



A COHERENT RISK MEASURE APPLIED TO THE BRAZILIAN LONG TERM HYDROTHERMAL SCHEDULING

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Joint work with
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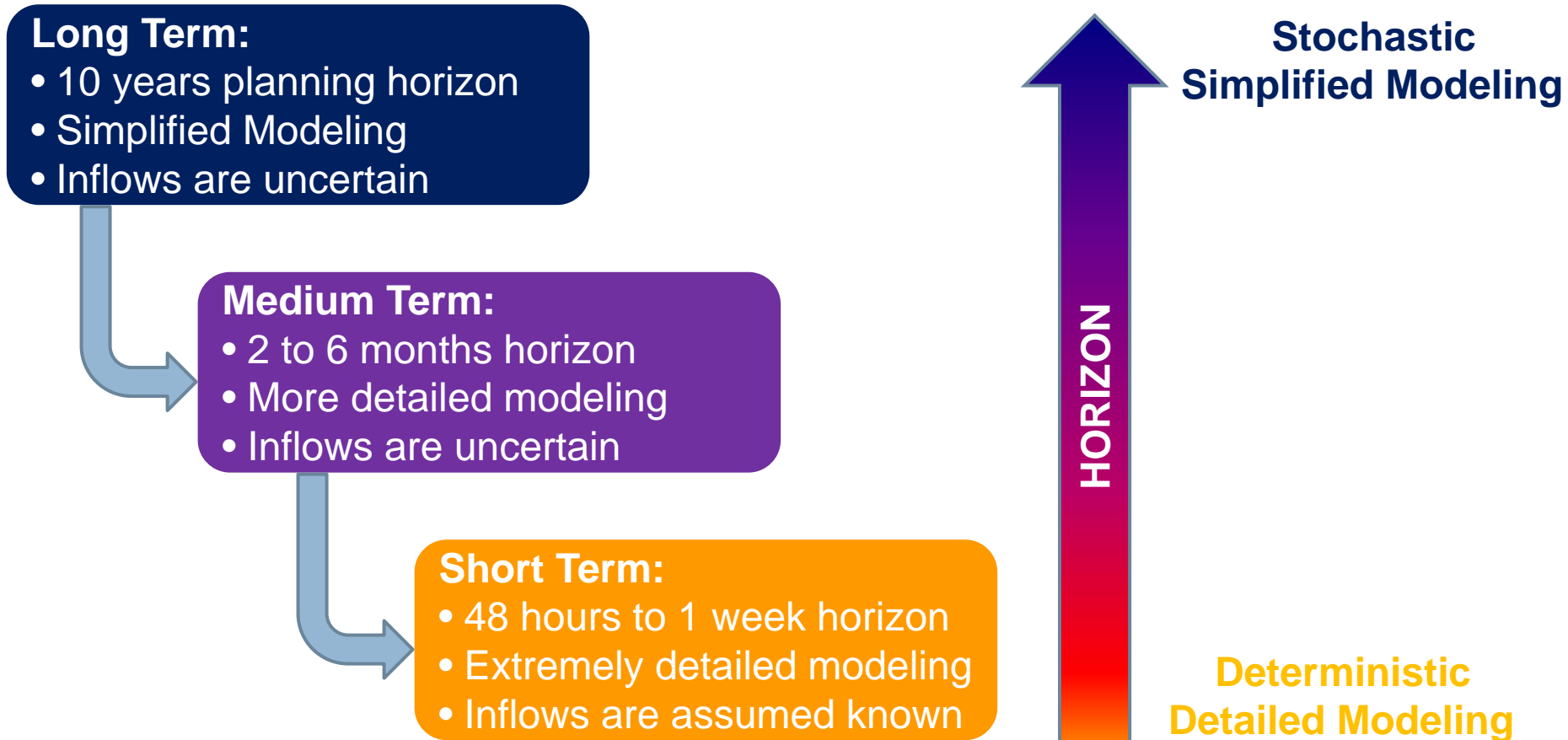
EPOC Workshop - Auckland, NZ

Agenda



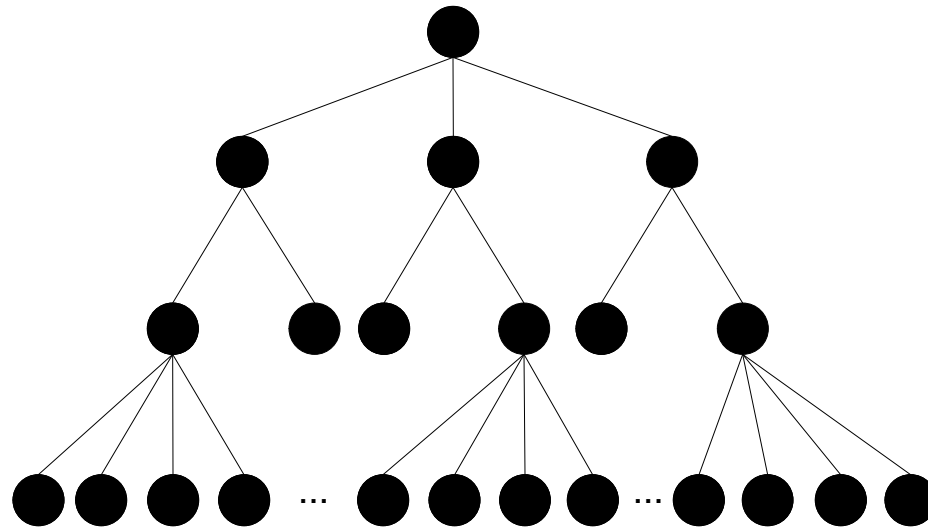
- Hydrothermal Scheduling in Brazil
- Risk Aversion
- Coherent Risk Measure
- Risk Averse SDDP Implementation
- Computational Results
- Final Remarks

Hydrothermal Scheduling in Brazil



Hydrothermal Scheduling in Brazil

- Minimize the Expected Operation Cost



Risk Aversion

- Expectation = Risk Neutral
 - ▣ Provides the best decision on average
- In a pure economical model it is acceptable to load shedding in order to obtain the smallest cost
- However, load shedding is an issue in political, economical and social terms
 - ▣ Therefore, one is Risk Averse

Coherent Risk Measure

- Proposed by Shapiro (2010)
- Convex combination between Expectation and CVaR

$$\rho(Z) = (1 - \lambda)\mathbb{E}[Z] + \lambda CVaR_{1-\alpha}[Z]$$

- Artzner *et al* (1999)
 - Convexity
 - Monotonicity
 - Positive homogeneity
 - Translation equivariance

CVaR Definition

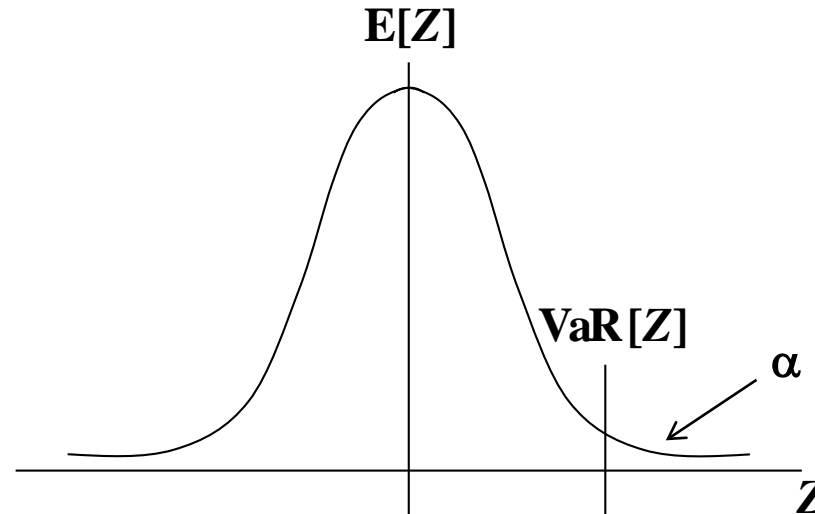
□ *Conditional Value-at-Risk - CVaR*

$$\text{CVaR}_{1-\alpha}[Z] = \inf_u \{u + \alpha^{-1} \mathbb{E}[Z - u]_+\},$$

CVaR Definition

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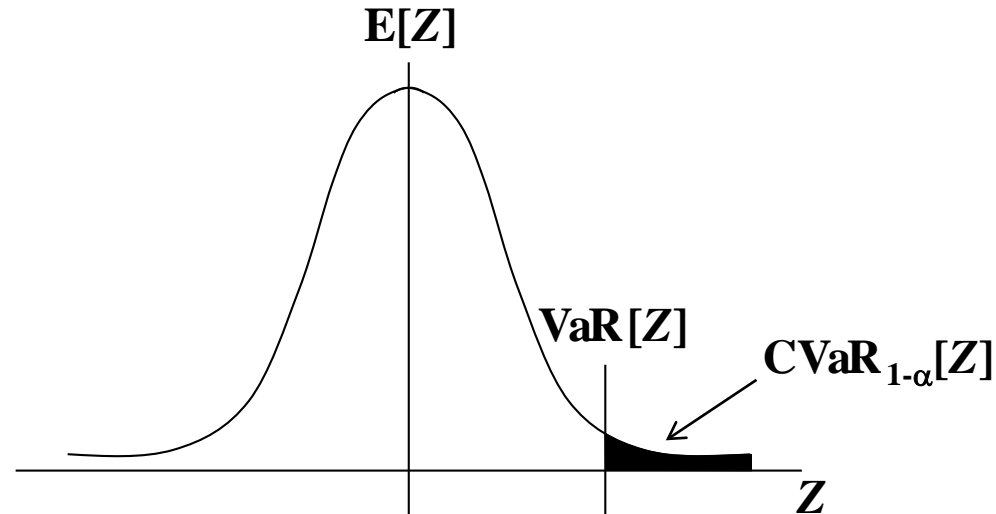
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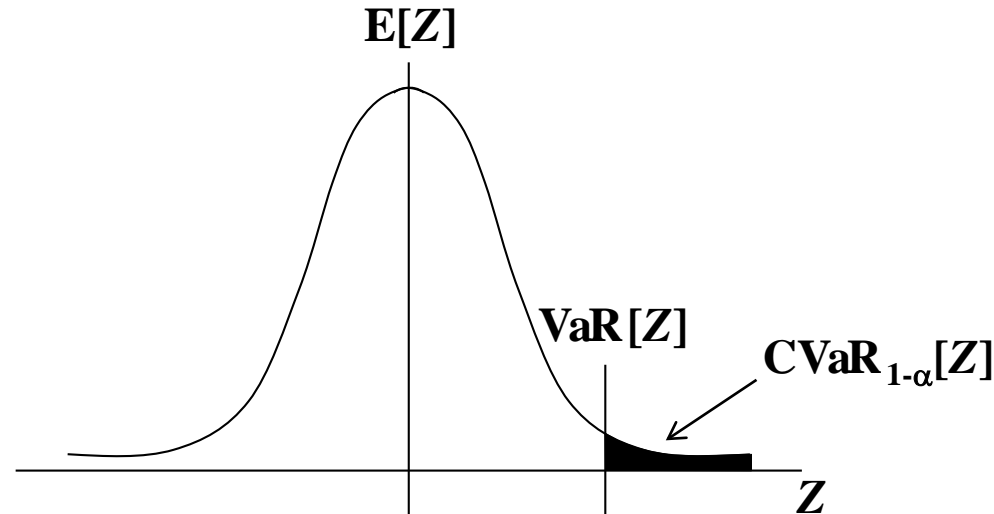
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Coherent Risk Measure

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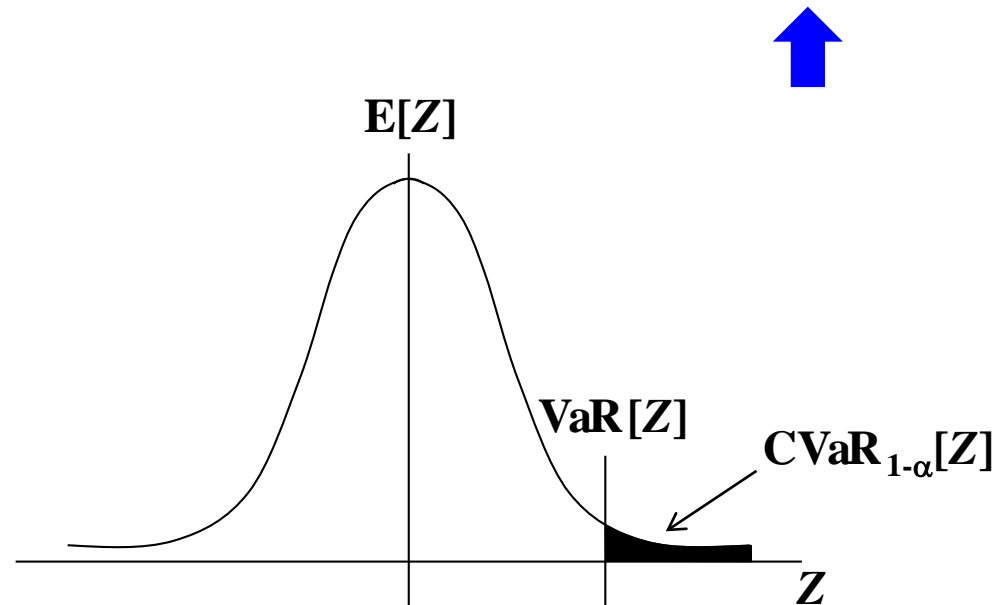
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Coherent Risk Measure

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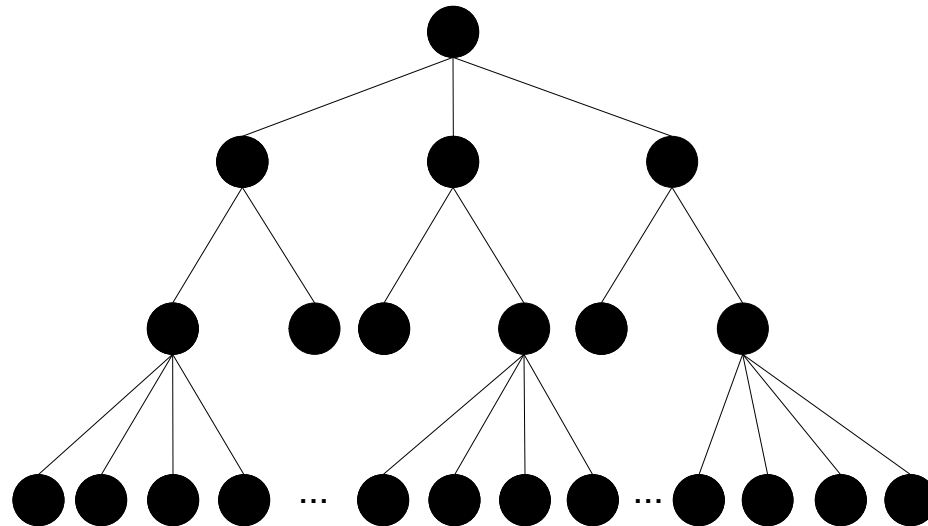
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Risk Averse SDDP Implementation

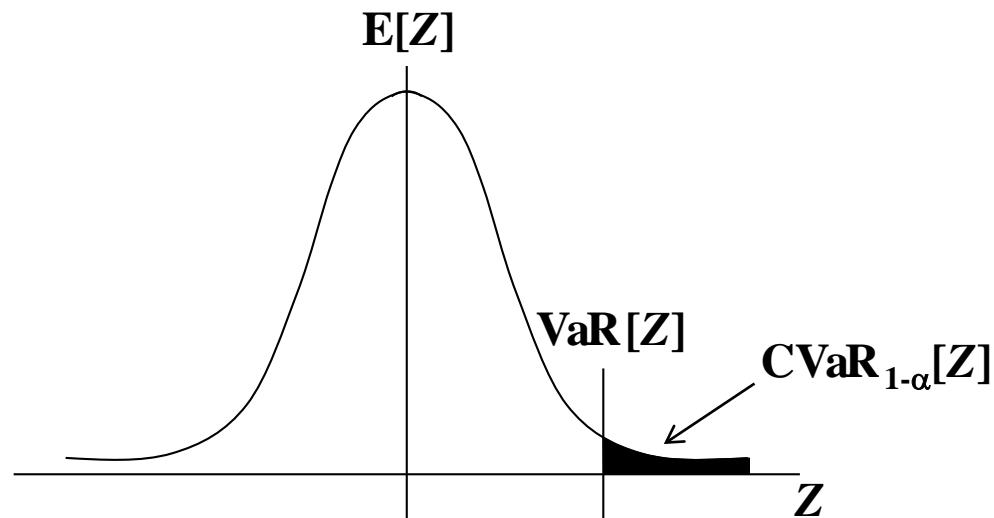
- Can be applied directly to the SDDP
 - ▣ One additional state variable - Philpott and de Matos (2011)

$$\rho(Z) = (1 - \lambda)\mathbb{E}[Z] + \lambda CVaR_{1-\alpha}[Z]$$



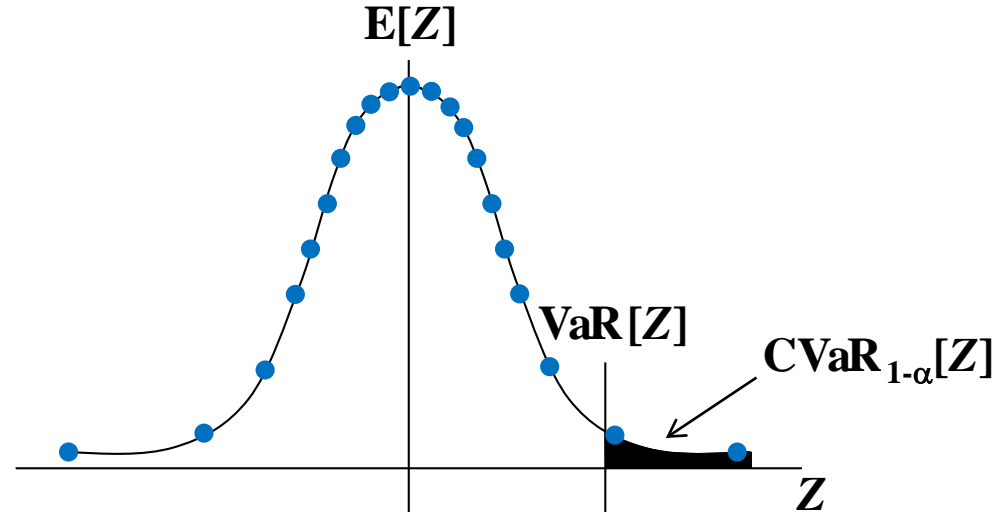
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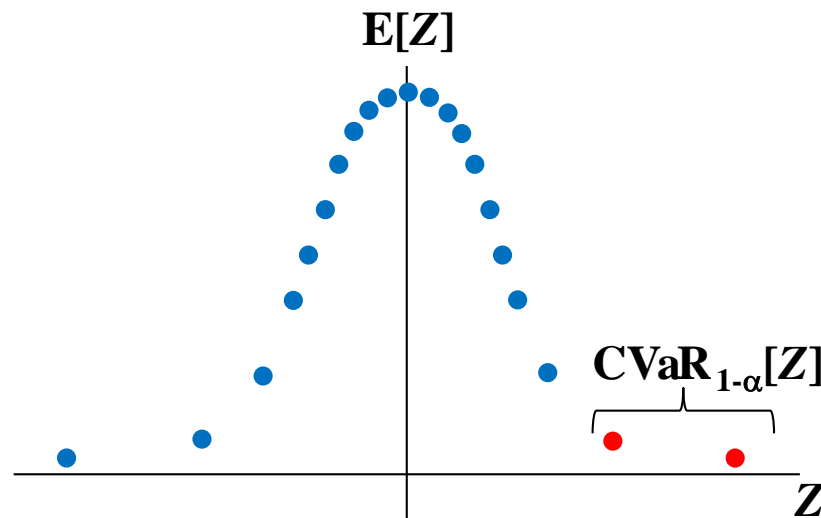
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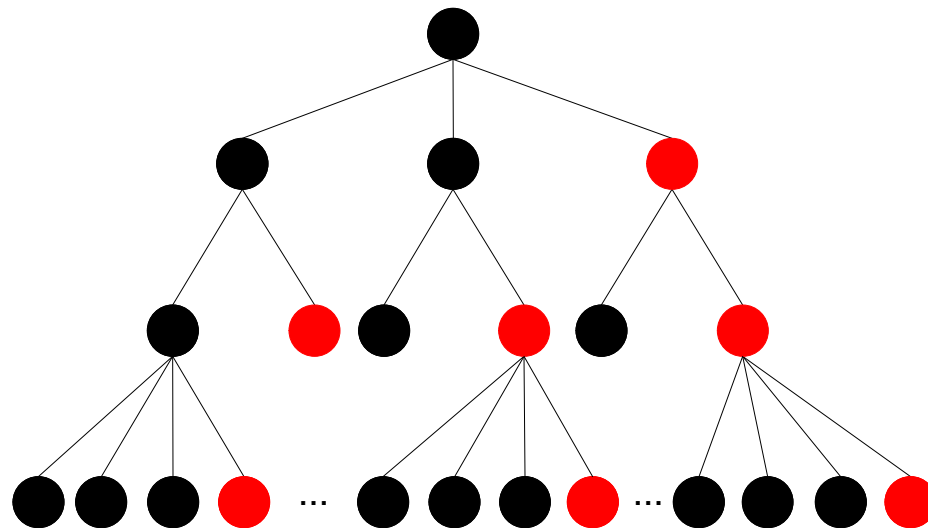
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Computational Results

- Brazilian Power System
 - 158 Hydro Plants
 - 151 Thermal Plants
- 5 Nodes (Buses)
- Hydro Plants
 - Aggregated in 4 EERs
- Data from January/2012



Computational Results

- Operation Policy computed for following parameters
 - ▣ 12,000 Cuts → Simulated over 5,000 scenarios
 - Scenarios were sampled within the Scenario Tree

$$\rho(Z) = (1 - \lambda)\mathbb{E}[Z] + \lambda CVaR_{1-\alpha}[Z]$$

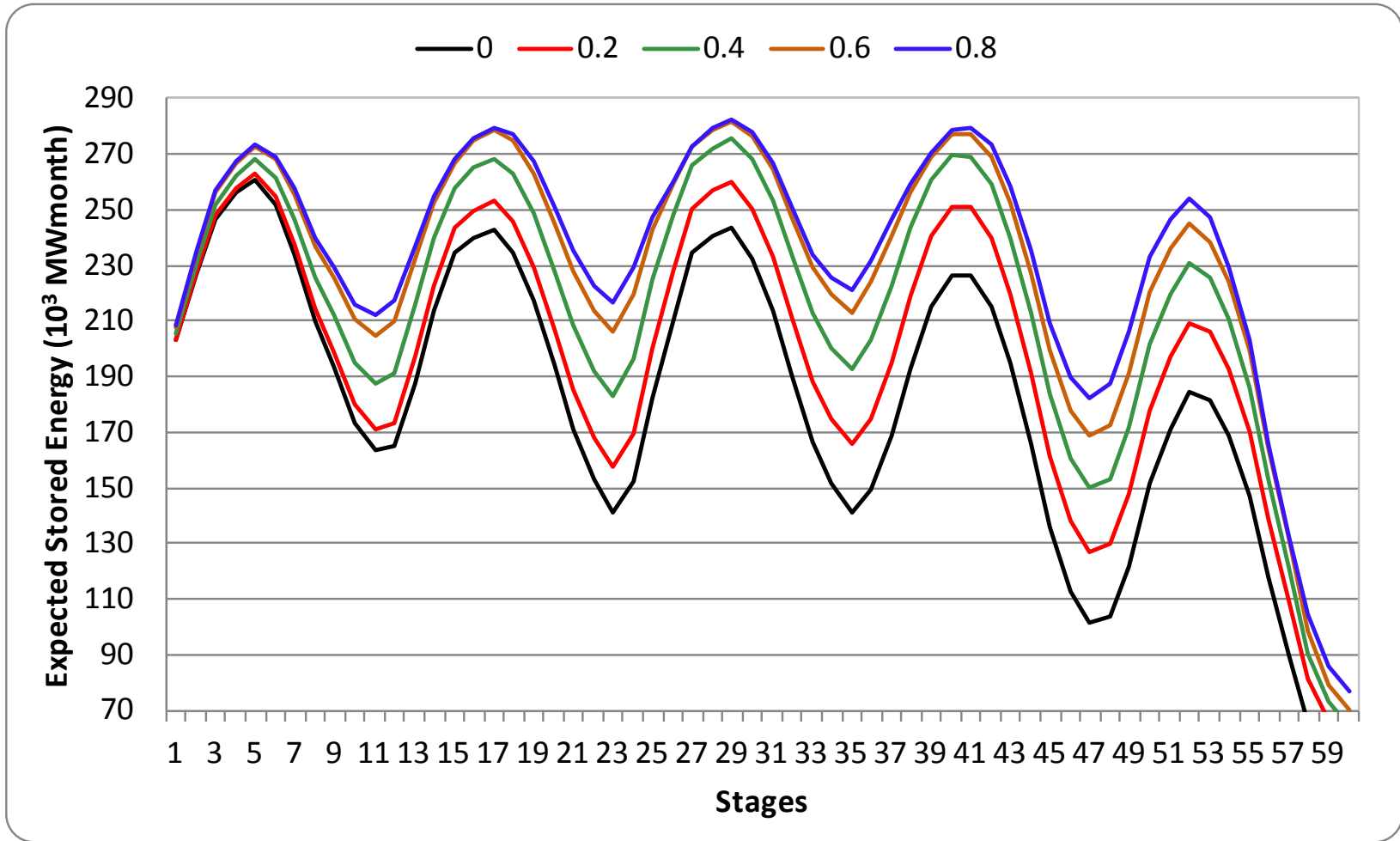
λ	α
0	10
0.2	50
0.4	90
0.6	
0.8	

Computational Results - $\alpha = 10\%$

Expected Stored Energy



LabPlan

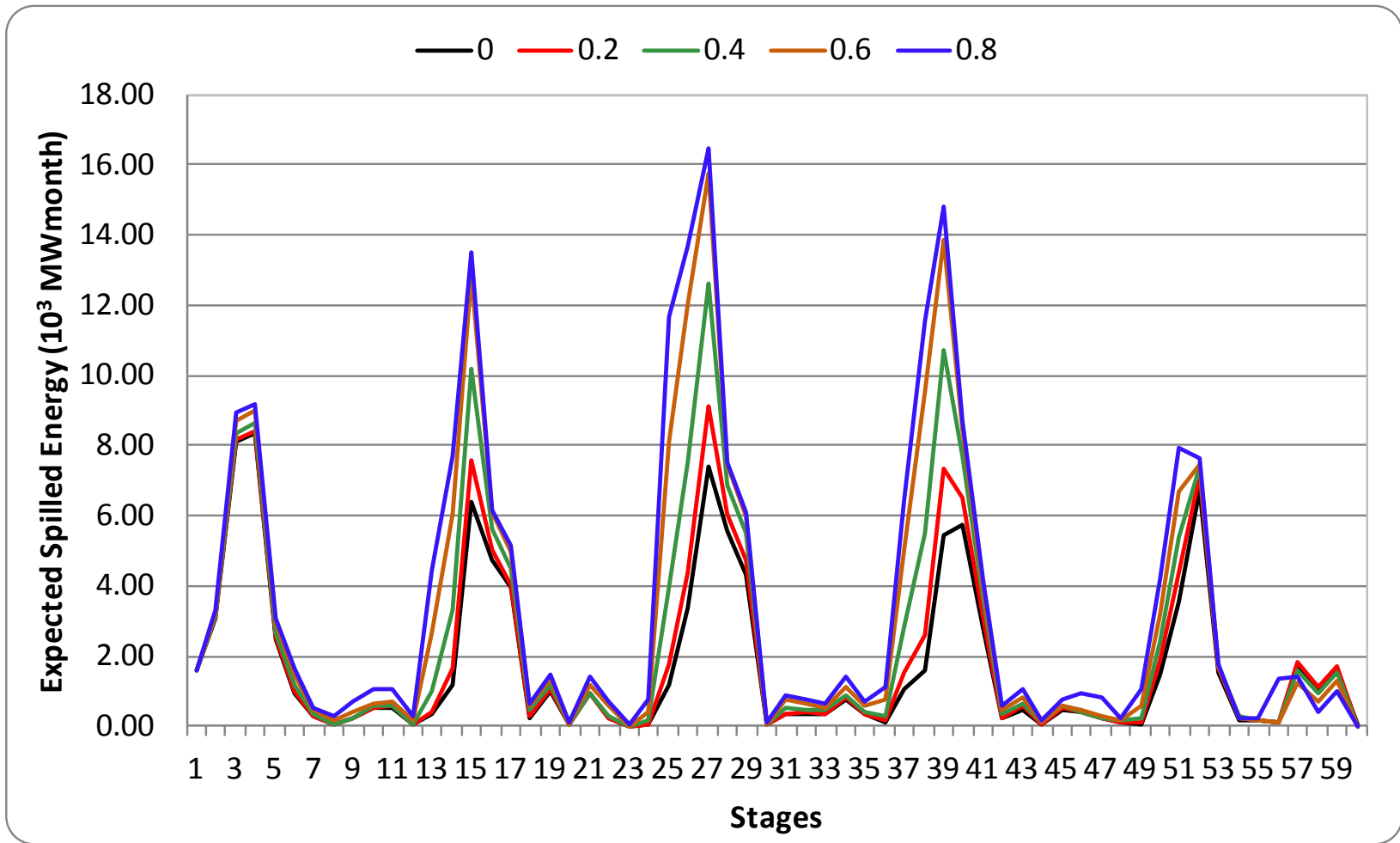


Computational Results - $\alpha = 10\%$

Expected Spilled Energy

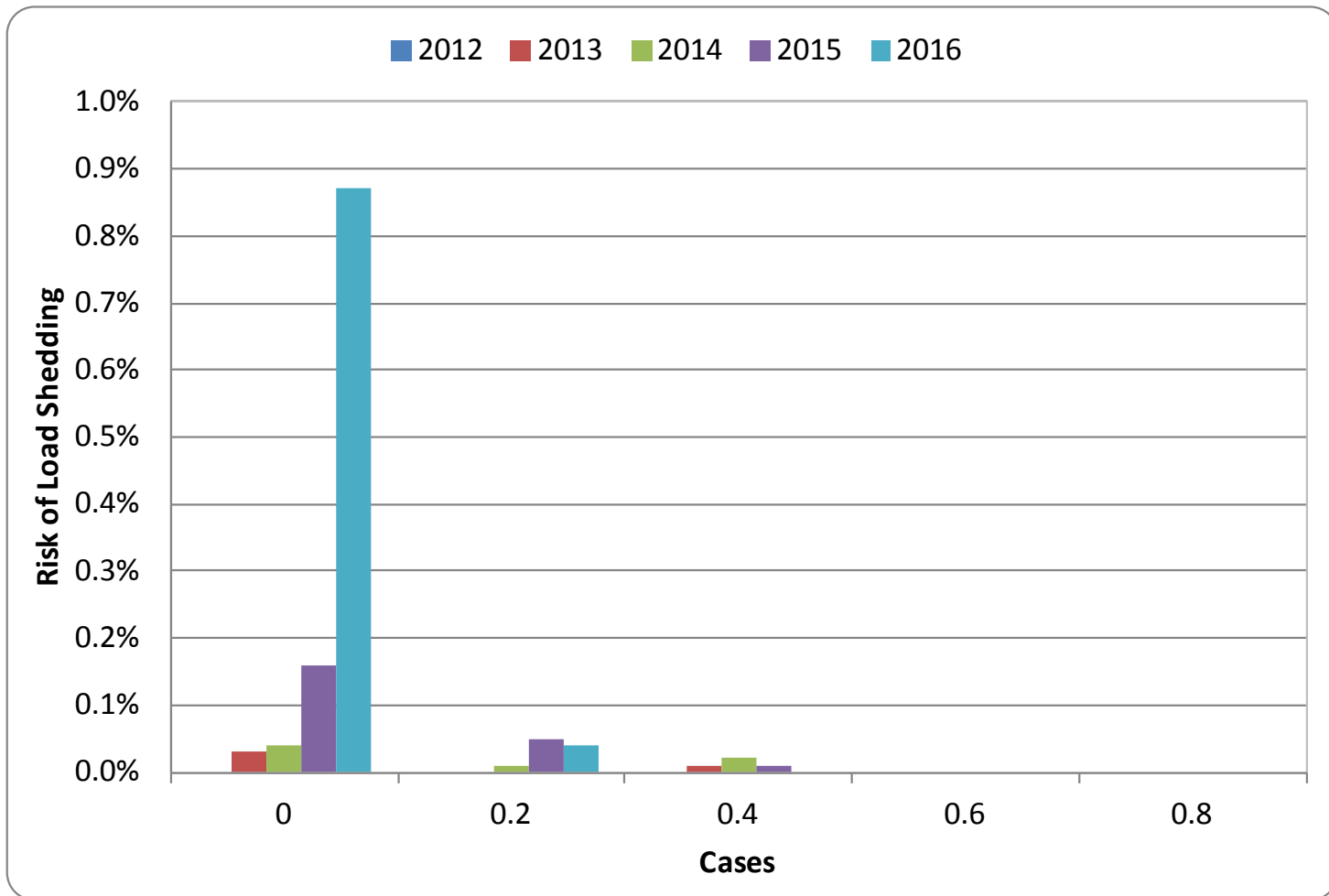


LabPlan



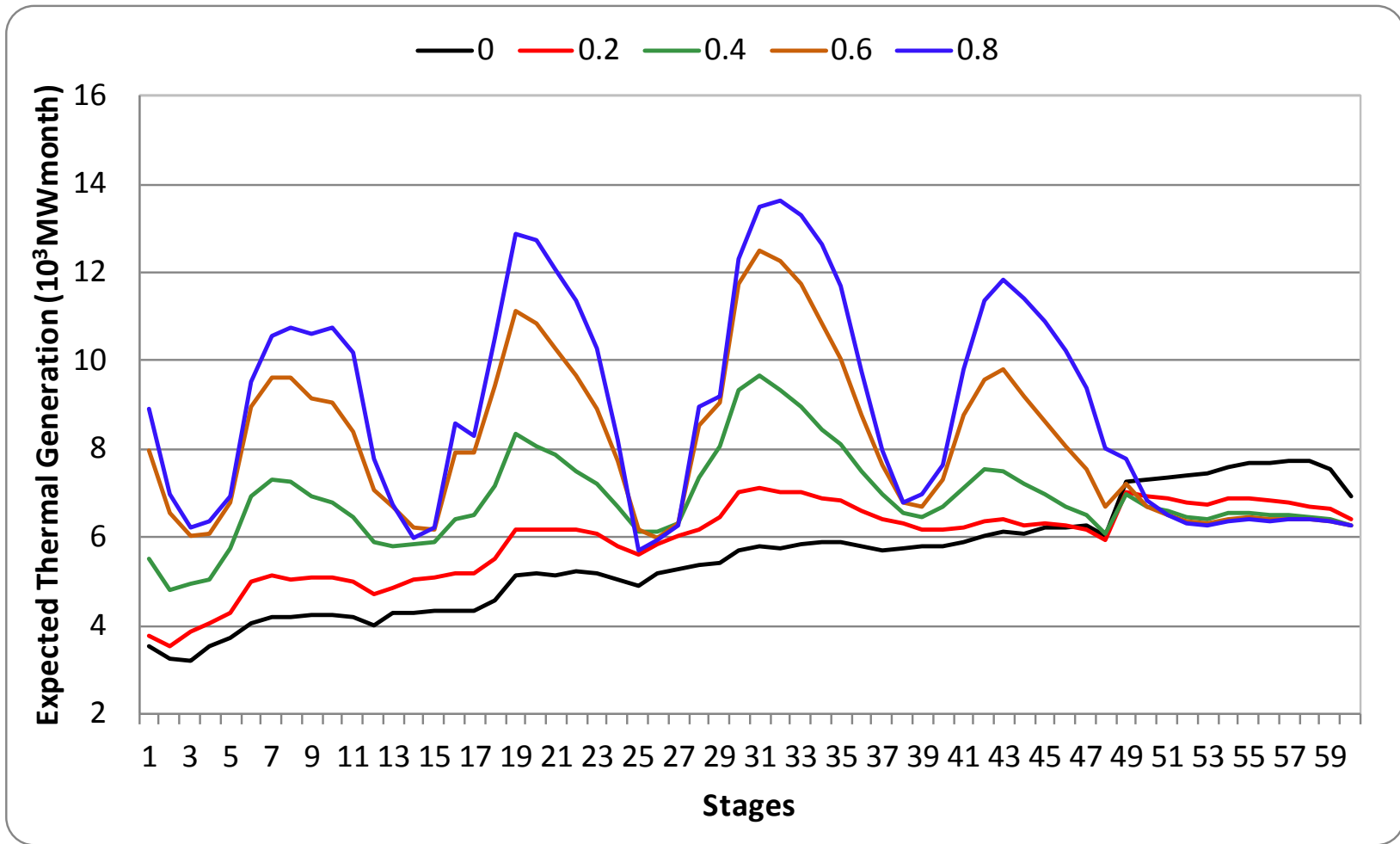
Computational Results - $\alpha = 10\%$

Risk of Load Shedding



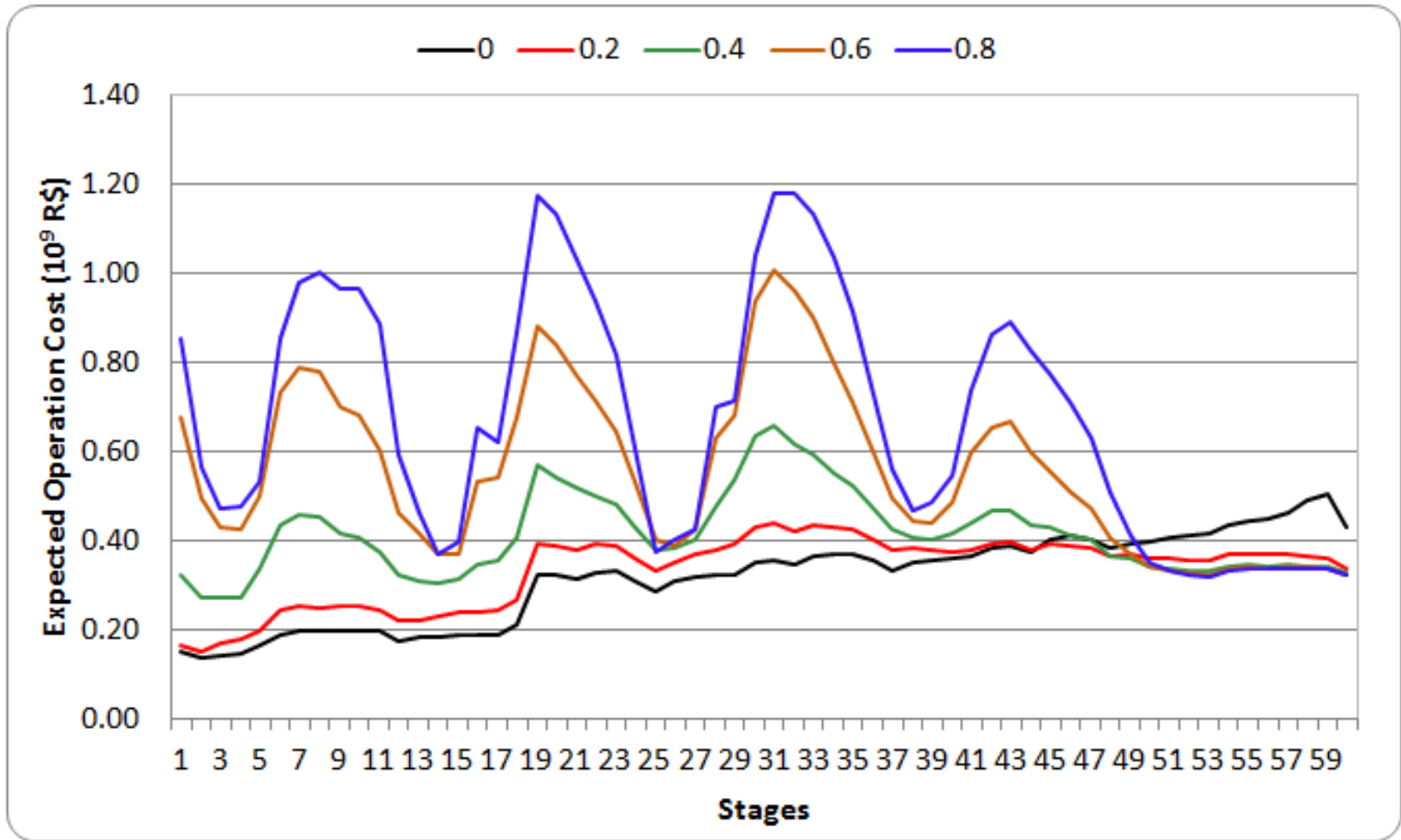
Computational Results - $\alpha = 10\%$

Expected Thermal Generation



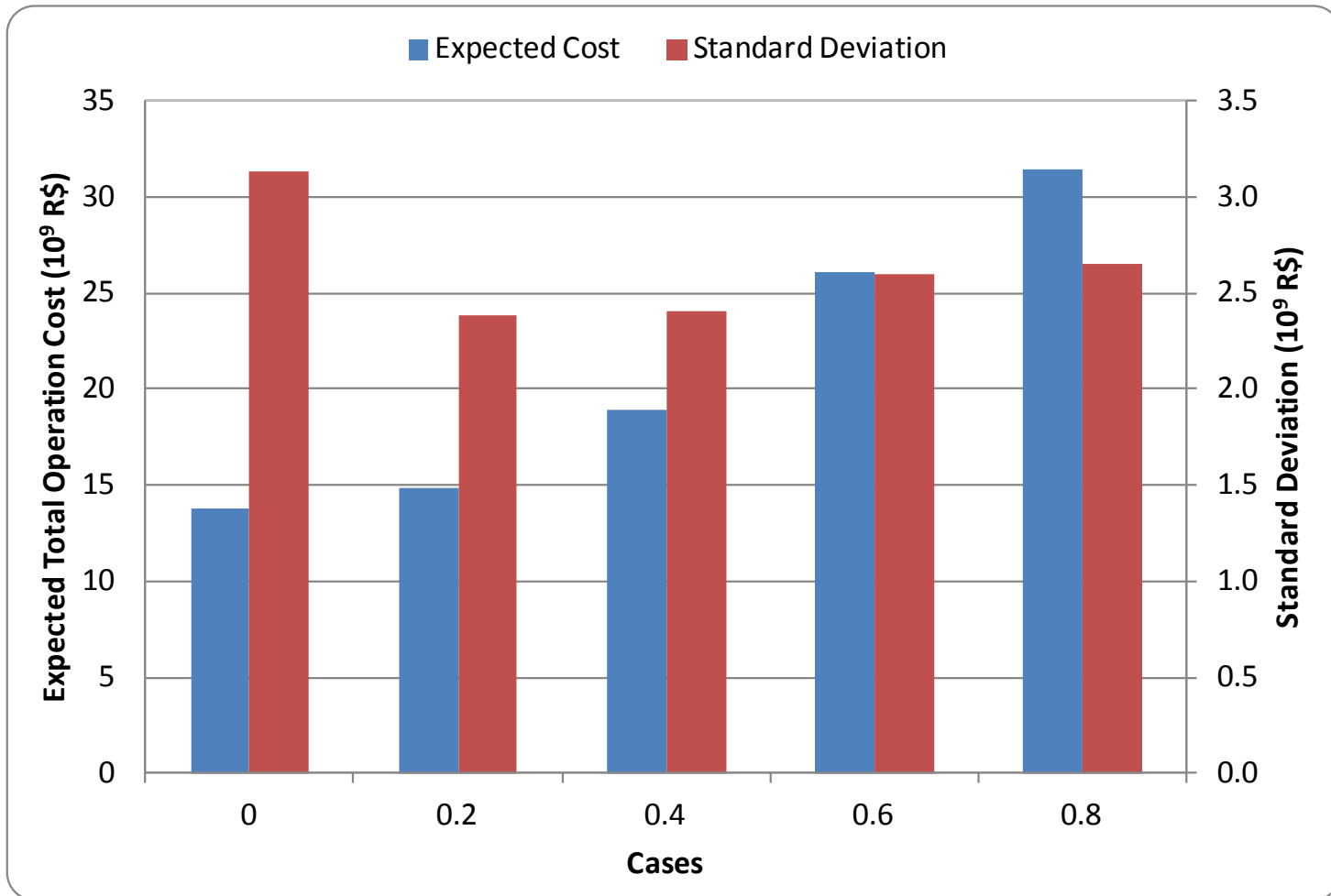
Computational Results - $\alpha = 10\%$

Expected Operation Cost per Stage



Computational Results - $\alpha = 10\%$

Expected Total Operation Cost



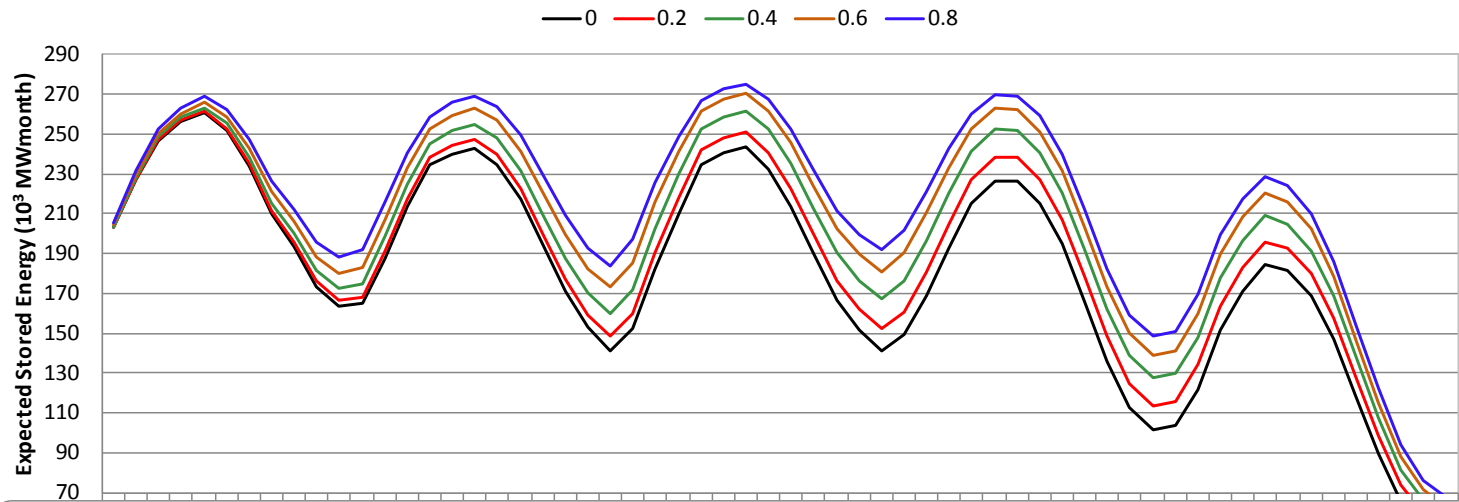
Computational Results

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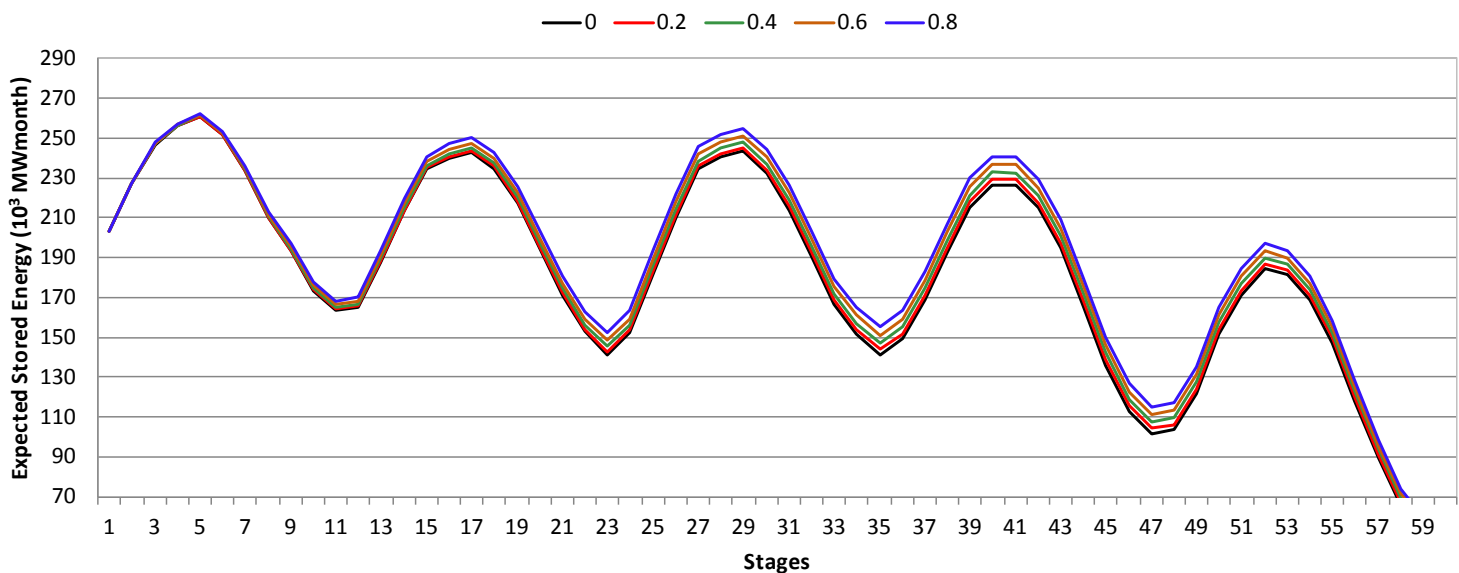


LabPlan

$\alpha = 50\%$



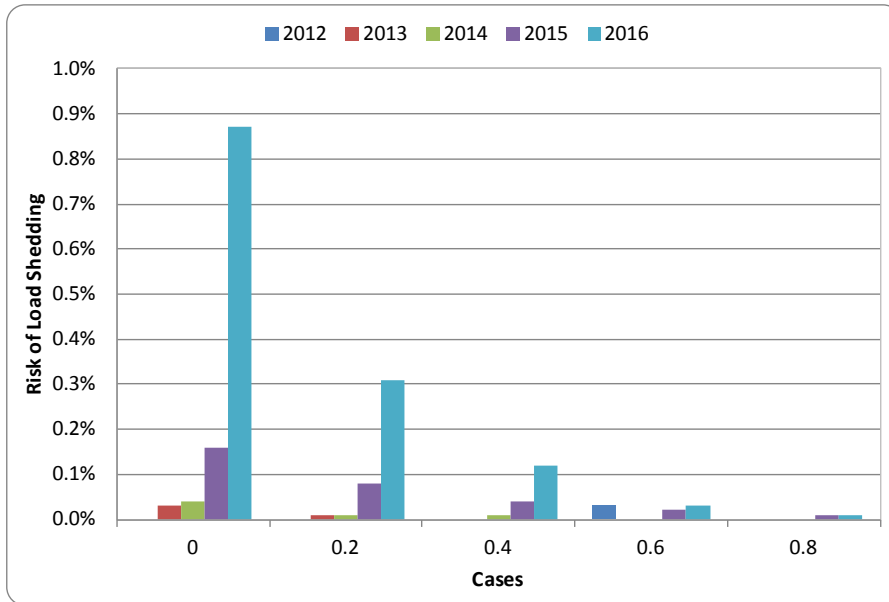
$\alpha = 90\%$



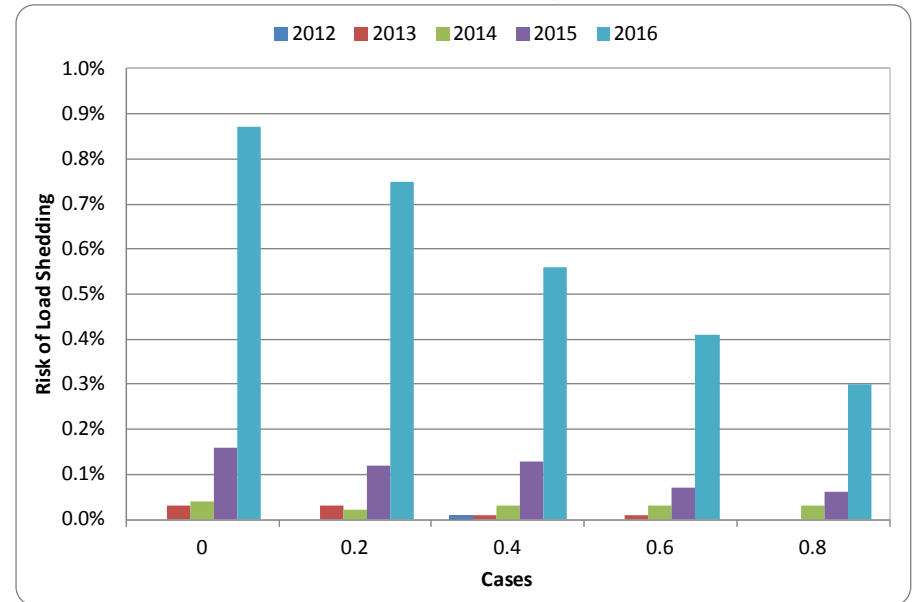
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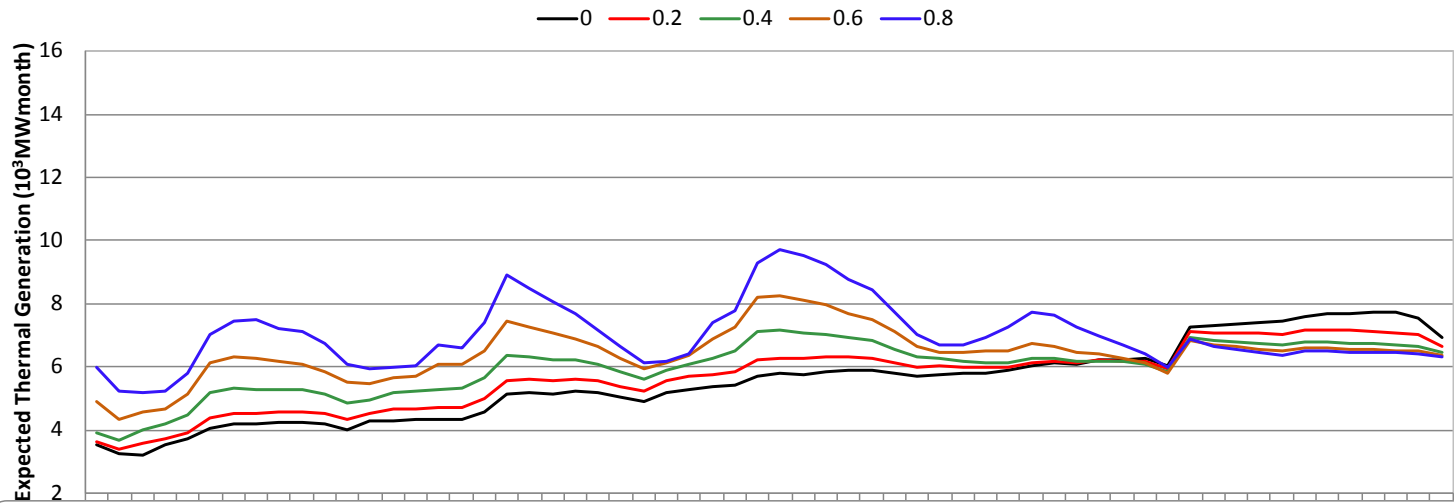
Computational Results

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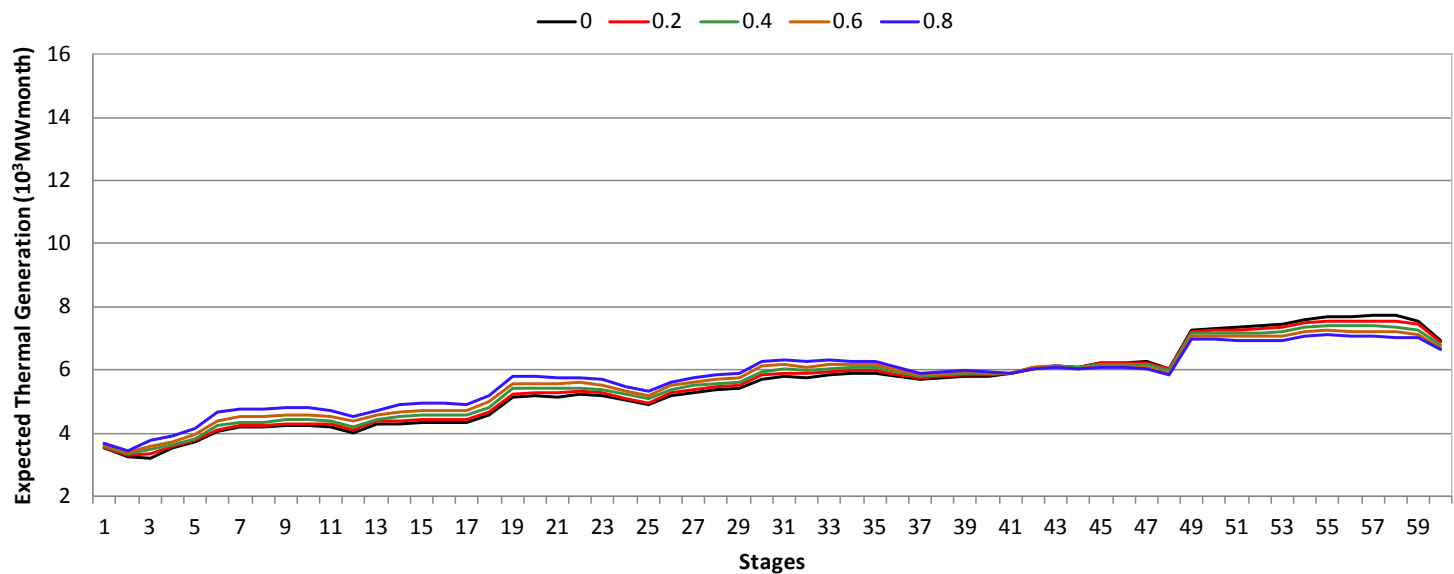


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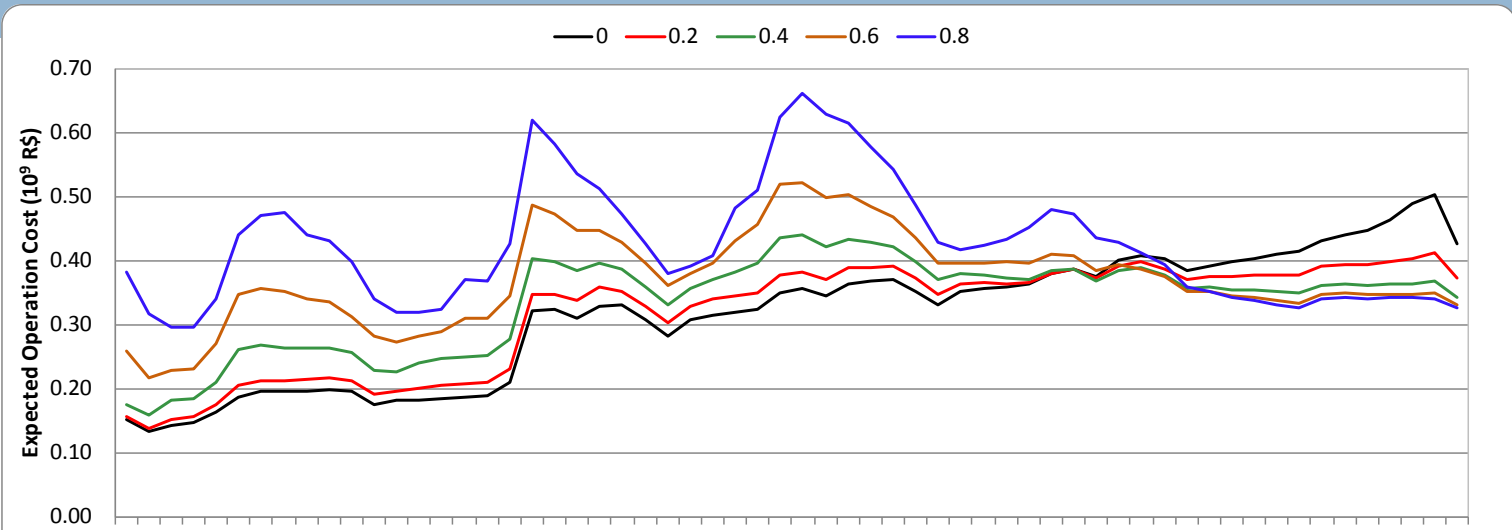
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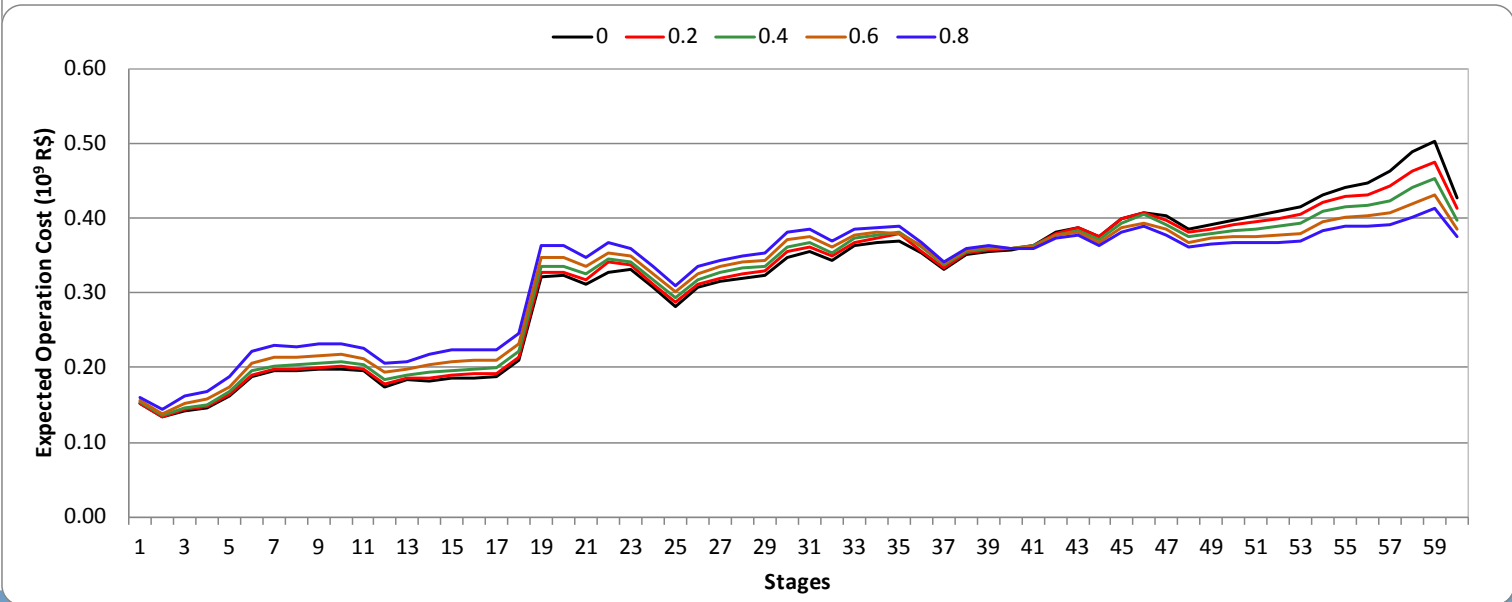
Computational Results

Expected Operation Cost per Stage

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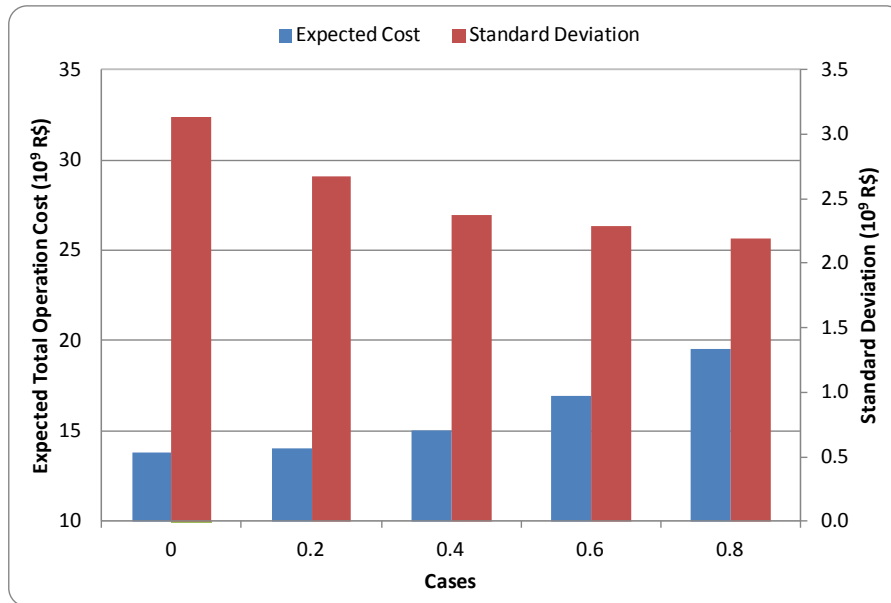
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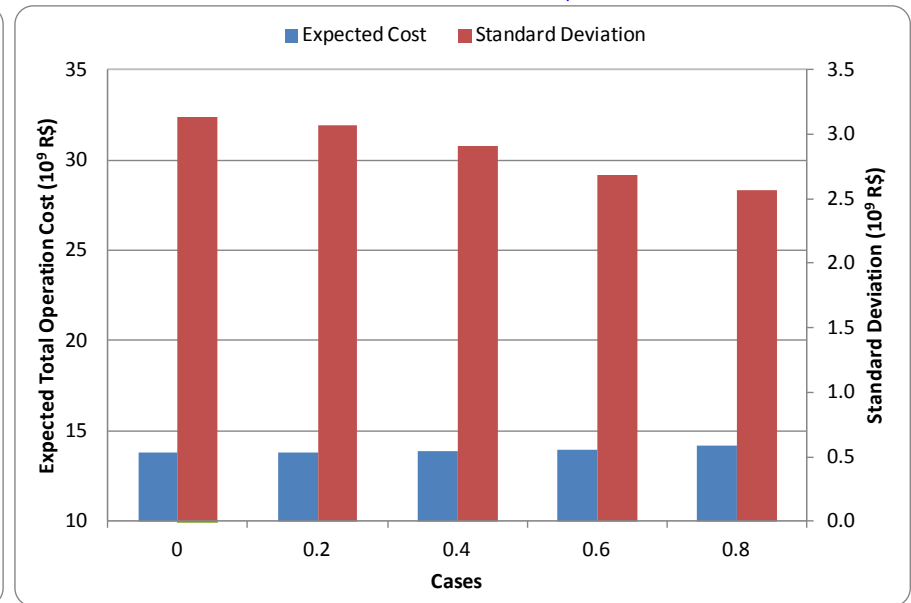
Computational Results

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Final Remarks



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 - ▣ It increases the storage in the Reservoirs
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- Consistent results with Risk Aversion
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- Easy and straight forward SDDP implementation

Final Remarks

- Consistent results with Risk Aversion
- In Hydrothermal Scheduling problems
 - ▣ It increases the storage in the Reservoirs
 - ▣ It reduces the risk of Load Shedding
 - ▣ However, it is more expensive on average
- Easy and straight forward SDDP implementation
- Difficulties
 - ▣ Upper Bound calculation for the SDDP stopping criterion
 - ▣ In Brazil the prices are given by the Lagrangean Multipliers, which in this case does not relate to prices directly

Thank you!

Kia Ora!

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