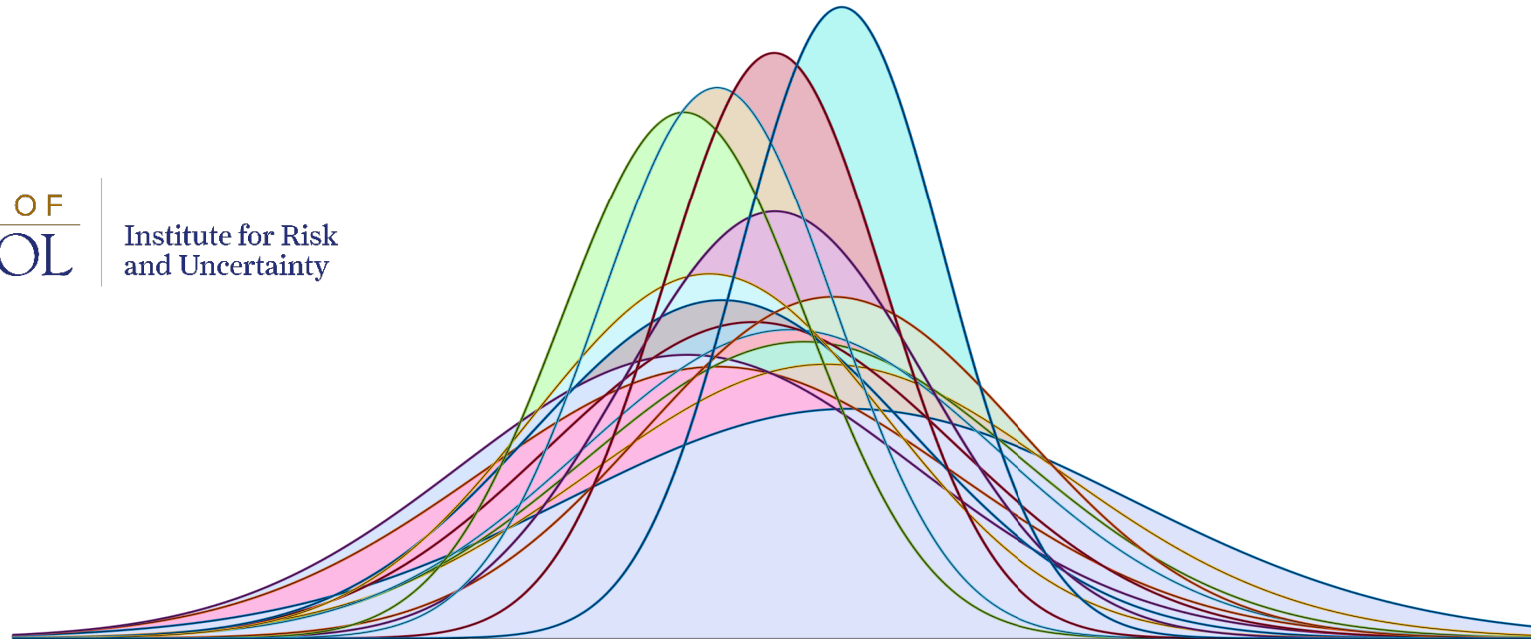


Distribution-free uncertainty propagation



UNIVERSITY OF
LIVERPOOL

Institute for Risk
and Uncertainty



REC₂₀₂₁

Scott Ferson, **Ander Gray**

Ander.Gray@Liverpool.ac.uk



Monte Carlo

- Most popular uncertainty propagation
- Non-intrusive
- Requires specified distributions
- Requires dependence assumptions
- Doesn't consider its own reliability
- Can't see individual variance components

First order error propagation

- In probability theory called “*Moment propagation*”
- Propagates mean & variance through mathematical expressions
- Although widely used, requires a lot of assumptions, and does not inform about distribution shape

Pessimism from Cullen and Frey (1999)

- Bad for risk analysis:
 - Requires stochastic independence
 - Moments must be perfectly known
 - Gives no information about tails without distribution assumption
 - Only accurate for near-linear models

Moment Arithmetic

- Generalisation with the following:
 - Independence between variables need not be assumed
 - Moment propagation formulae may be evaluated with intervals
 - Assumptions about input distributions is no longer necessary
 - Works for non-linear functions

Moment Arithmetic

- Computer arithmetic on random variable's *range, mean* and *variance*
- 7 binary operations: $+$, $-$, $*$, $/$, $^$, $\min$, $\max$
- 9 unary operations: rescale , shift , $\exp$, $\log$, $\ln$, $1/x$, x^2 , $\sqrt{x}$, $|x|$
- For independence and unknown dependence (Fréchet)

<https://github.com/AnderGray/MomentArithmetic.jl>

Interval arithmetic

	<i>Least possible value</i>	<i>Greatest possible value</i>
$k + X$ (shifting)	$\mathbf{k} + \underline{\mathbf{X}}$	$\mathbf{k} + \overline{\mathbf{X}}$
kX (rescaling)	$\begin{cases} \mathbf{k}\underline{\mathbf{X}}, & \text{if } 0 \leq k \\ \mathbf{k}\overline{\mathbf{X}}, & \text{if } k < 0 \end{cases}$	$\begin{cases} \mathbf{k}\overline{\mathbf{X}}, & \text{if } 0 \leq k \\ \mathbf{k}\underline{\mathbf{X}}, & \text{if } k < 0 \end{cases}$
e^X	$e^{\underline{\mathbf{X}}}$	$e^{\overline{\mathbf{X}}}$
$\ln(X)$ for $0 < X$	$\ln(\underline{\mathbf{X}})$	$\ln(\overline{\mathbf{X}})$
$\log_{10}(X)$ for $0 < X$	$\log_{10}(\underline{\mathbf{X}})$	$\log_{10}(\overline{\mathbf{X}})$
$\frac{1}{X}$ for $0 \notin X$	$1/\overline{\mathbf{X}}$	$1/\underline{\mathbf{X}}$
X^2	$\begin{cases} 0, & \text{if } 0 \in X \\ \min(\underline{\mathbf{X}}^2, \overline{\mathbf{X}}^2), & \text{otherwise} \end{cases}$	$\max(\underline{\mathbf{X}}^2, \overline{\mathbf{X}}^2)$
$ X $ (absolute value)	$\begin{cases} 0, & \text{if } 0 \in X \\ \min(\underline{\mathbf{X}} , \overline{\mathbf{X}}), & \text{otherwise} \end{cases}$	$\max(\underline{\mathbf{X}} , \overline{\mathbf{X}})$
$\sqrt{X}$ for $0 \leq X$	$\sqrt{\underline{\mathbf{X}}}$	$\sqrt{\overline{\mathbf{X}}}$
$X + Y$	$\underline{\mathbf{X}} + \underline{\mathbf{Y}}$	$\overline{\mathbf{X}} + \overline{\mathbf{Y}}$
$X - Y$	$\underline{\mathbf{X}} - \overline{\mathbf{Y}}$	$\overline{\mathbf{X}} - \underline{\mathbf{Y}}$
$X \times Y$	$\min(\underline{\mathbf{X}}\underline{\mathbf{Y}}, \underline{\mathbf{X}}\overline{\mathbf{Y}}, \overline{\mathbf{X}}\underline{\mathbf{Y}}, \overline{\mathbf{X}}\overline{\mathbf{Y}})$	$\max(\underline{\mathbf{X}}\underline{\mathbf{Y}}, \underline{\mathbf{X}}\overline{\mathbf{Y}}, \overline{\mathbf{X}}\underline{\mathbf{Y}}, \overline{\mathbf{X}}\overline{\mathbf{Y}})$
$\frac{X}{Y}$ for $0 \notin Y$	$\min(\underline{\mathbf{X}}/\underline{\mathbf{Y}}, \underline{\mathbf{X}}/\overline{\mathbf{Y}}, \overline{\mathbf{X}}/\underline{\mathbf{Y}}, \overline{\mathbf{X}}/\overline{\mathbf{Y}})$	$\max(\underline{\mathbf{X}}/\underline{\mathbf{Y}}, \underline{\mathbf{X}}/\overline{\mathbf{Y}}, \overline{\mathbf{X}}/\underline{\mathbf{Y}}, \overline{\mathbf{X}}/\overline{\mathbf{Y}})$
X^Y for $0 < X$ or $0 < Y$	$\min(\underline{\mathbf{X}}^{\underline{\mathbf{Y}}}, \underline{\mathbf{X}}^{\overline{\mathbf{Y}}}, \overline{\mathbf{X}}^{\underline{\mathbf{Y}}}, \overline{\mathbf{X}}^{\overline{\mathbf{Y}}})$	$\max(\underline{\mathbf{X}}^{\underline{\mathbf{Y}}}, \underline{\mathbf{X}}^{\overline{\mathbf{Y}}}, \overline{\mathbf{X}}^{\underline{\mathbf{Y}}}, \overline{\mathbf{X}}^{\overline{\mathbf{Y}}})$
$\min(X, Y)$	$\min(\underline{\mathbf{X}}, \underline{\mathbf{Y}})$	$\min(\overline{\mathbf{X}}, \overline{\mathbf{Y}})$
$\max(X, Y)$	$\max(\underline{\mathbf{X}}, \underline{\mathbf{Y}})$	$\max(\overline{\mathbf{X}}, \overline{\mathbf{Y}})$

Arithmetic with independence

	Mean	Variance
$X + Y$	$\mathbf{EX} + \mathbf{EY}$	$\mathbf{VX} + \mathbf{VY}$
$X - Y$	$\mathbf{EX} - \mathbf{EY}$	$\mathbf{VX} + \mathbf{VY}$
$X \times Y$	$\mathbf{EXEY}$	$(\mathbf{EX})^2 \mathbf{VY} + (\mathbf{EY})^2 \mathbf{VX} + \mathbf{VXVY}$
$\frac{X}{Y}$ for $0 \notin Y$	$E(X \times (1/Y))$	$V(X \times (1/Y))$
X^Y for $0 < X$ or $0 < Y$	$E(\exp(\ln(X) \times Y))$	$V(\exp(\ln(X) \times Y))$
$\max(X, Y)$	$\begin{cases} \mathbf{EX}, & \text{if } Y < X \\ \mathbf{EY}, & \text{if } X < Y \\ \text{"Bertsimas max"} , & \text{otherwise} \end{cases}$	$\begin{cases} \mathbf{VX}, & \text{if } Y < X \\ \mathbf{VY}, & \text{if } X < Y \\ \text{env}(\max(\mathbf{VX}, \mathbf{VY}), 0) , & \text{otherwise} \end{cases}$
$\min(X, Y)$	$\begin{cases} \mathbf{EX}, & \text{if } X < Y \\ \mathbf{EY}, & \text{if } Y < X \\ \text{"Bertsimas min"} , & \text{otherwise} \end{cases}$	$\begin{cases} \mathbf{VX}, & \text{if } X < Y \\ \mathbf{VY}, & \text{if } Y < X \\ \text{env}(\max(\mathbf{VX}, \mathbf{VY}), 0) , & \text{otherwise} \end{cases}$

Yields intervals

Arithmetic without dependence assumptions (Fréchet)

	Mean	Variance
$k + X$ (shifting)	$\mathbf{k} + \mathbf{EX}$	$\mathbf{VX}$
kX (rescaling)	$\mathbf{kEX}$	$\mathbf{k^2VX}$
e^X	rowe(exp)	rowevar(exp)
$\ln(X)$ for $0 < X$	rowe(ln)	rowevar(ln)
$\log_{10}(X)$ for $0 < X$	rowe(log ₁₀)	rowevar(log ₁₀)
$\frac{1}{X}$ for $0 \notin X$	rowe(reciprocal)	rowevar(reciprocal)
X^2	$(\mathbf{EX})^2 + \mathbf{VX}$	rowevar(square)
$ X $ (absolute value)	$\begin{cases} \mathbf{EX}, & \text{if } 0 \leq \underline{X} \\ -\mathbf{EX}, & \text{if } \overline{X} \leq 0 \\ [\mathbf{EX} , \mathbf{EX} + \sqrt{\mathbf{VX}}(\pi - \text{atan}(\frac{ \mathbf{EX} }{\sqrt{\mathbf{VX}}}))], & \text{if } 0 \in X \end{cases}$	$\max(0, \mathbf{EX}^2 + \mathbf{VX} - E(X)^2)$
$\sqrt{X}$ for $0 \leq X$	rowe($\sqrt{}$)	rowevar($\sqrt{}$)
$X + Y$	$\mathbf{EX} + \mathbf{EY}$	$(\sqrt{\mathbf{VX}} \pm \sqrt{\mathbf{VY}})^2$
$X - Y$	$\mathbf{EX} - \mathbf{EY}$	$(\sqrt{\mathbf{VX}} \pm \sqrt{\mathbf{VY}})^2$
$X \times Y$	$\mathbf{EXEY} \pm \sqrt{\mathbf{VXVY}}$	"Goodman"
$\frac{X}{Y}$ for $0 \notin Y$	$E(X \times (1/Y))$	$V(X \times (1/Y))$
X^Y for $0 < X$ or $0 < Y$	$E(\exp(\ln(X) \times Y))$	$V(\exp(\ln(X) \times Y))$
$\max(X, Y)$	"Bertsimas max"	$\text{env}(\max(\mathbf{VX}, \mathbf{VY}), 0)$
$\min(X, Y)$	"Bertsimas min"	$\text{env}(\max(\mathbf{VX}, \mathbf{VY}), 0)$

Scattered across the literature

- Formula for variance (Goodman)
- Transformations with constant-sign derivatives (Rowe)
- Formulas can often be tighter (Kreinovich)
- Max and min functions (Bertsimas)
- Homegrown inequalities

Rowe formulas for moment transformations

- Formulas for constant-sign-derivative transformations
- $t = \{\exp, \ln, \log, 1/x, \sqrt{x}\}$

$$\text{rowe}(t) = \text{env}\left(Pt(\underline{X}) + (1 - P)t\left(EX + \frac{VX}{EX - \underline{X}}\right), Qt(\overline{X}) + (1 - Q)t\left(EX + \frac{VX}{EX - \overline{X}}\right)\right) \quad (1)$$

$$\text{rowevar}(t) = \text{env}\left(\frac{t(\underline{\nu}) - t(\underline{X})}{(\underline{\nu} - \underline{X})^2}(VX + (\underline{\nu} - EX)^2), \frac{t(\overline{\nu}) - t(\overline{X})}{(\overline{\nu} - \overline{X})^2}(VX + (\overline{\nu} - EX)^2)\right) \quad (2)$$

$$P = 1/(1 + (EX - \underline{X})^2/VX)$$

$$Q = 1/(1 + (EX - \overline{X})^2/VX)$$

Rowe, N. C., [Absolute bounds on the mean and standard deviation of transformed data for constant-sign-derivative transformations](#), SIAM journal on scientific and statistical computing 9 (6) (1988) 1098–1113.

Goodman variance for product

$$V(XY) = (EX)^2VY + (EY)^2VX + 2EXEYE_{11} + 2EXE_{12} + 2EYE_{21} + E_{22} - E_{11}^2$$

$$E_{ij} = E[(X - EX)^i(Y - EY)^j] \text{ (e.g. } E_{11} \text{ is covariance)}$$

$$\begin{aligned} E_{11} &= E[(X - EX)(Y - EY)] \\ &= E[XY - XEY - YEX + EXEY] \\ &= E[XY] - EXEY \end{aligned}$$

Goodman, L. A., [On the exact variance of products](#), Journal of the American statistical association 55 (292) (1960) 708–713.

Goodman variance for product

$$\begin{aligned}E_{21} &= E[(X - EX)^2(Y - EY)] \\&= E[X^2Y + E[X]^2Y - X^2E[Y] + 2E[X]E[Y]X - 2E[X]XY - E[X]^2E[Y]] \\&= E[X^2Y] + E[X]^2E[Y] - E[X^2]E[Y] + 2E[X]^2E[Y] - 2E[X]E[XY] - E[X]^2E[Y] \\&= E[X^2Y] - E[X^2]E[Y] + 2E[X]^2E[Y] - 2E[X]E[XY]\end{aligned}$$

$$\begin{aligned}E_{22} &= E[(X - EX)^2(Y - EY)^2] \\&= E[E[X]^2E[Y]^2 - 2E[X]E[Y]^2X + E[Y]^2X^2 - 2E[X]^2E[Y]Y + 4E[X]E[Y]XY \\&\quad - 2E[Y]X^2Y + E[X]^2Y^2 - 2E[X]XY^2 + X^2Y^2] \\&= E[X]^2E[Y]^2 - 2E[X]^2E[Y]^2 + E[Y]^2E[X^2] - 2E[X]^2E[Y]^2 + 4E[X]E[Y]E[XY] \\&\quad - 2E[Y]E[X^2Y] + E[X]^2E[Y^2] - 2E[X]E[XY^2] + E[X^2Y^2] \\&= -3E[X]^2E[Y]^2 + E[X^2]E[Y]^2 + E[X]^2E[Y^2] + 4E[X]E[Y]E[XY] \\&\quad - 2E[Y]E[X^2Y] - 2E[X]E[XY^2] + E[X^2Y^2]\end{aligned}$$

Using the tables with intervals

- Tables can be readily evaluated with *interval arithmetic*
- EX , VX , and bounds can be intervals
- Repeated variables
- Subintervalisation: n^m interval calculations
- Since only 2 variables are intervals, subintervalisation works well
- $n \approx 20 ; m = 2$

Missing information

Mean and variance bounds can be estimated from range

```
julia> A = Moments(mean = missing, var = missing, range = interval(2.3, 7))
moment:      ~ ( mean = [2.3,7.0], var = [0.0,5.5225], range = [2.3,7.0] )

julia> B = Moments(mean = 3, var = missing, range = interval(2.3, 7))
moment:      ~ ( mean = 3, var = [0.0,2.8000000000000007], range = [2.3,7.0] )

julia> C = Moments(mean = interval(5, 10), var = missing, range = interval(2.3, 7))
moment:      ~ ( mean = [5.0,7.0], var = [0.0,5.4], range = [2.3,7.0] )

julia> D = Moments(mean = interval(10, 30), var = missing, range = interval(2.3, 7))
ERROR: ArgumentError: Provided information not valid. Mean n Range = ∅.
[10, 30] n [2.29999, 7] = ∅

julia> E = Moments(mean = 3, var = interval(15,18), range = interval(2.3, 7))
ERROR: ArgumentError: Provided information not valid. Variance n VarBounds = ∅.
[15, 18] n [0, 2.80001] = ∅
```

Mean and variance
bounded from ranges

Tighter variance when
mean is given

Given mean is too wide

Inconsistent mean

Inconsistent variance

Example: mink dietary mercury

$$TDI = FMR \times \left(\frac{C_{\text{fish}} \times P_{\text{fish}}}{AE_{\text{fish}} \times GE_{\text{fish}}} + \frac{C_{\text{inverts}} \times P_{\text{inverts}}}{AE_{\text{inverts}} \times GE_{\text{inverts}}} \right)$$

$C_{\text{fish}}, C_{\text{inverts}}$ mercury concentration in fish or invertebrate tissue

$P_{\text{fish}}, P_{\text{inverts}}$ proportion of fish or inverts in the mammal's diet

$GE_{\text{fish}}, GE_{\text{inverts}}$ gross energy of fish or invertebrate tissue

$AE_{\text{fish}}, AE_{\text{inverts}}$ assimilation efficiency for dietary fish or inverts

FMR normalized free metabolic rate

BW body mass of the mammal

Hibernian Mink



Input ranges and moments

	Min	Max	Mean	Stdev	Units
GE_{fish}	500	2000	1200	240	Kcal per kg
GE_{inverts}	300	1700	1050	225	Kcal per kg
AE_{fish}	0.77	0.98	0.91	?	
AE_{inverts}	0.72	0.96	0.87	?	
C_{fish}	0.1	0.3	?	?	mg per kg
C_{inverts}	0.02	0.06	?	?	mg per kg
P_{fish}	0.9	0.9	0.9	0	
P_{inverts}	0.1	1.0	0.9	0	
BW	400	800	608	66.9	gram

$$TDI = FMR \mid * \mid (C_{\text{fish}} * P_{\text{fish}} / (AE_{\text{fish}} \mid * \mid GE_{\text{fish}})) + C_{\text{inverts}} * P_{\text{inverts}} / (AE_{\text{inverts}} \mid * \mid GE_{\text{inverts}})$$

$$FMR = [0.412 \pm 0.058] * \text{mag}(BW) \wedge [0.862 \pm 0.026] * 1 \text{ Kcal per kg per day}$$

(vertical bars denote independence between the operands)

Total daily intake, mg per kg per day

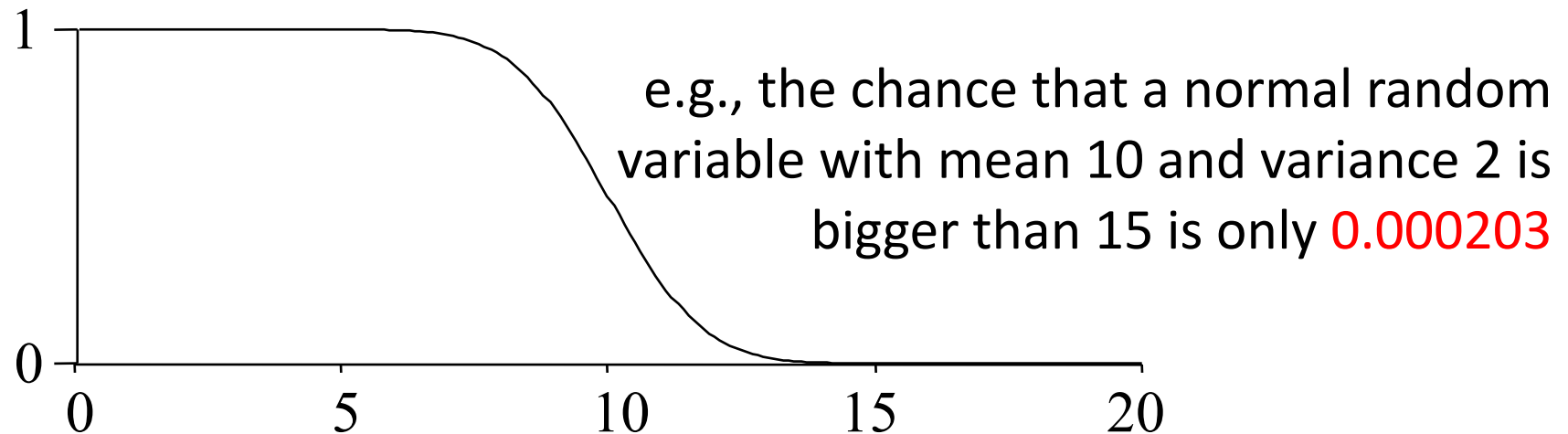
Mean	[0.06,0.47]
Stdev	[0.04, 0.51]
Max	1.19

Useful information even though some inputs
were totally unknown

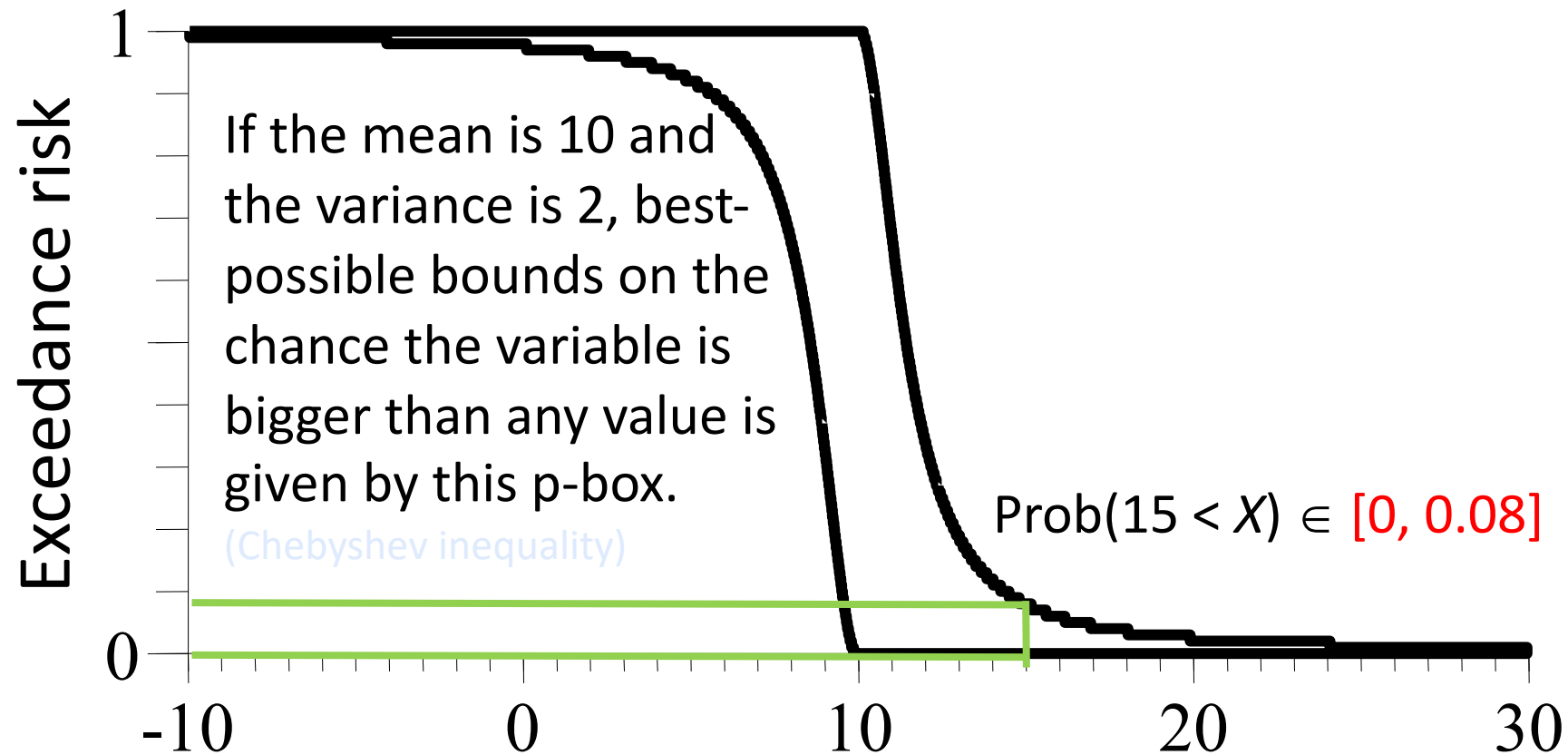
What do range and moments say about risks?

Traditional normal theory

- Distribution determined by mean and variance
- Probability in tails drops quickly

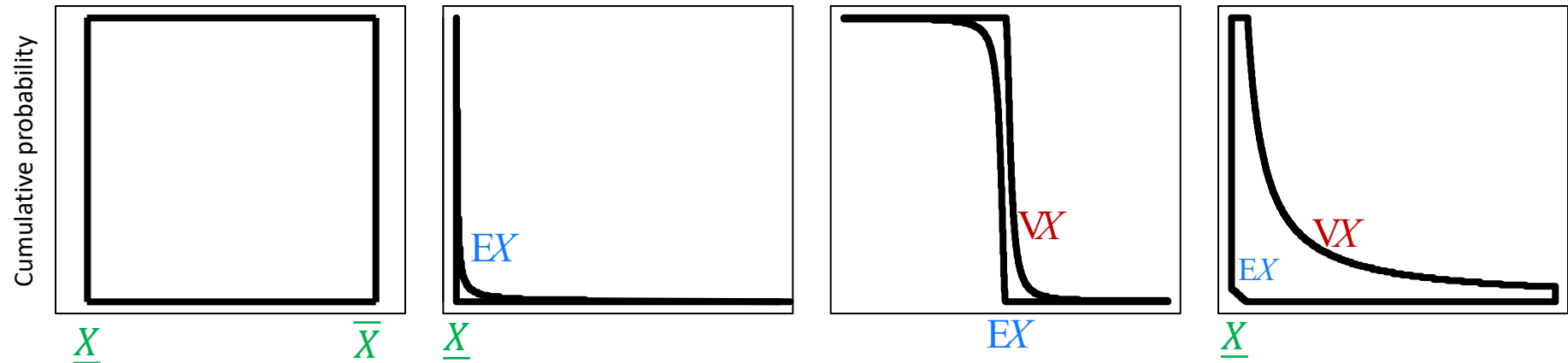


What do moments say about risks?

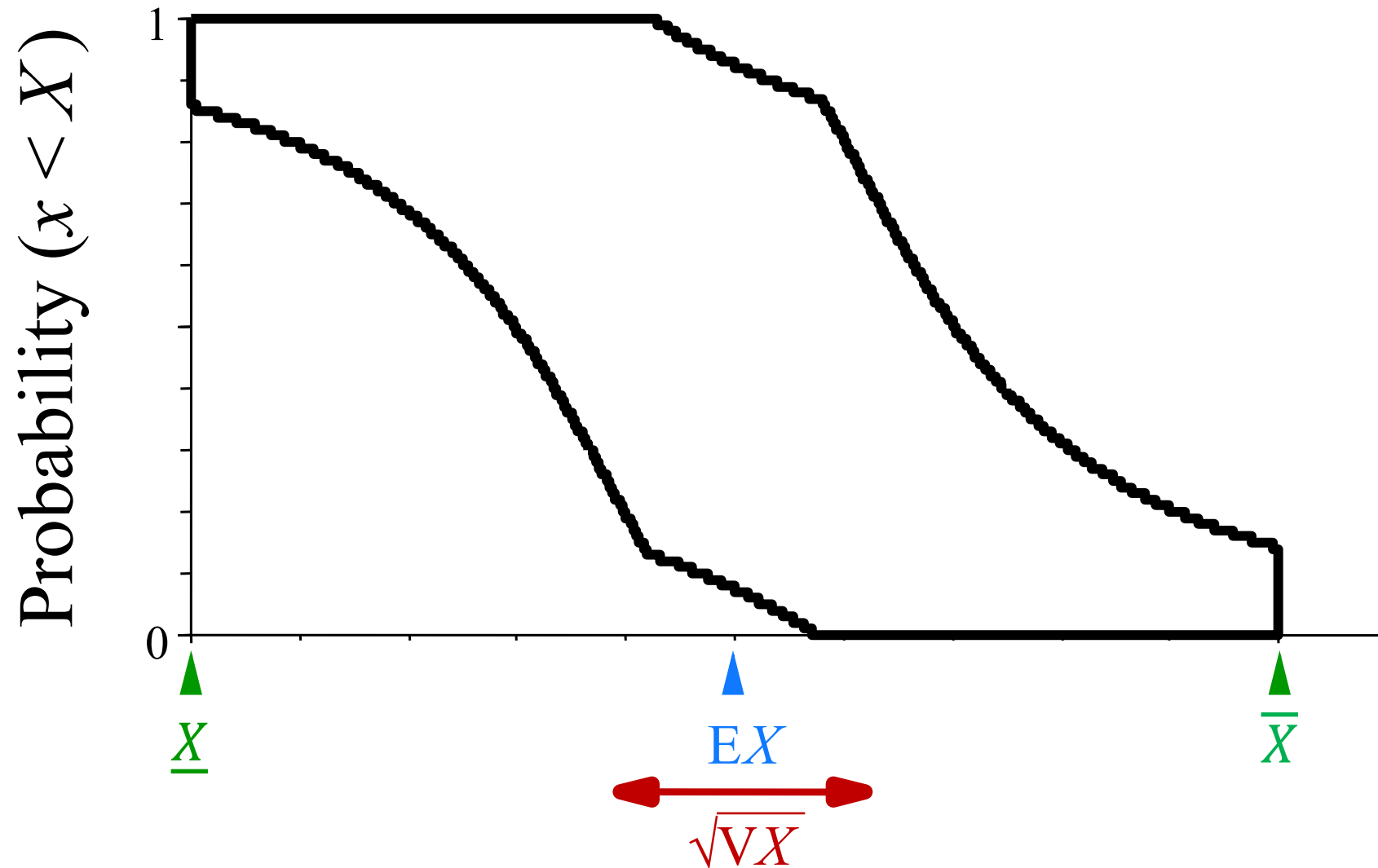


Constraining probabilities and quantiles

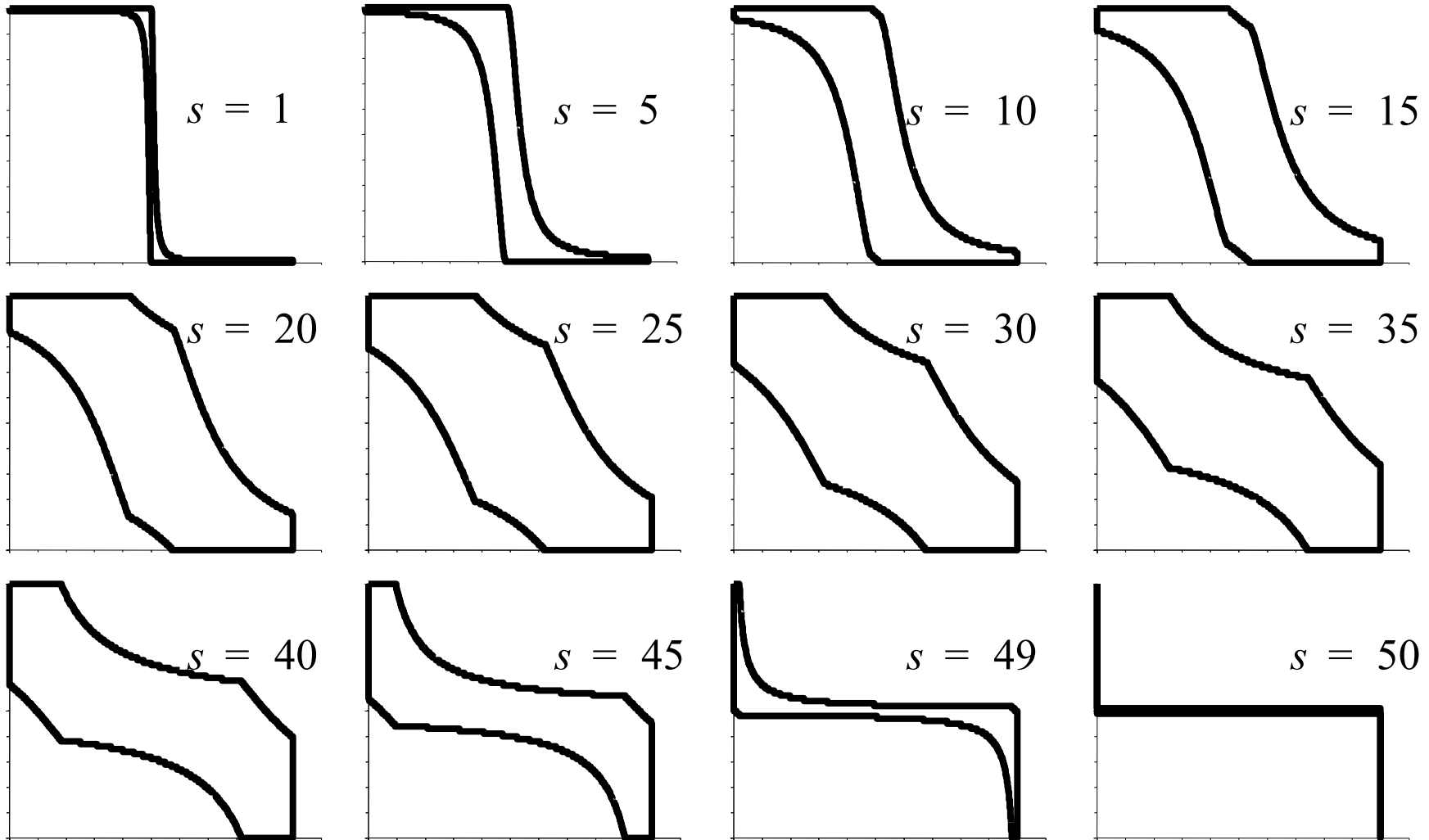
- Interval $\{\underline{X}, \bar{X}\}$
- Markov $\{\underline{X}, EX\}, \{EX, \bar{X}\}$
- Chebyshev $\{EX, VX\}$
- Cantelli $\{\underline{X}, EX, VX\}, \{EX, VX, \bar{X}\}$



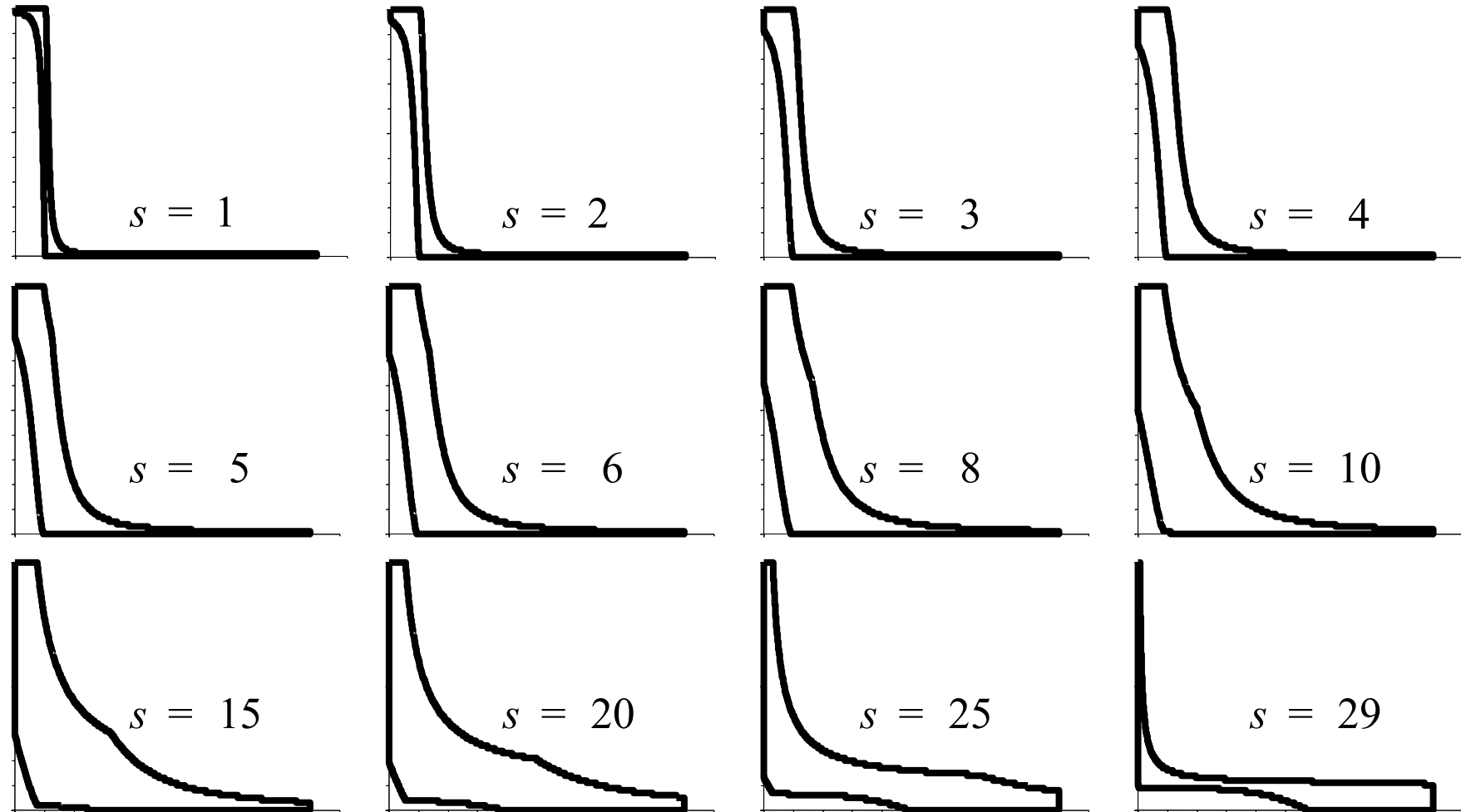
Range and moments *together*



$\{\text{min} = 0, \text{max} = 100, \text{mean} = 50, \text{stdev} = s\}$



$\{\text{min} = 0, \text{max} = 100, \text{mean} = 10, \text{stdev} = s\}$



Moment arithmetic: *distribution-free* risk analysis

- Fully probabilistic (with tails)
- Rigorous rather than approximate
- Needs no distributions, dependencies, derivatives
- Some moments/bounds can be totally unknown
- Bounds may be too wide to be useful
- Requires dependency tracking
- Intrusive

Grazie

Covariance arithmetic

$$Z = X + Y$$

$$VZ = VX + 2\text{Cov}[X, Y] + VY$$

$$Z = X - Y$$

$$VZ = VX - 2\text{Cov}[X, Y] + VY$$

$$Z = X * Y$$

$$VZ = \text{Goodman}$$

$$Z = X / Y$$

$$VZ = V[X * (1/Y)]$$

Springer, Melvin Dali. *The algebra of random variables*. No. 519.2 S67. 1979

Covariance tracking

$$Z = X + Y$$

$$\text{Cov}[Z, X] = VX + \text{Cov}[X, Y]$$

$$Z = X - Y$$

$$\text{Cov}[Z, X] = VX - \text{Cov}[X, Y]$$

$$Z = X * Y$$

$$\text{Cov}[Z, X] = E[X^2 Y] - E[XY]EX$$

(Numerator) $Z = X / Y$

$$\text{Cov}[Z, X] = E[X^2 / Y] - E[X / Y]EX$$

(Denominator) $Z = X / Y$

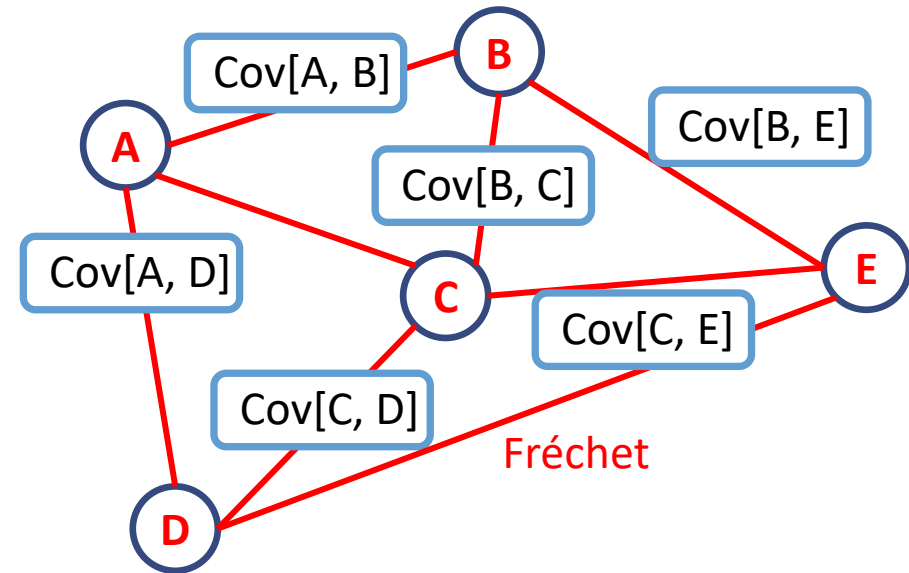
$$\text{Cov}[Z, Y] = EX - E[X / Y]EY$$

Springer, Melvin Dali. *The algebra of random variables*. No. 519.2 S67. 1979

Covariance tracking

- Graphical model for storing dependencies
- Nodes are moment variables
- Edges are covariances
- Make a new node and edge when a variable is created

$$\begin{array}{c} \text{Cov}[B, C] \\ \swarrow \quad \searrow \\ E = B + C \\ \swarrow \quad \searrow \\ \text{Cov}[C, E] \quad \text{Cov}[B, E] \\ \\ F = D + E \quad \text{Fréchet} \end{array}$$



Covariance tracking

The graph is an ill-specified covariance matrix

	A	B	C	D
A	VA^2	$C[A, B]$	$C[A, C]$	$C[A, D]$
B	$C[B, A]$	VB^2	$C[B, C]$	???
C	$C[C, A]$	$C[C, B]$	VC^2	???
D	$C[D, A]$	???	???	VD^2

$$D = A + C$$