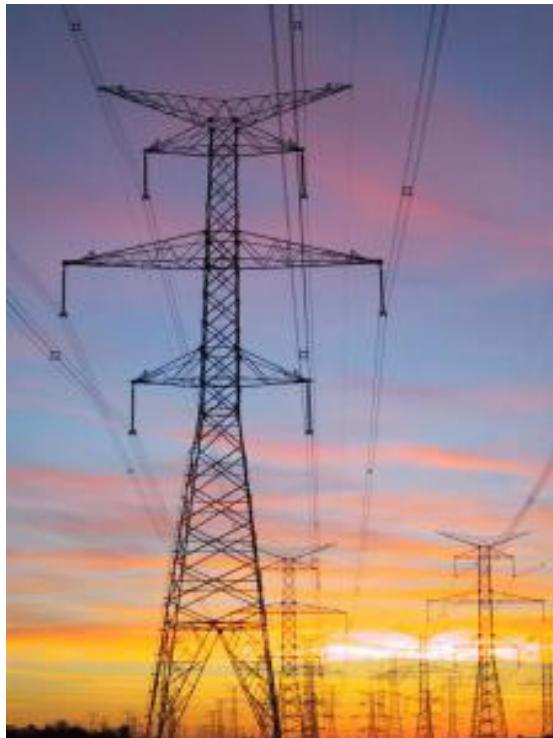




Impact of Unit Failure on Forward Contracting



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Outline

- Introduction: unit availability
- Problem characterization
- Proposed solution
- Numerical simulations
- Future work



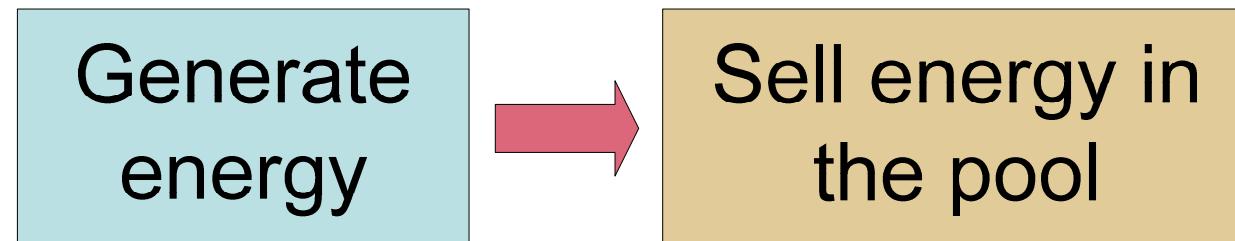
Power Producer

- Selling in the pool:
 - Risky because of the volatility of the prices
- Selling through forward contracts:
 - Hedging strategy against pool price volatility
 - Risky due to the probability of unit failure



Understanding the problem

- High pool prices volatility
- Risk-neutral producer



Expected profit ↑
Standard deviation of profit ↑



Understanding the problem

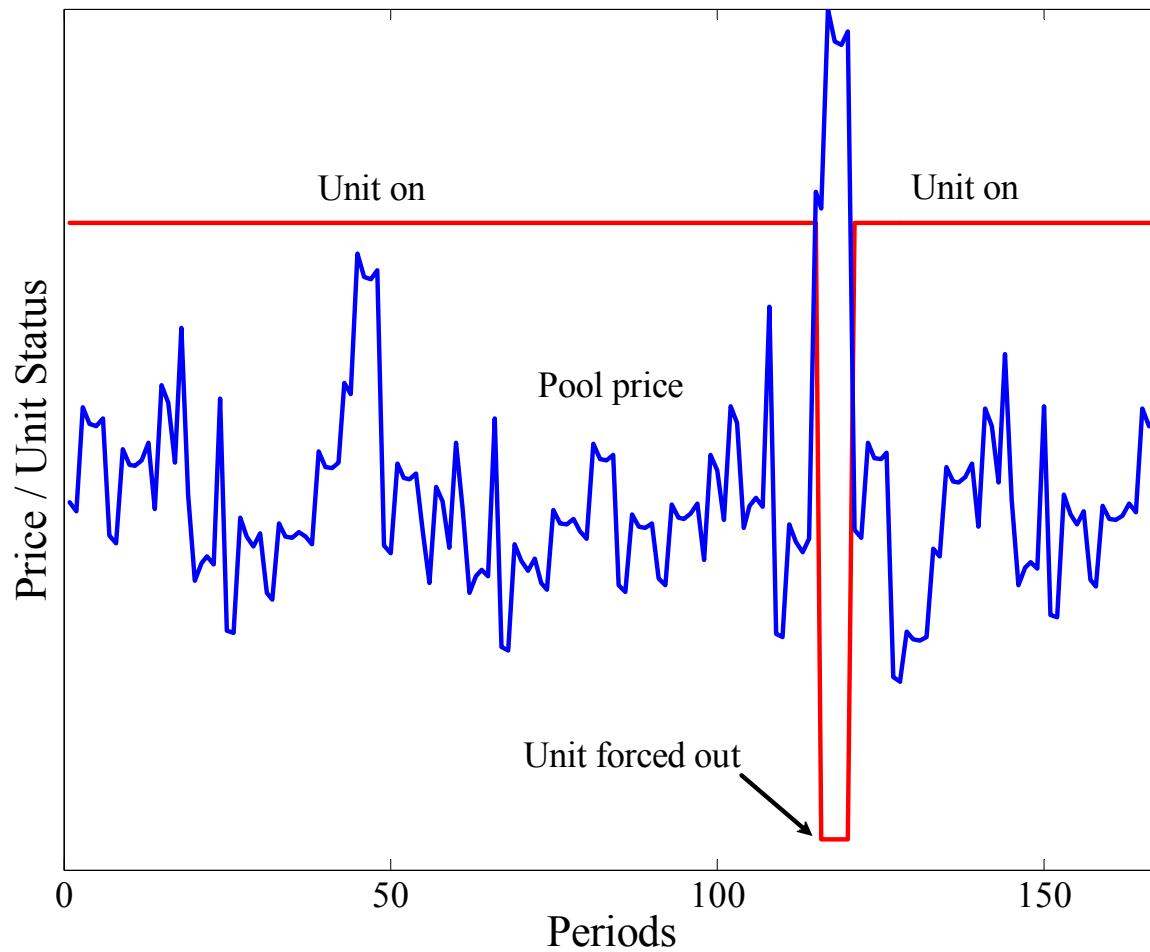
- High pool prices volatility
- Risk-averse producer
- Unit failures neglected



Problem: If the unit fails, the producer has to buy energy in the pool even if the price is high



Understanding the problem





Understanding the problem

- High pool prices volatility
- Risk-averse producer
- Unit failures considered



Expected profit ↓
Standard deviation of profit ↓



Aim

Analyze the effect of the unit failure on forward contracting decisions for different values of risk level



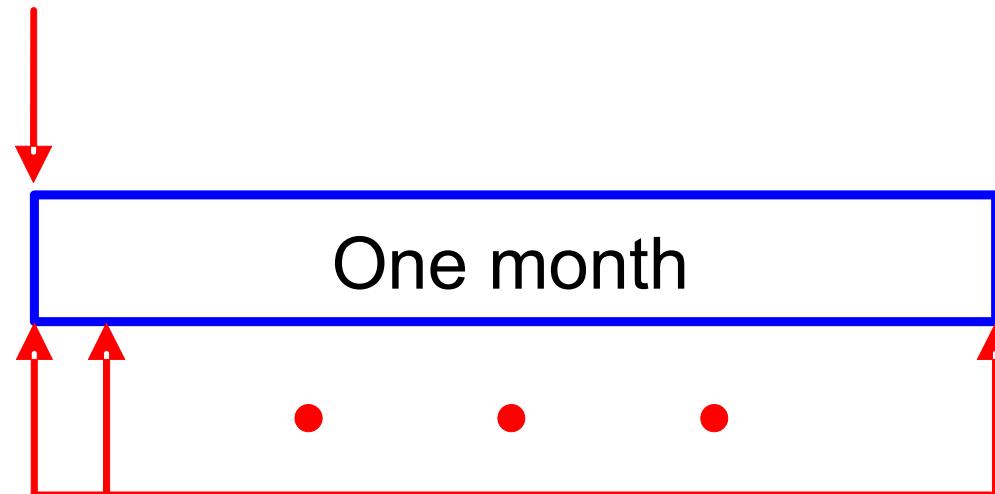
Decision Framework

- Planning horizon: e.g. , one year
- Two decisions:
 - Forward contracting at weekly / monthly /quarterly intervals
 - Pool trading throughout the planning horizon



Decision Framework

Forward contracting

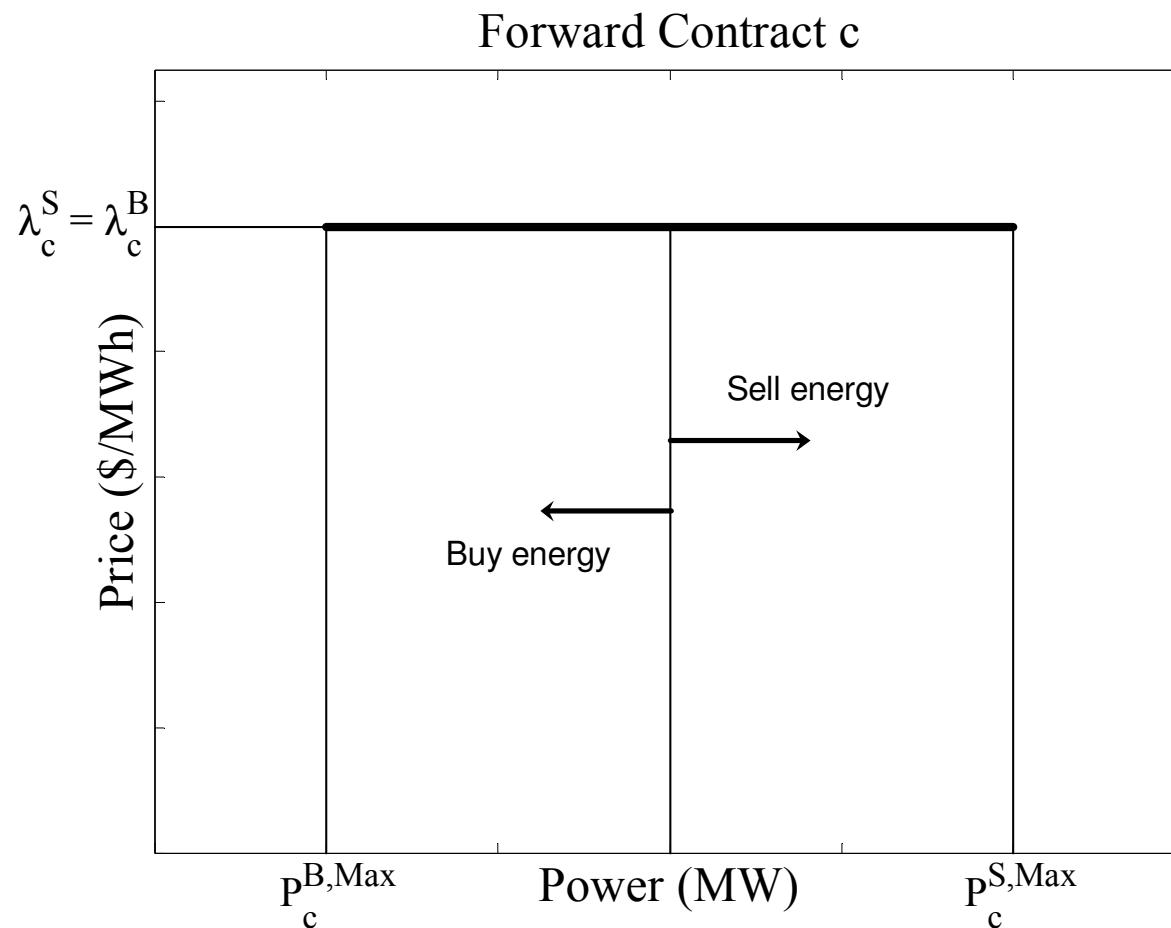


Pool trading



Forward Contracting

- Buy/sell a fixed MW power quantity (up to a maximum) during a certain future time period at a specified price





Forward Contracting

- Advantages:
 - Hedging against pool price volatility
 - Lower standard deviation of profit
- Disadvantages:
 - Lower expected profit
 - Impossibility of selling energy through forward contracts if the unit is forced out



Uncertainty

- Uncertainty of pool price
- Uncertainty of unit availability



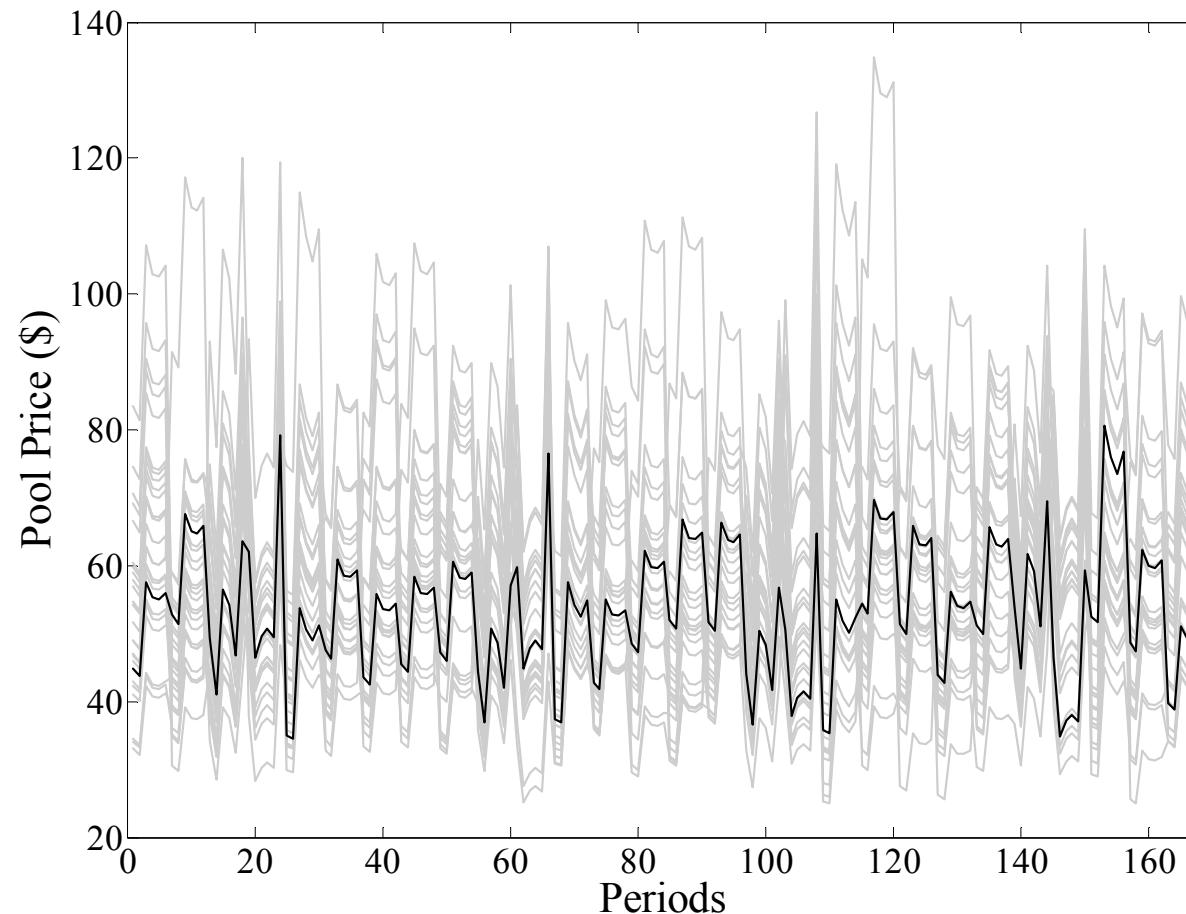
Modeling uncertainty

- Price uncertainty:
 - Scenario generation covering the study horizon
 - Scenario reduction
- Unit availability:
 - Scenario generation based on *MTTF* (Mean Time To Failure) and *MTTR* (Mean Time To Repair)
 - Scenario reduction



Characterizing price uncertainty

Pool price is treated as a stochastic variable:
Price scenario generation





Modeling unit availability

- Failure time modeled through an exponential distribution with mean $MTTF$
- Repair time modeled through an exponential distribution with mean $MTTR$

$$FOR (\%) = \frac{MTTR}{MTTF + MTTR} \times 100$$



Modeling unit availability

- We simulate *Up Time* and *Down Time* as:

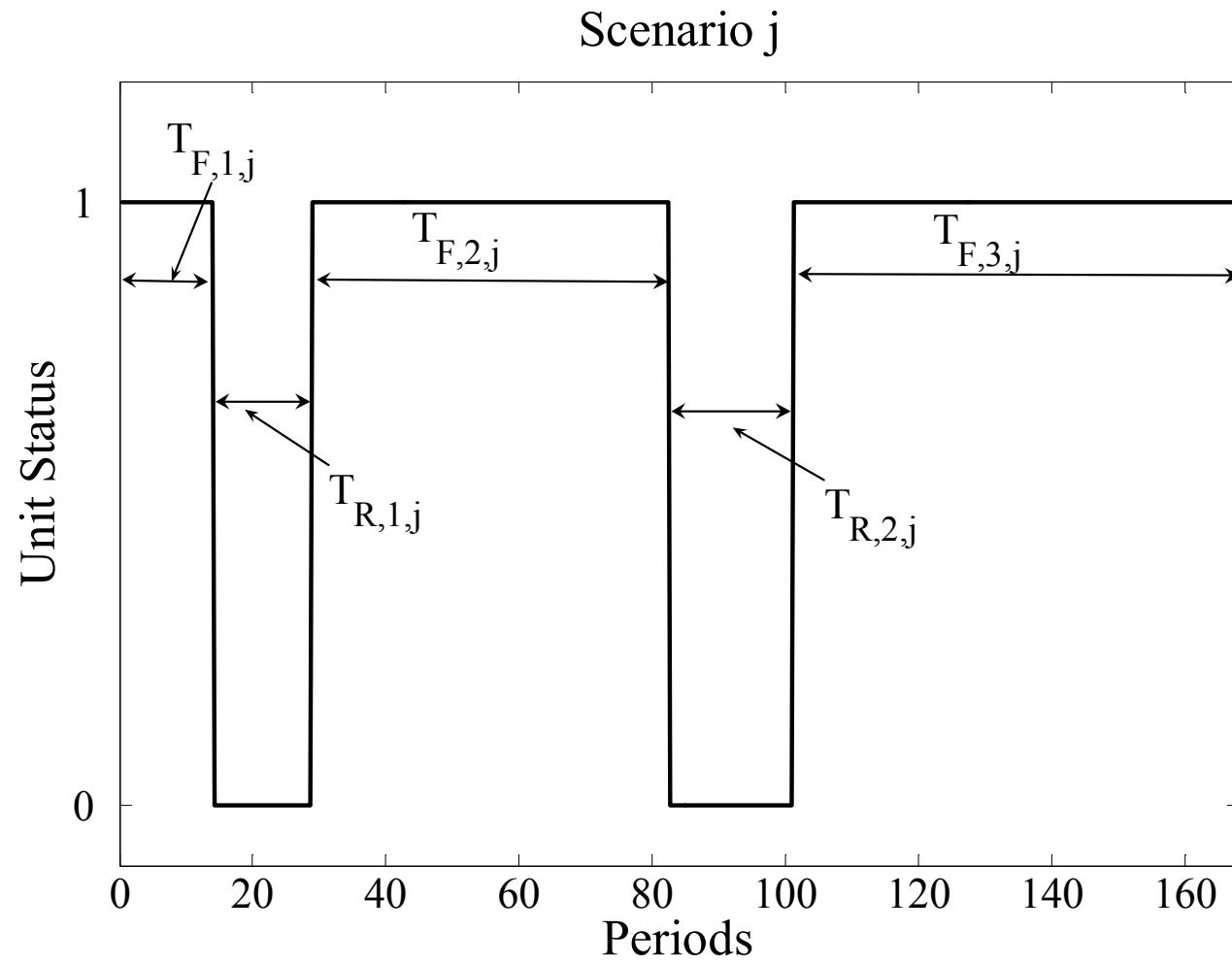
$$\text{Up Time} = -MTTF \times \log(u_1)$$

$$\text{Down Time} = -MTTR \times \log(u_2)$$

u_1 and u_2 are random variables uniformly distributed between 0 and 1



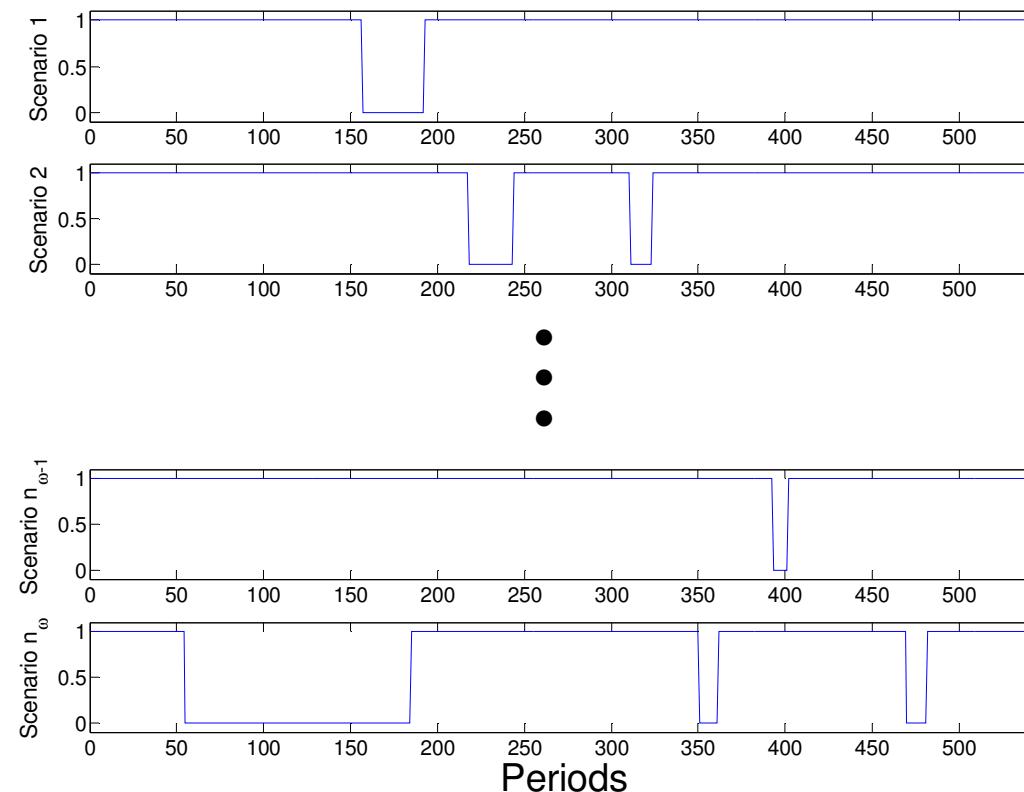
Modeling unit availability





Modeling unit availability

Status of the unit is treated as a stochastic variable:
Unit status scenario generation





Scenario reduction

- Needed to attain problem tractability
- Reduction technique based on the O.F. of a problem with one scenario (*Fast forward selection method*)
- Assessment via observation of the evolution of the objective function value with the # of scenarios



Model

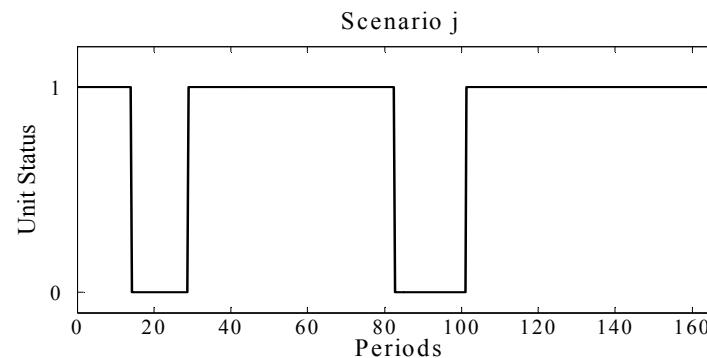
- Two-stage stochastic programming approach:
 - Stochastic variables: pool price and unit availability
 - First stage variables: forward contracting decisions
 - Second stage variables: pool trading



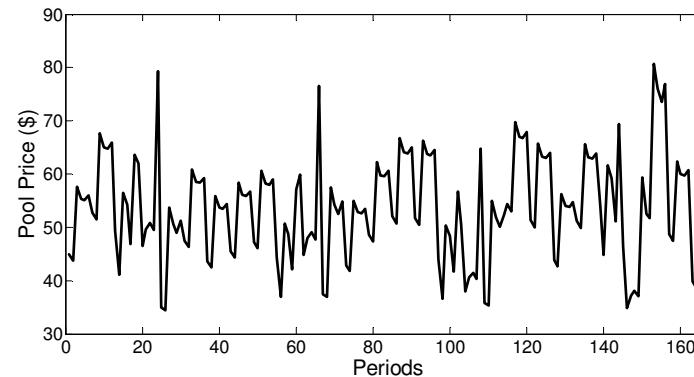
Model

- Each scenario of the optimization problem contains:

[1 0 0 1 0 1 1 1 ... 0 1 1]



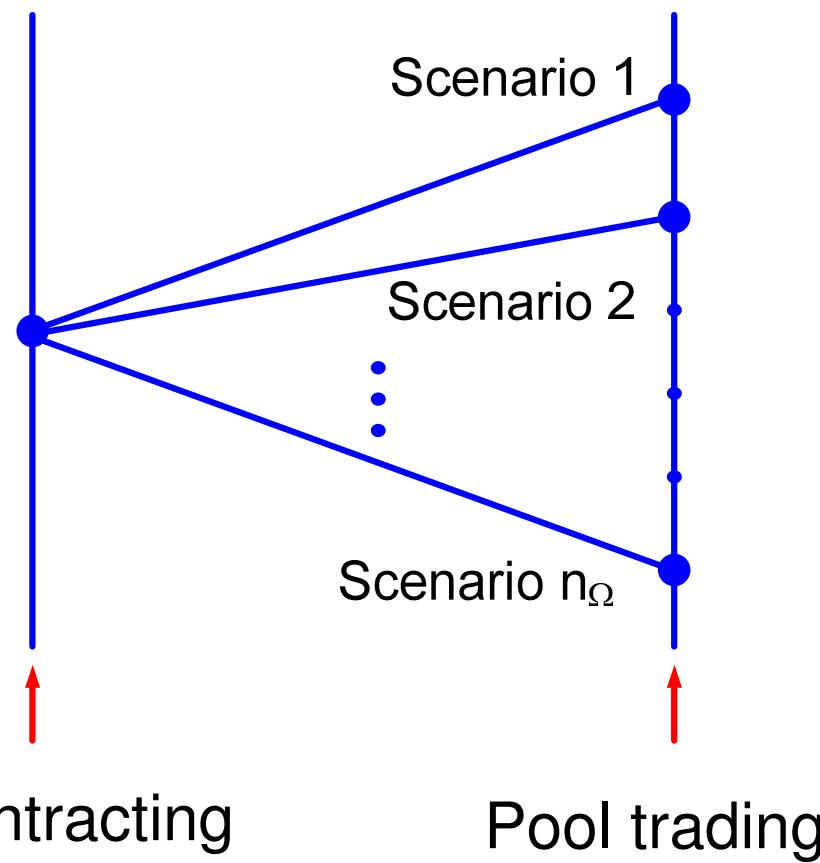
[24 37 41 21 39 52]





Stochastic programming approach

Two-stage stochastic programming: Scenario tree



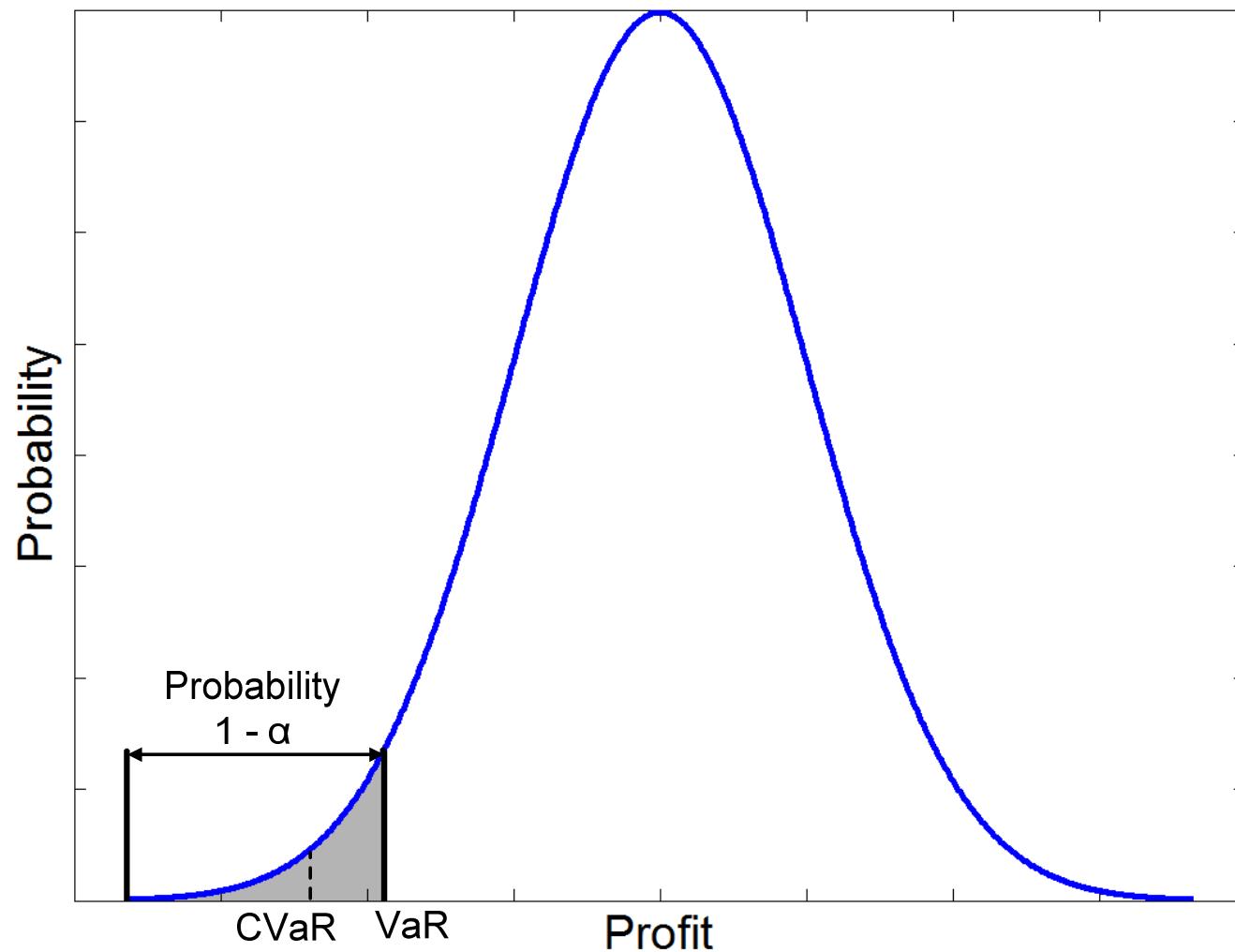


Risk management

- Tradeoff between expected profit and risk due to profit variability
- Risk measure: CVaR (average profit in scenarios with lowest profit)
- CVaR advantage: linear formulation



Risk management





Problem formulation

- Maximize $Expected\ profit + \beta \times CVaR$
- Subject to constraints associated with:
 - Energy balance
 - Forward contracts
 - CVaR

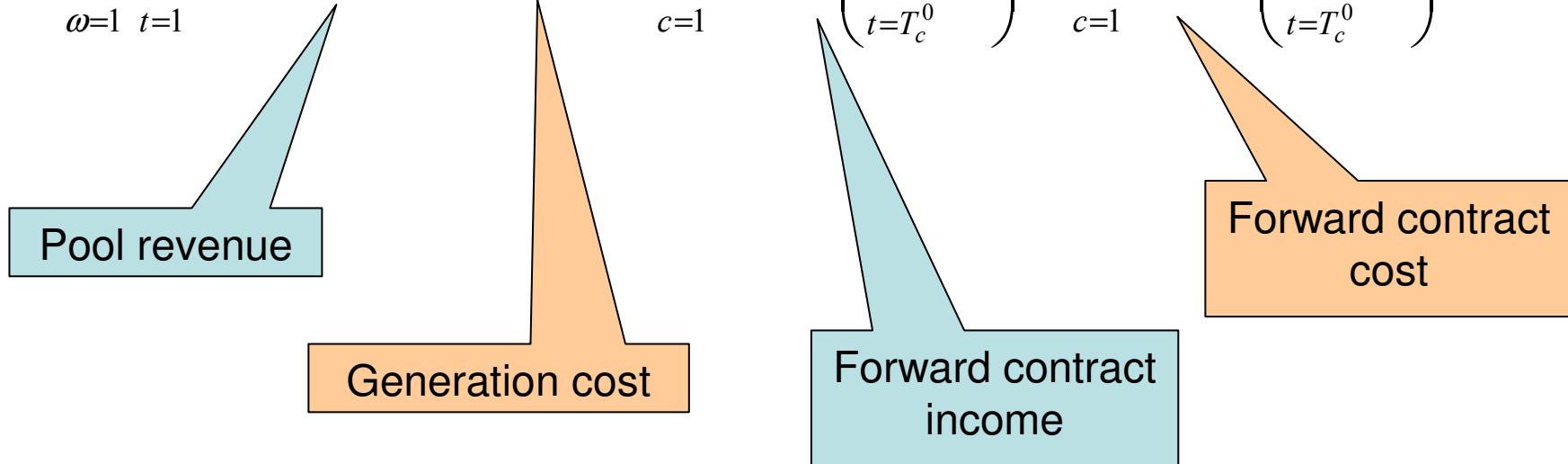
The tradeoff between expected profit and risk is enforced through the weighting factor $\beta \in [0, \infty)$



Expected profit

$E\{\text{profit}\} = E\{\text{revenue from pool involvement} - \text{generation cost}\} +$
 $+ \text{forward contract purchases} - \text{forward contract cost} =$

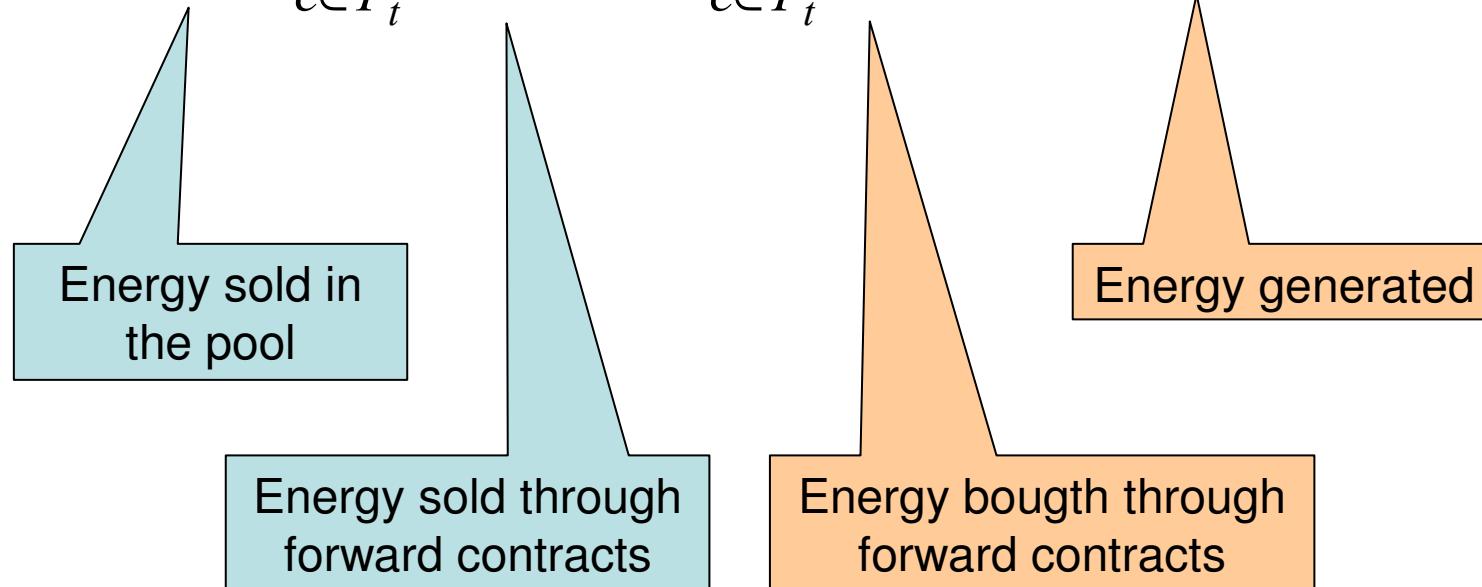
$$= \sum_{\omega=1}^{N_W} \sum_{t=1}^{N_T} \pi_{\omega} (\lambda_{t\omega}^P E_{t\omega}^P - E_{t\omega}^G C) + \sum_{c=1}^{N_C} \lambda_c^S P_c^S \left(\sum_{t=T_c^0}^{T_c} L_t \right) - \sum_{c=1}^{N_C} \lambda_c^B P_c^B \left(\sum_{t=T_c^0}^{T_c} L_t \right)$$





Energy balance

$$E_{t\omega}^P + \sum_{c \in F_t} P_c^S L_t = \sum_{c \in F_t} P_c^B L_t + E_{t\omega}^G \quad \forall t, \forall \omega$$





Bounds

$$E_{t\omega}^G \leq k_{tw} P^{\max} L_t$$

$$E_{t\omega}^G \geq k_{tw} P^{\min} L_t$$

Availability
Parameter

$$P_c^S, P_c^B \geq 0$$

$$P_c^B \leq P_c^{B,\max}$$

$$P_c^S \leq P_c^{S,\max}$$

$\forall t, \forall \omega$

$\forall t, \forall \omega$

$\forall c$

$\forall c$

$\forall c$



Risk measure (CVaR)

CVaR is incorporated in the objective function through a weighting parameter β

$$\begin{aligned} CVaR = & \xi - \frac{1}{1-\alpha} \sum_{\omega \in \Omega} \pi_\omega \eta_\omega \\ & - \sum_{t=1}^{N_T} \lambda_{t\omega}^P E_{t\omega}^P - \sum_{c=1}^{N_C} \lambda_c^S P_c^S \left(\sum_{t=T_c^0}^{T_c} L_t \right) + \sum_{c=1}^{N_C} \lambda_c^B P_c^B \left(\sum_{t=T_c^0}^{T_c} L_t \right) + \\ & + \sum_{t=1}^{N_T} E_{t\omega}^G C + \xi - \eta_\omega \leq 0 \quad \forall \omega \\ \eta_\omega \geq 0 \end{aligned}$$

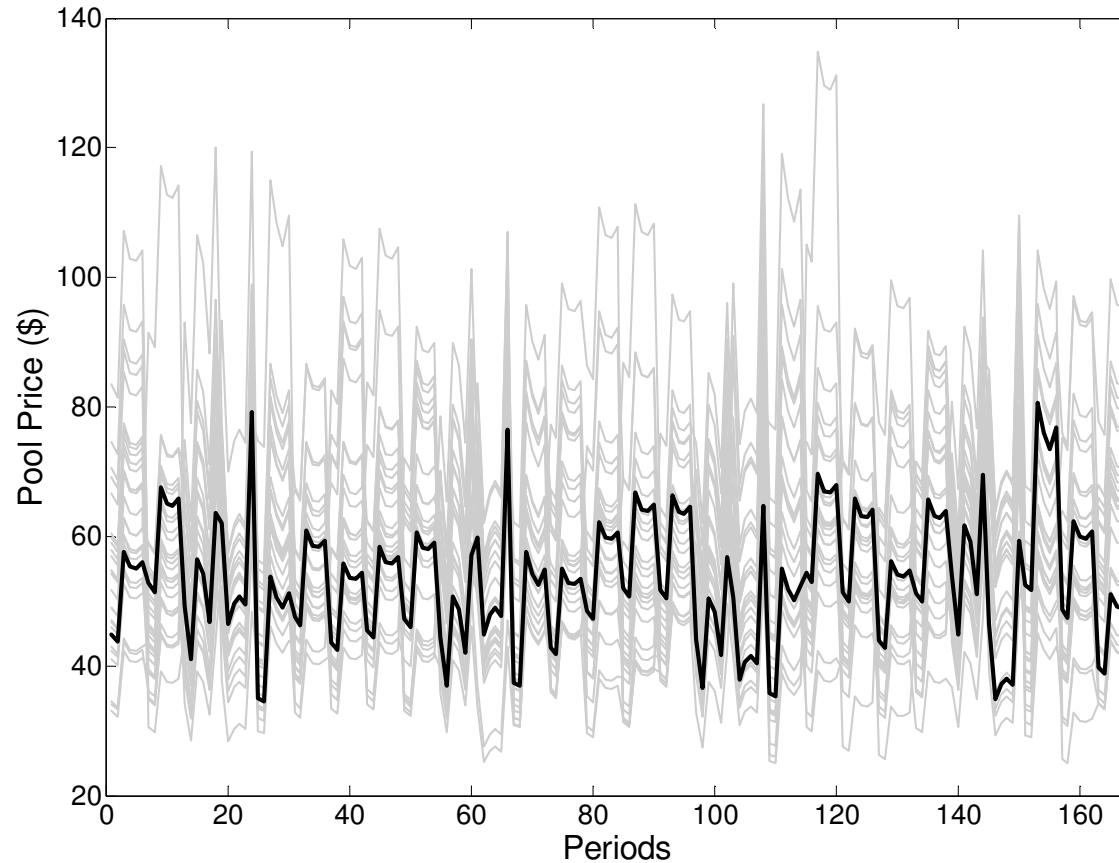


Numerical simulations. Stochastic Variable Characterization

- One month (28 days)
- Each day is divided into 6 periods of 4 hours each
- Total number of periods: $28 \times 6 = 168$



Numerical simulations. Prices scenarios



Original tree: 288 price scenarios → Final tree: 25 price scenarios

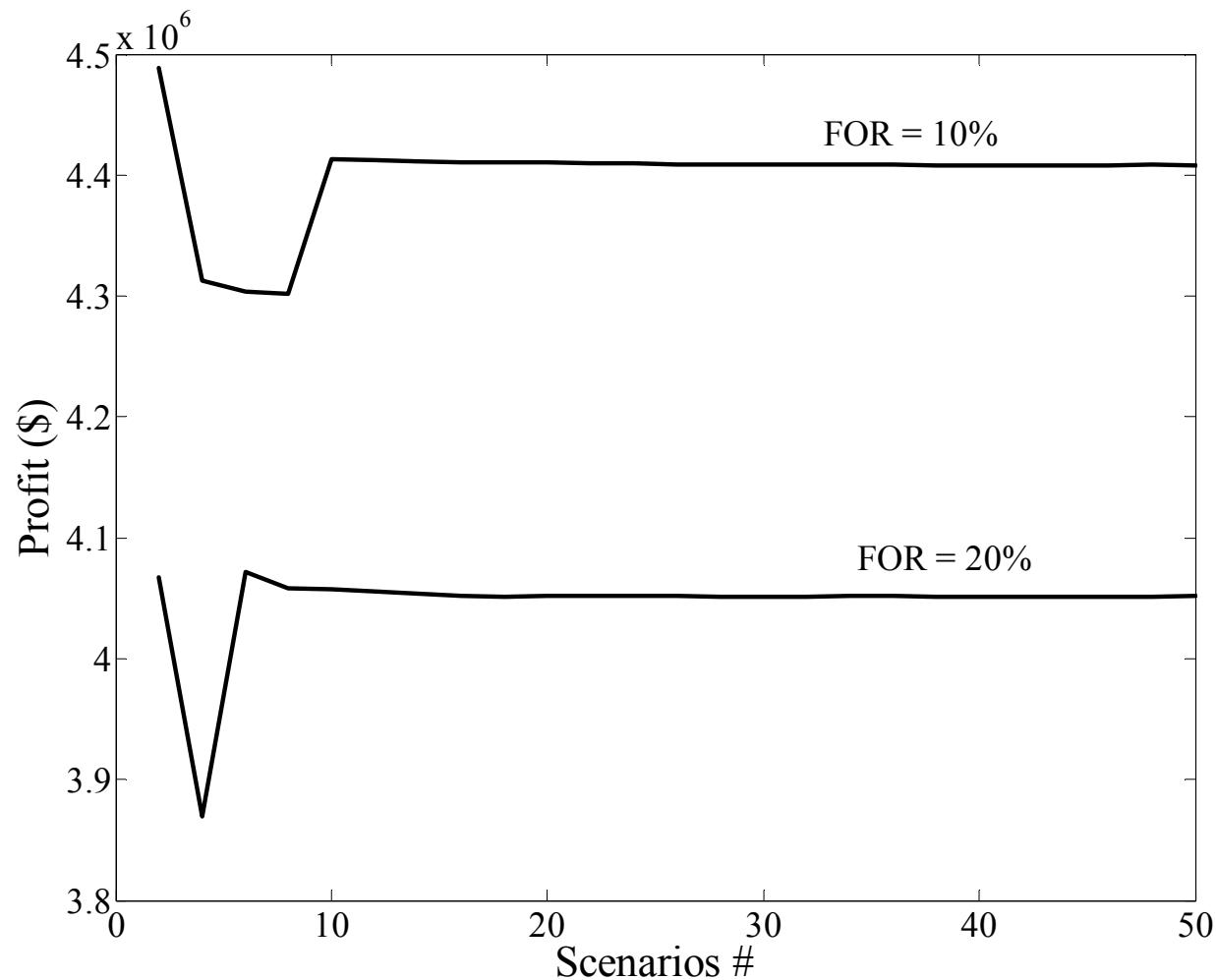


Numerical simulations. Unit Status

- Simulation for three different cases:
 - Case 1: $FOR = 0\%$ (no failure)
 - Case 2: $FOR = 10\%$ ($MTTF=400$ $MTTR=40$)
 - Case 3: $FOR = 20\%$ ($MTTF=250$ $MTTR=60$)
- Original tree: 5000 availability scenarios →
- Final tree: 50 availability scenarios



Numerical simulations. Reducing availability scenarios





Numerical simulations. Forward contracts

- One month decision framework
- Five forward contracts:
 - Fixed price for each contract
 - Maximum power that can be sold/bought
 - 4 contracts with a duration of 1 week
 - 1 contract with a duration of the whole month



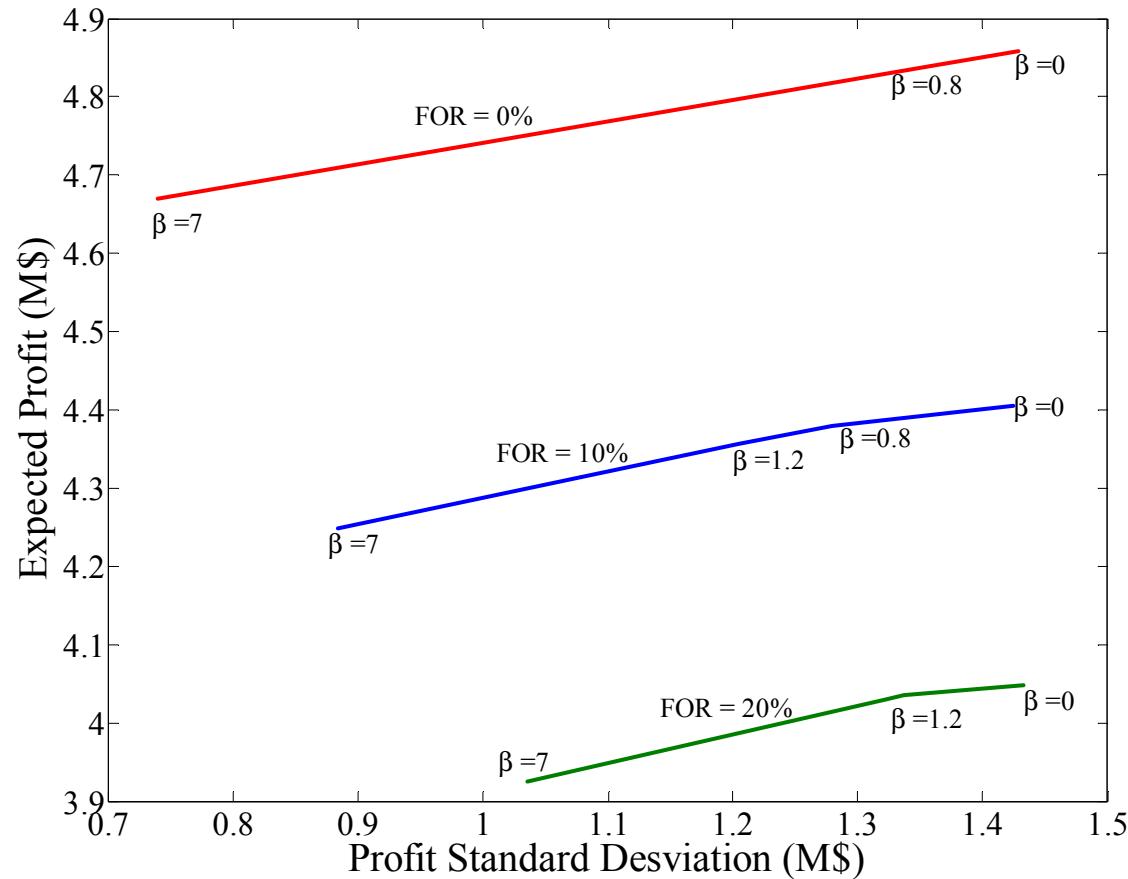
Numerical simulations. Forward contracts

# contract	$P_c^{S,Max}$	$P_c^{B,Max}$	λ_c^B	λ_c^S
1	100	100	52.88	52.88
2	100	100	50.75	50.75
3	100	100	53.60	53.60
4	100	100	53.25	53.25
5	50	50	52.57	52.57

Week 1	Week 2	Week 3	Week 4
Contract 1			
	Contract 2		
		Contract 3	
			Contract 4
Contract 5			



Results

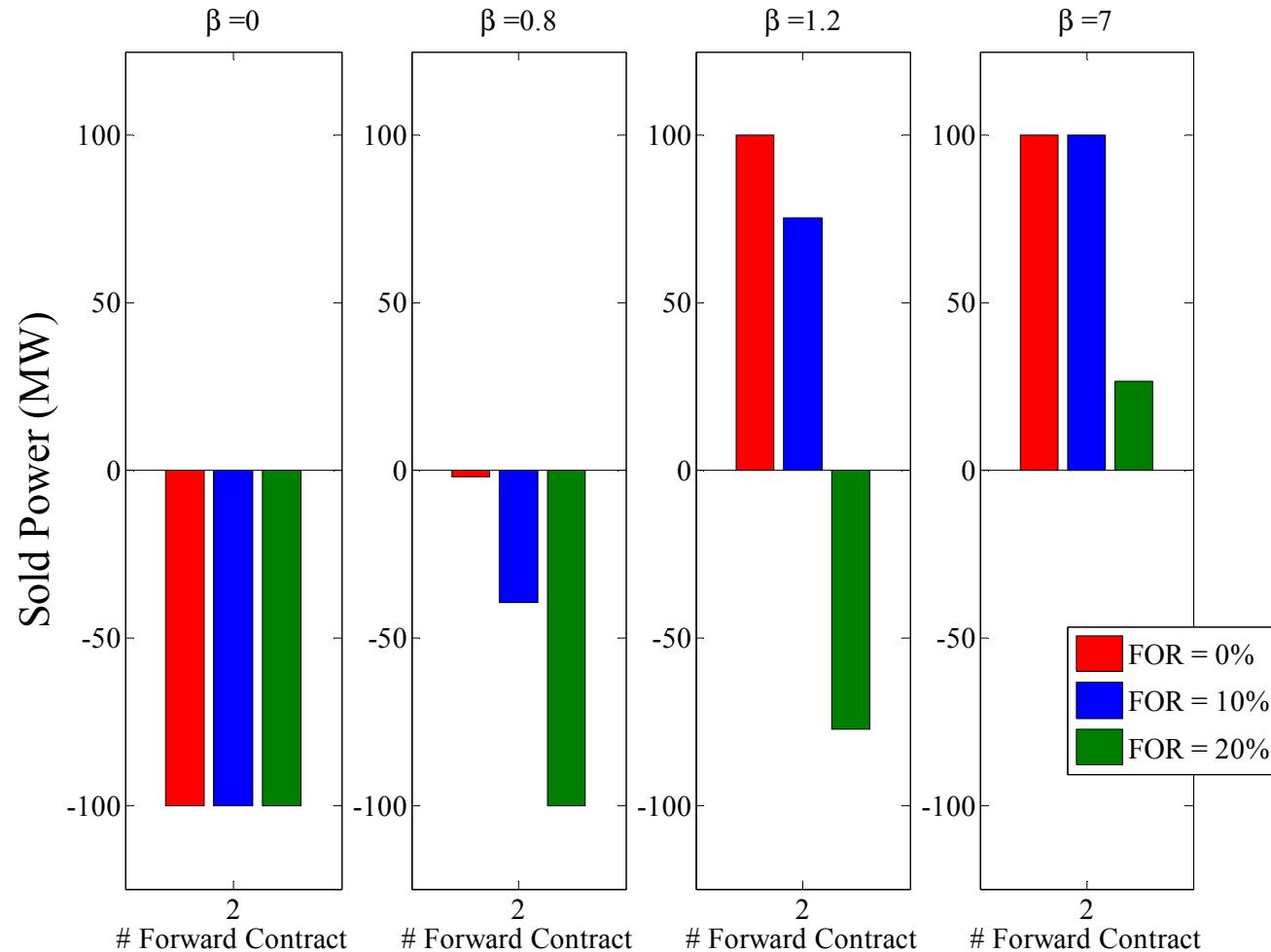


$\beta \uparrow \Rightarrow \begin{cases} \text{expected profit } \downarrow \\ \text{profit standard desviation } \downarrow \end{cases}$

$FOR \uparrow \Rightarrow \text{expected profit } \downarrow \downarrow$



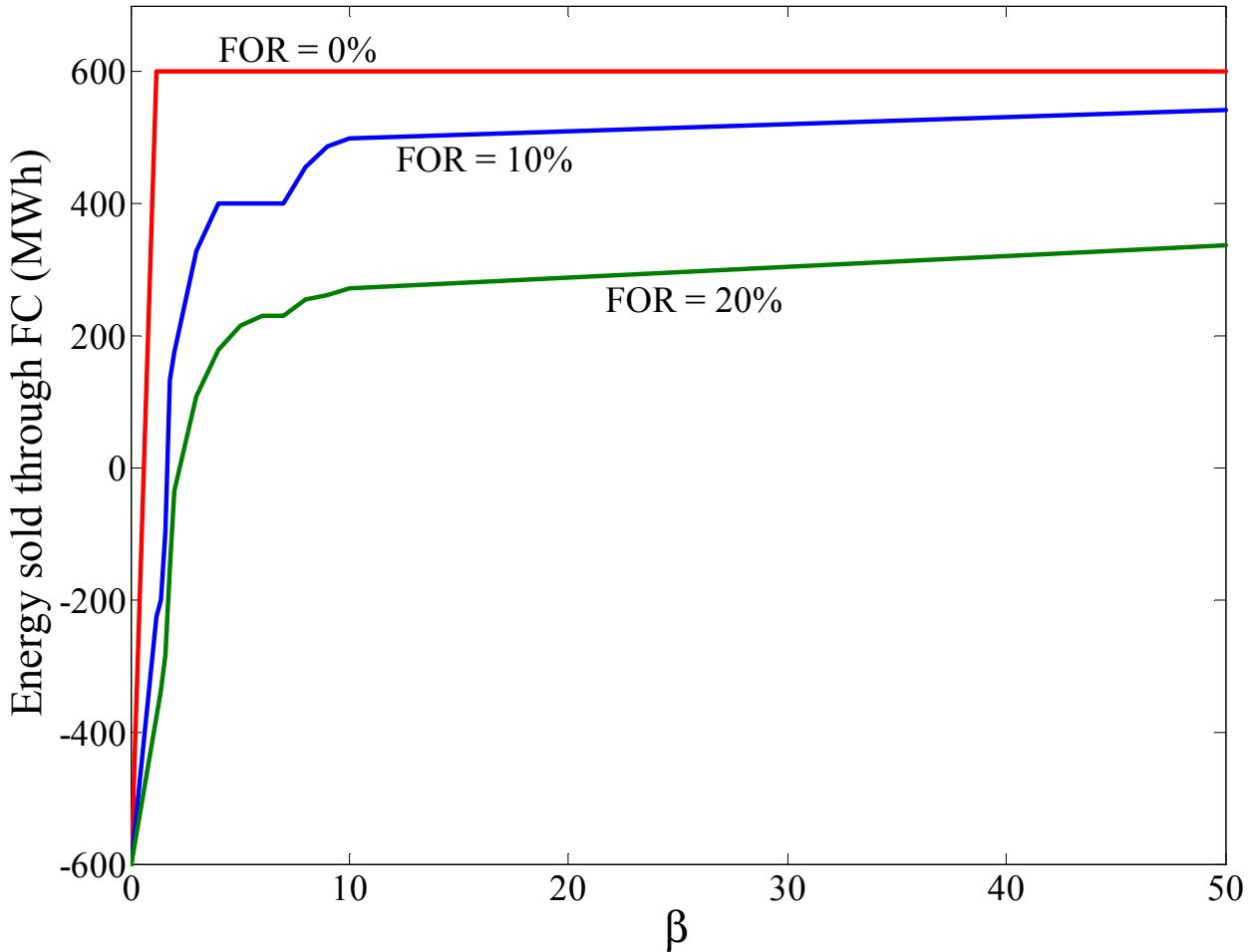
Results



The higher the FOR, the lower the energy sold through FC



Results



The producer sells less energy through forward contracts if the possibility of a unit failure increases



Conclusion

- The energy sold through forward contracts by a unit decreases if its FOR increases



Simulation Data

- GAMS: cplex solver
- 2,6 GHz
- Linear optimization problem with continuous variables
- $4.2 \cdot 10^5$ variables
- $6.3 \cdot 10^5$ constraints
- Solution time: 3 minutes



Future work

- Forward contracts with increasing prices
- Include financial options
- Possibility of selling to end users directly (“gentailer”)
- Analyze different scenario reduction techniques



Thanks for your attention!

GSEE: <http://www.uclm.es/area/gsee/>