

# **Stochastic Programming**



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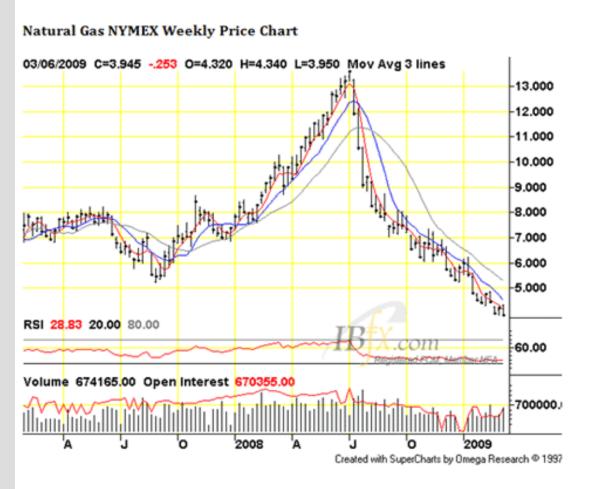


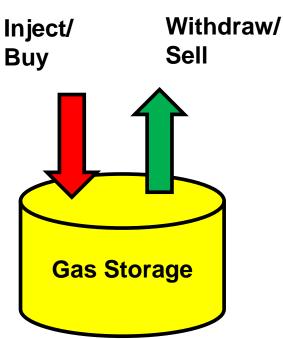




### **Example Model: Gas Price Model**

2010





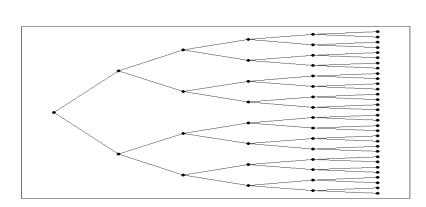


## n-Stage Stochastic Programs

- Construct Scenario Tree:
  - Start with today's price and use a (discrete) distribution
  - Realizations: up, down
- Stochastic Linear Program (block structure)
  - Nested Bender's Decomposition (OSLSE, FortSP, AIMMS)
  - In practice Deterministic Equivalent with Barrier method

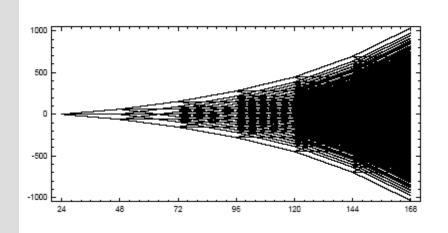
$$Z_{H\!N} = \min_{\mathbf{x}_1} \quad \left\{ \ c_1 \mathbf{x}_1 + E_{\xi_2} \bigg[ \min_{\mathbf{x}_2} c_2 \mathbf{x}_2 + E_{\xi_3 \mid \xi_2} \bigg[ \min_{\mathbf{x}_3} c_3 \mathbf{x}_3 + \ldots + E_{\xi_T \mid \xi_{T-1} \mid \ldots \mid \xi_2} \ \min_{\mathbf{x}_T} c_T \mathbf{x}_T \bigg] \bigg] \right\}$$

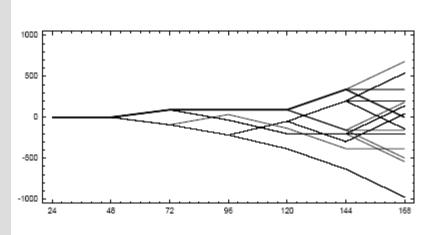
subject to:





# ScenRed (Römisch et. al., HU Berlin)





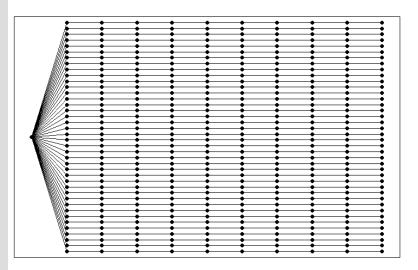
- Find good approximation of original scenario tree of significant smaller size
- Available since 2002
- Integrated in GAMS system
- No extra cost

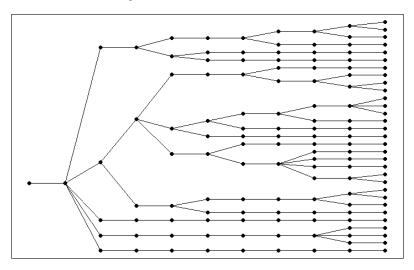


### **Tree Generation: ScenRed2**

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Construct a true scenario tree from independent scenarios:





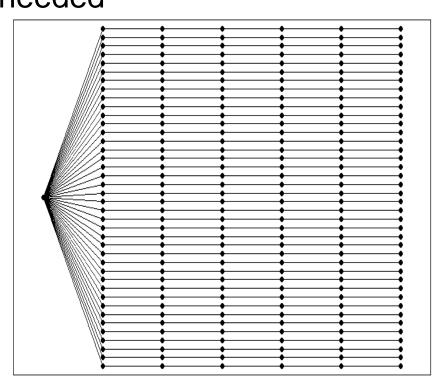
Reconstruct underlying distribution from a set of scenarios



### 2-Stage Stochastic Programs

2010

- SP Solver DECIS (Gerd Infanger, Stanford, USA)
  - Stores only one instance of the problem and generates scenario sub-problems as needed
  - Solution Strategies
    - Deterministic Equivalent (all scenarios)
    - Sampling: Crude Monte Carlo/ Importance sampling





## **AML and Stochastic Programming (SP)**

- Algebraic Modeling Languages/Systems good way to represent optimization problems
  - Algebra is a universal language
  - Hassle free use of optimization solvers
  - Simple connection to data sources (DB, Spreadsheets, ...) and analytic engines (GIS, Charting, ...)
- Large number of (deterministic) models in production
  - Opportunity for seamless introduction of new technology like Global Optimization, Stochastic Programming, ...
  - AML potential framework for SP



# Simple Example











## Simple Example: Newsboy (NB) Problem

- Data:
  - A newsboy faces a certain demand for newspapers d = 45
  - He can buy newspapers for fixed costs per unit c = 30
  - He can sell newspapers for a fixed price r = 60
  - For hold units he has to pay a disposal fee h = 10
  - He has to satisfy his customers demand or has to pay a penalty p = 5
- Decisions:
  - How many newspapers should he buy:
    How many newspapers should he sell:
    45
    45
- Derived Outcomes:
  - How many newspapers need to be disposed:
    How many customers are lost:



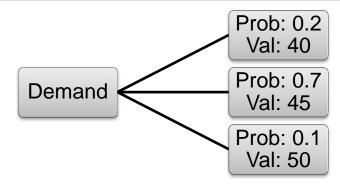
### Simple NB Problem – GAMS Formulation

```
*
          LostSales = demand - UnitsSold
lSales.. L = e = d - S;
*
          Inventory = UnitsBought - UnitsSold
          I = e = X - S;
Inv..
*
          Profit, to be maximized
Profit.. Z = e = r*S - c*X - h*I - p*L;
Model nb / all /;
solve nb max z use lp;
```



### **NB Problem – Add Uncertainty**

Uncertain demand d



- Decisions to make:
  - How much newspaper should he buy "here and now" (without knowing the outcome of the uncertain demand)?
    - → First-stage decision
  - How many customers are lost after the outcome becomes known?
    - → Second-stage or recourse decision
  - Recourse decisions can be seen as
    - penalties for bad first-stage decisions
    - variables to keep the problem feasible



### Stochastic NB Problem – GAMS Extension

```
* Make d uncertain
randvar d discrete 0.2 40
0.7 45
0.1 50
```

\* Define non-default stages stage 2 d I L S stage 2 lSales Inv



# New GAMS (EMP) Keywords



### **Excursus: EMP, what?**

With new modeling and solution concepts do not:

- overload existing GAMS notation right away!
- · attempt to build new solvers right away!

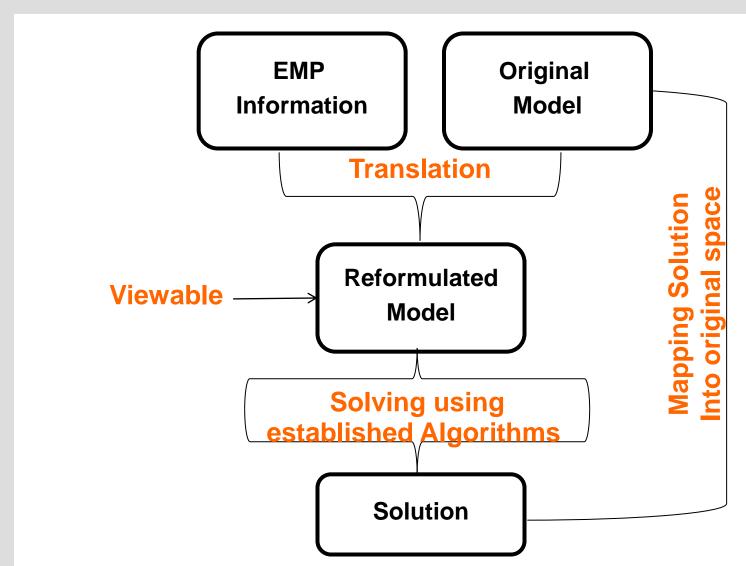
#### But:

- Use existing language features to specify additional model features, structure, and semantics
- Express extended model in symbolic (source) form and apply existing modeling/solution technology
- · Package new tools with the production system

#### → Extended Mathematical Programming (EMP)



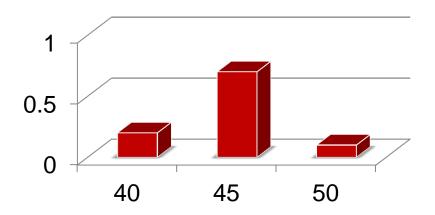
### **JAMS: a GAMS EMP Solver**



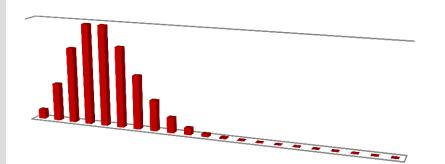


### **Random Variables**

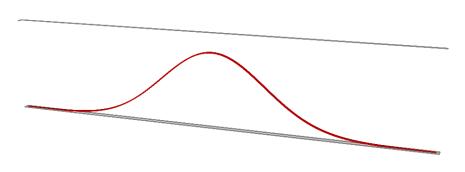
### **Discrete Distribution**



#### **Poisson Distribution**



#### **Normal Distribution**



**Exponential Distribution** 



### Random Variables (RV) [randVar]

Defines both discrete and parametric random variables:

```
randVar rv discrete prob val {prob val}
```

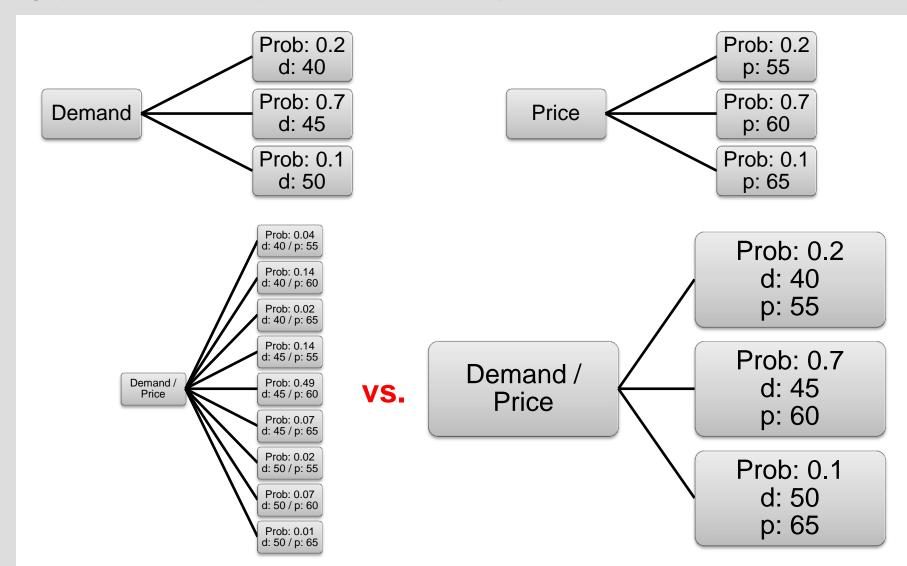
 The distribution of discrete random variables is defined by pairs of the probability prob of an outcome and the corresponding realization val

```
randVar rv distr par {par}
```

• The name of the parametric distribution is defined by distr, par defines a parameter of the distribution



#### **Joint Random Variables**





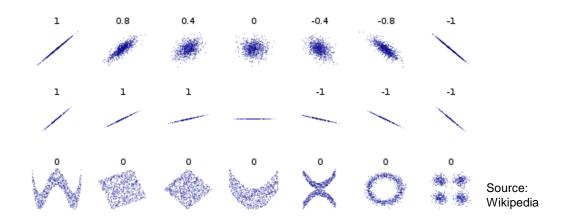
### Joint RVs [jRandVar]

Defines discrete random variables and their joint distribution:

- At least two discrete random variables rv are defined and the outcome of those is coupled
- The probability of the outcomes is defined by prob and the corresponding realization for each random variable by val



### Correlation between RVs [correlation]



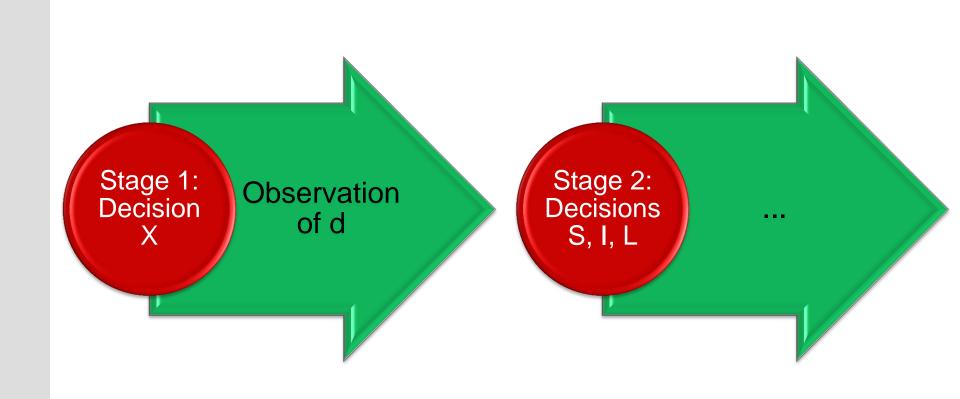
• Defines a correlation between a pair of random variables:

correlation rv rv val

 rv is a random variable which needs to be specified using the randvar keyword and val defines the desired correlation (-1 ≤ val ≤ 1)



## **Stages**





### Stages [stage]

• Defines the stage of random variables (rv), equations (equ) and variables (var):

```
stage stageNo rv | equ | var {rv | equ | var}
```

- StageNo defines the stage number
- The default StageNo for the objective variable and objective equation is the highest stage mentioned
- The default StageNo for all the other random variables, equations and variables not mentioned is 1



### **Chance Constraints**

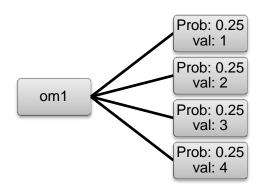
```
OBJ.. Z =e= X1 + X2;

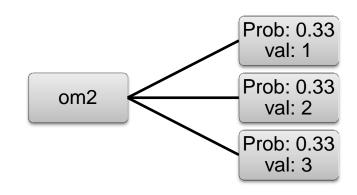
E1.. om1*X1 + X2 =g= 7;

E2.. om2*X1 + 3*X2 =g= 12;

Model sc / all /;

solve sc min z use lp;
```





chance E1 0.6 chance E2 0.6



#### **Chance Constraints**

Just in case: X1 = 2 and X2 = 3 are optimal.



### **Chance Constraints [chance]**

Defines individual or joint chance constraints (CC):

```
chance equ {equ} [holds] minRatio [weight|varName]
```

- Individual CC: A single constraint equ has to hold for a certain ratio (0 ≤ minRatio ≤ 1) of the possible outcomes
- Joint CC: A set of constraints equ has to hold for a certain ratio (0 ≤ minRatio ≤ 1) of the possible outcomes
- If weight is defined, the violation of a CC gets penalized in the objective (weight violationRatio)
- If varName is defined the violation get multiplied by this existing variable



### Expected Value [ExpectedValue]

This is the default objective:

```
ExpectedValue [x EV x]
```

- If only ExpectedValue is defined, the expect value of the GAMS objective variable will be optimized (same as if it would be omitted at all)
- If the variable pair  $x \in V_x$  is defined, GAMS will replace its objective variable by  $EV_x$ , which will become the expected value of x



### Conditional Value at Risk [cVaR]

 As an alternative to the expected value, the conditional value at risk (cVaR) can be optimized:

```
cVaR [x cVaR x] theta
```

- If only cVaR theta is defined, the cVaR of the GAMS objective variable to the quantile level theta will be optimized
- If the variable pair  $x \text{ cVaR}\_x$  is defined, GAMS will replace its objective variable by  $\text{cVaR}\_x$ , which will become the cVaR of x to the quantile level theta



## Combining EV and cVaR

It is also possible optimize a combination of the expected value and the conditional value at risk like this:

```
defobj..
  obj =e= lambda*EV_r + (1-lambda)*CVaR_r;

ExpectedValue r EV_r
cvarlo r CVaR_r 0.1
...
```



### **Output Extraction**

- The expected value of the solution can be accessed via the regular . ⊥ and . M fields
- In addition, the following information can be stored in a parameter by scenario:

```
    level: Levels of variables or equations
```

- marginal: Marginals of variables or equations
- randvar: Realization of a random variable
- opt: Probability of each scenario
- This needs to be stored in a separate dictionary:



## **Adding Uncertainty to Transport**

```
Model transport /all/;
file emp / '%emp.info%' /; put emp '* problem %gams.i%'/;
$onput
randvar b('new-york') normal 325 50
randvar b('chicago') normal 300 50
randvar b('topeka') normal 275 50
stage 2 b demand
$offput
putclose emp;
Set scen scenarios / s1*s6 /;
Parameter
   s b(scen, j) demand realization by scenario
   s x(scen,i,j) shipment per scenario
   s s(scen);
Set dict / scen .scenario.''
          b .randvar .s b
          x .level .s x /;
Solve transport using emp minimizing z scenario dict;
```



# Summary



### **Available GAMS SP Solvers**

	$\mathbf{DE}$	DECIS	LINDO
chance			
correlation			$\checkmark$
cVaR			
expectedValue			
jrandVar		$\sqrt{}$	$\checkmark$
randVar (discrete)			$\checkmark$
randVar (parametric)		-	



#### Conclusion

- Deterministic examples from all kind of application areas exist already (e.g. ~400 in the GAMS Model Library)
- Easy to add uncertainty to existing deterministic models, to
  - ... either use specialized algorithms (DECIS, LINDO)
  - ... or create Deterministic Equivalent and select from wide range of existing GAMS solver links (DE, free)
- New SP examples in the GAMS EMP Library
- More work to be done:
  - Scenario tree support
  - Sampling
- **-** ..



### **Contacting GAMS**

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