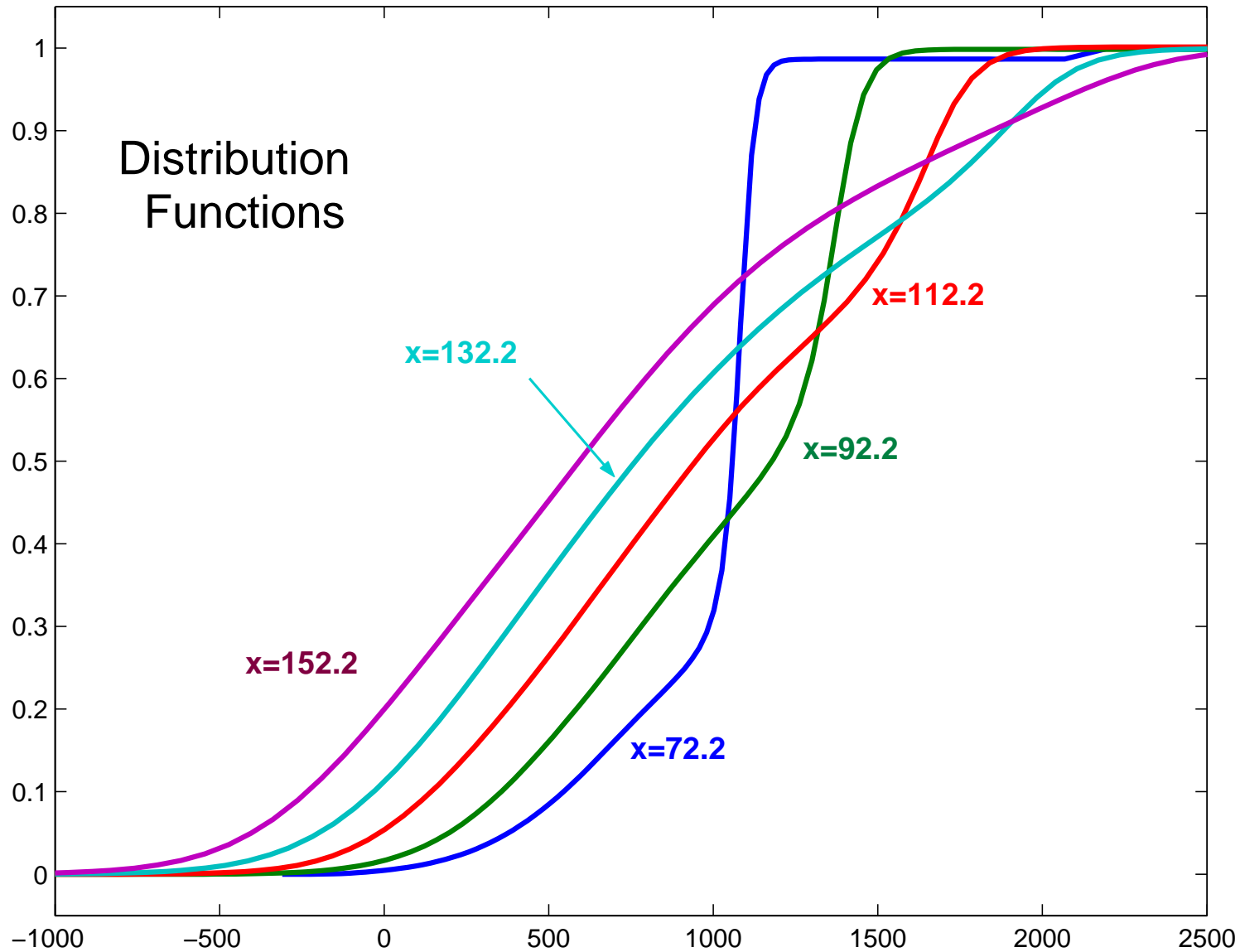


**... but is maximum expected
return the “real” objective?**

The Returns' Densities



Choosing the Returns' Distribution



Decision Criteria

Reducing the choice of a distribution function
to the choice of a “number”

- maximize expected return (scaled?),
- max. $E\{\text{return}\}$ & minimize customers lost,
- minimize Value-at-Risk (VaR),
- minimize the (buffered) probability of failure,
- minimizing a Measure
- variants & combinations of the above

Maximizing Expected Utility

“generic” stochastic optimization problem:

$\max E\{f_0(\xi, x)\}$ such that $f_i(\xi, x) \leq 0, i = 1, \dots, m,$

Risk-averse or risk-seeking \implies *utility function*

von Neuman-Morgenstern: under “rationality” (axiomatics) there exists a utility function u such that $\bar{x} \in \operatorname{argmax} E\{u(f_0(\xi, x))\}$ (subject to the constraints) identifies the preferred return’s distribution

Modeling hurdle: no blueprint for u ’s design!

Robust Optimization

“generic” optimization problem: $\max \gamma$

so that $\gamma - f_0(\xi, x) \leq 0,$

$$f_i(\xi, x) \leq 0, \quad i = 1, \dots, m,$$

“robust” counterpart: $\max \gamma$

so that $\gamma - f_0(\xi, x) \leq 0, \quad \forall \xi \in \mathcal{U} \subset \Xi$

$$f_i(\xi, x) \leq 0, \quad i = 1, \dots, m, \quad \forall \xi \in \mathcal{U},$$

Challenges:

- formulate a computationally tractable robust counterpart
- specify reasonable uncertainty for set \mathcal{U}

Reliability: Chance Constraints

Satisfy constraints with probability $\alpha \in (0, 1]$

$$\min f_0(x) \text{ so that } \text{prob. } [x \in S(\xi)] \geq \alpha$$

Variant:

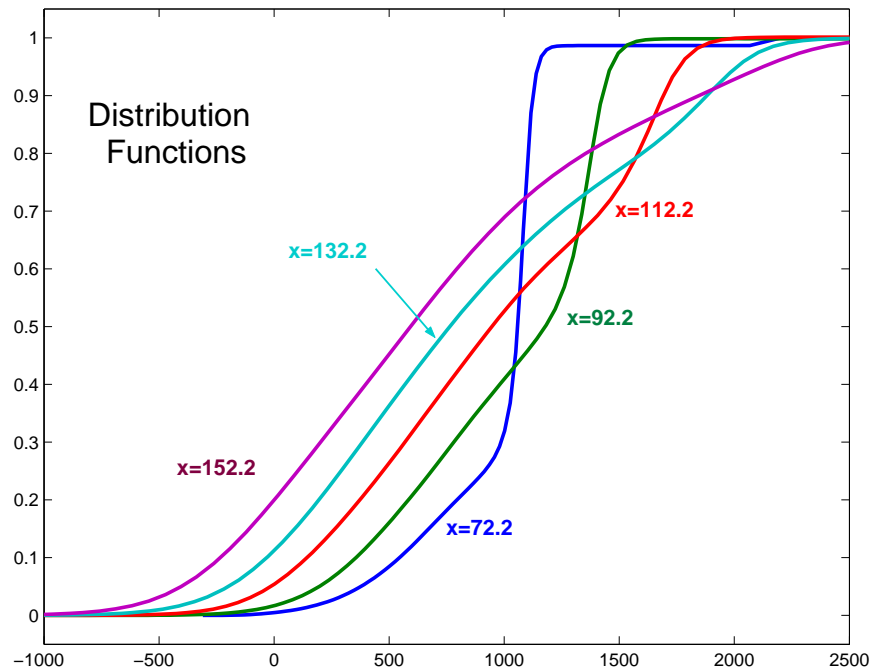
$$\min f_0(x)$$

$$\text{so that } \text{prob. } [f_i(\xi, x) \leq 0] \geq \alpha_i, i \in I$$

α_i dictated by

- contractual obligations, company policy, guess, etc.
- often nonconvex; convex alternative, buffered prob.

Stochastic Dominance



Stochastic Dominance: $D_x(s) \leq D_{\hat{x}}(s), \quad \forall s$

\implies probability of the return to be $\leq s$

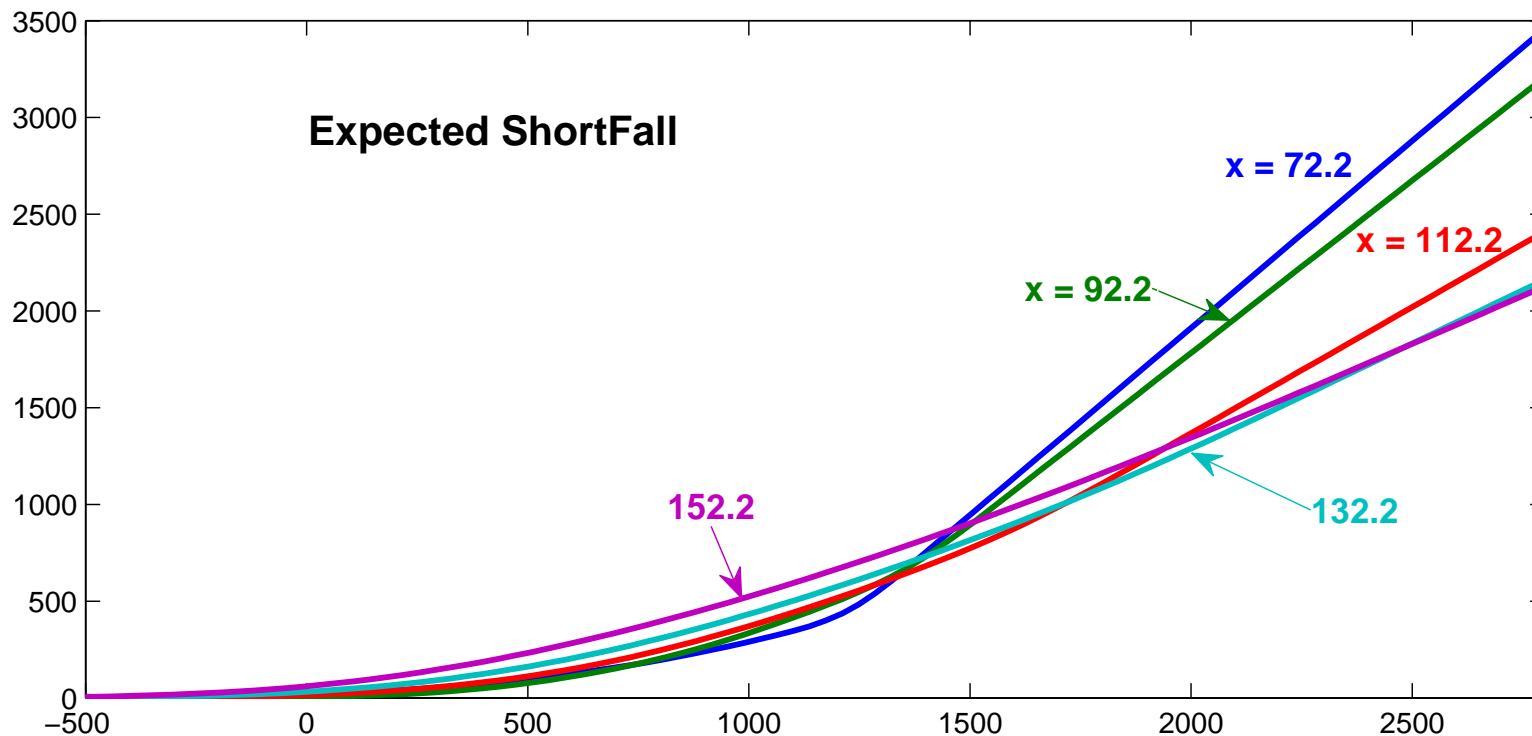
always smaller when choosing x rather than \hat{x}

unfortunately unusual

Second order Stochastic Dominance

$$D^2(s) = \int_{-\infty}^s D(\xi) d\xi = E\{(s - \xi)_+\}$$

D^2 : the expected shortfall function



Stochastic Dominance Constraint

NewsVendor problem

$$\max rx, \quad x \geq 0$$

$$\text{such that } D_x^2(s) \leq G^2(s), \quad s \in [\alpha, \beta]$$

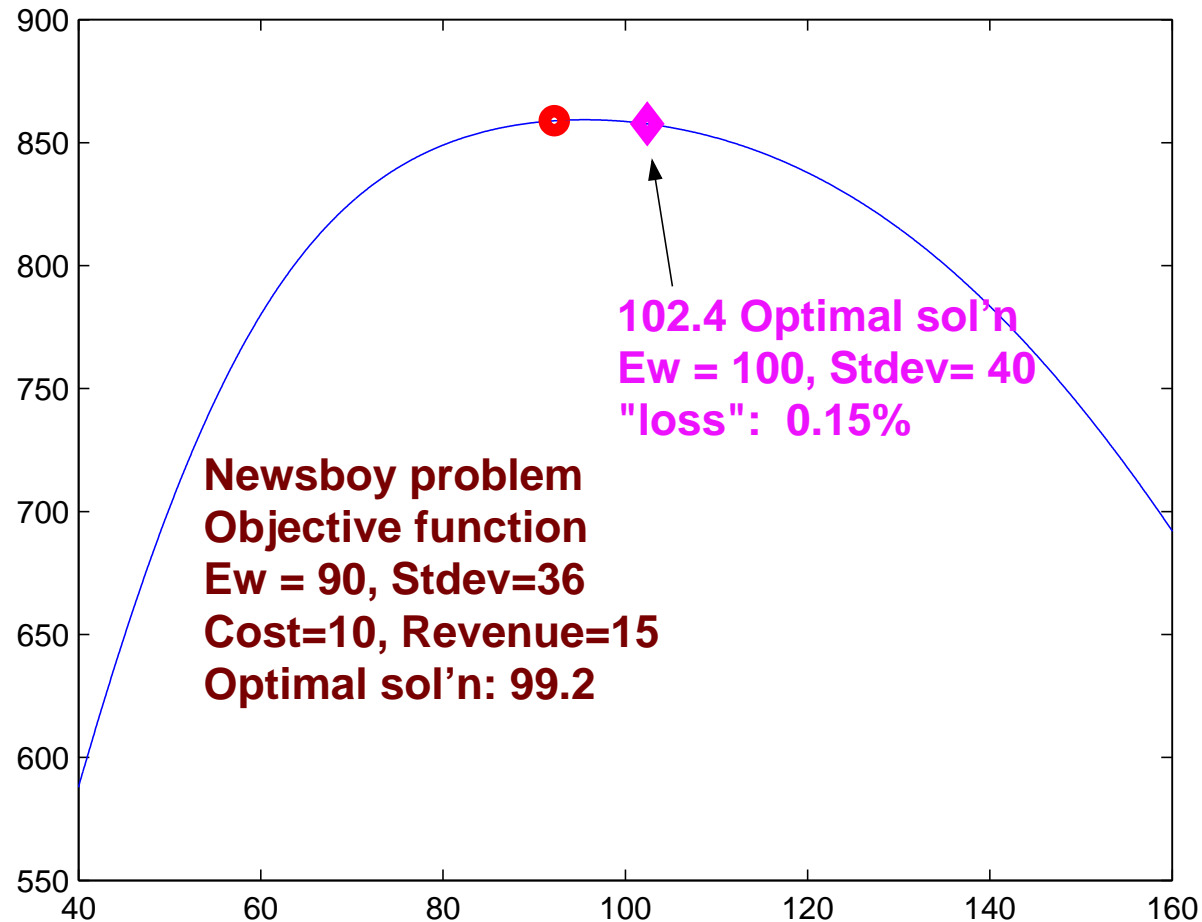
given a “desirable” distribution function G

D_x : distribution of actual return, decision x

$$rx \text{ when } \xi \leq x \text{ and } (c + r)\xi - cx \text{ when } \xi < x$$

leads to a semi-infinite optimization problem
used in portfolio optimization, for example

Perturbing the Probability Measure



stress testing via distribution contamination

A few references - Books

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3. A. King & S. Wallace, *Modeling with Stochastic Programming*, Springer, 2012.
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A few references - Tutorials/Surveys

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2. Upcoming tutorial by A. Shapiro and previous tutorial by J. Luedtke
3. Overview of Risk-averse models:
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