

## Construction Simulation Optimization Using VRTs

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### INTRODUCTION & AIM

Simulation is often coupled with an optimization algorithm to optimize the construction operations. However, the major drawback of the current stochastic simulation optimization methods is that they require a long computational time and may present inferior solutions in the final Pareto front.

#### Objective

To investigate the benefits of incorporating Variance Reduction Techniques (VRTs) in discrete event simulation optimization problems with a large and discrete set of decision variables.

Three VRTs are used: Common Random Numbers (CRN), Antithetic Variance (AV), and combined CRN+AV

#### Anticipated Benefits

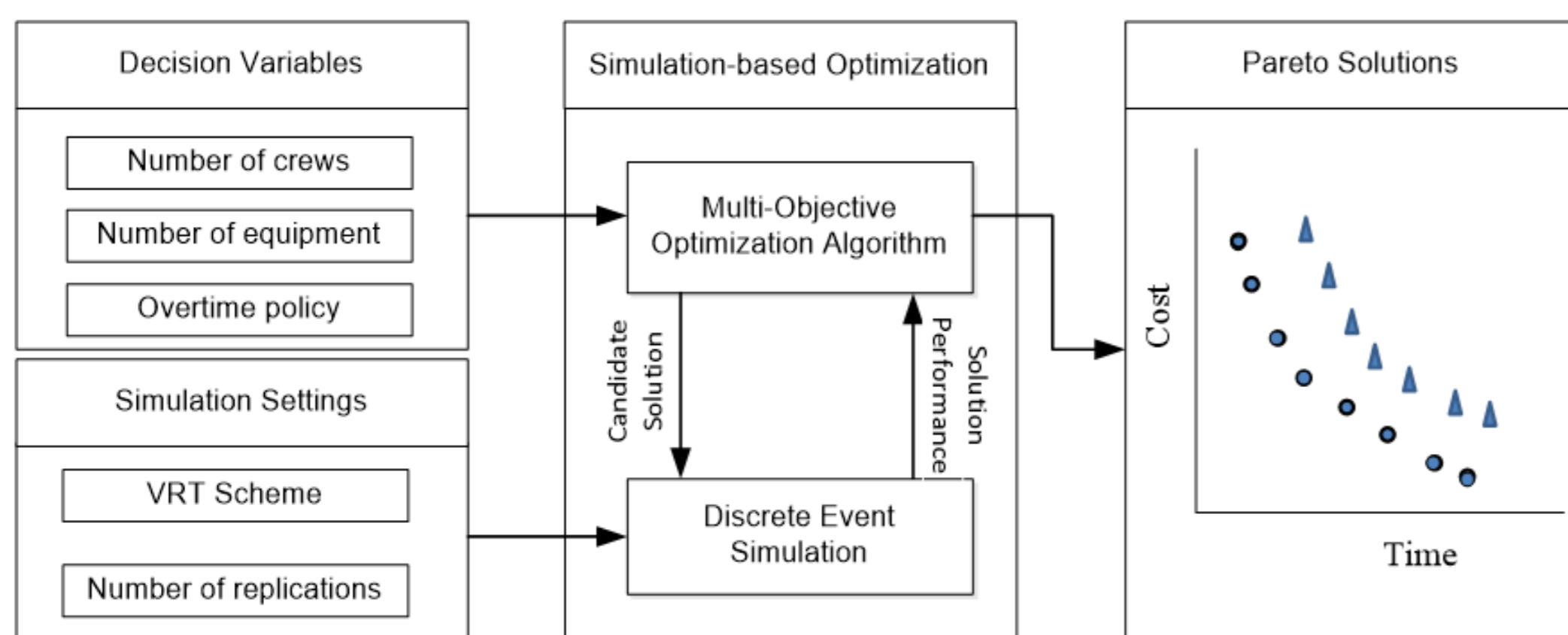
Reduce the computation time.

Improve the quality of the optimal solutions.

Increase the confidence in the optimality of the solutions.

### METHOD

#### Proposed Framework



#### Integration Algorithms

##### DES Optimization Algorithm Under CRN

- 1 Generate and store seed number
- 2 Initialize population
- 3 **FOR** each population until termination **DO**
- 4 **FOR** each solution in a population
- 5 Run simulation
- 6 **IF**  $j = 1$  **Then** generate streams using seed number
- 7 **ELSE** continue from last random number
- 8 **END IF**
- 9 Calculate duration and cost
- 10 Sort population and apply genetic operators
- 11 **RETURN** Pareto solutions

##### DES Optimization Algorithm Under AV & CM

- 1 Generate and store seed number if CM is used
- 2 Initialize population
- 3 **FOR** each population until termination **DO**
- 4 **FOR** each solution in a population
- 5 Run simulation
- 6 **FOR** each standard replication
- 7 **IF**  $p = 1$  **Then** generate streams using seed number
- 8 **ELSE** continue from last random number
- 9 **END IF**
- 10 **FOR** each antithetic replication
- 11 Use complimentary random numbers
- 12 Calculate duration and cost
- 13 Sort population and apply genetic operators
- 14 **RETURN** Pareto solutions

### RESULTS & DISCUSSION

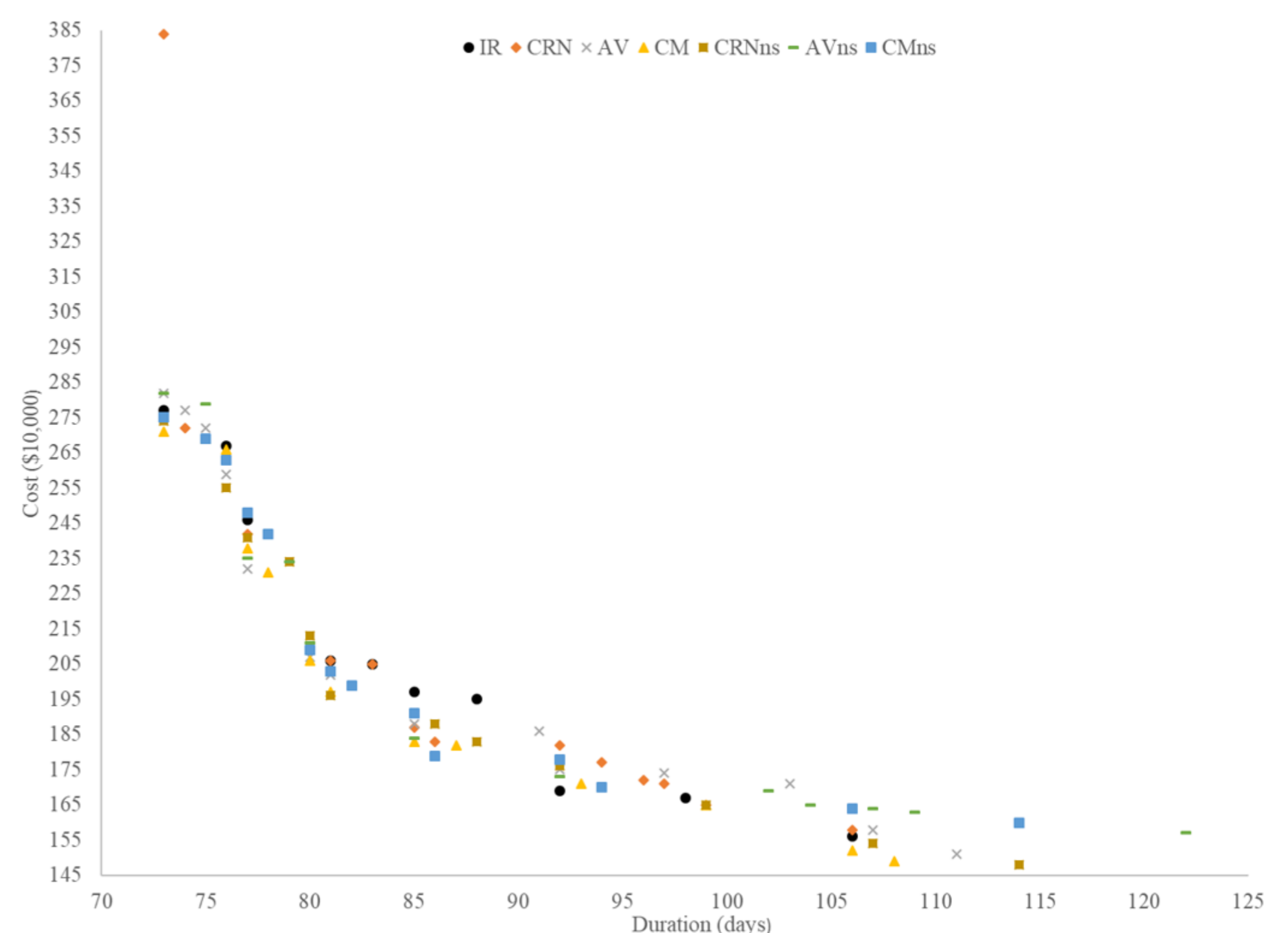
#### Performance Metrics

$$Time\ Savings\ (T_s) = \frac{T_{IR} - T_{VRT}}{T_{IR}} \times 100$$

$$\Delta HyperVolume(HV) = \frac{HV_{IR} - HV_{VRT}}{HV_{IR}} \times 100$$

$$Inferior\ Solutions\ (IS) = \frac{IS_{scheme}}{PS_{scheme}} \times 100$$

Obtained Pareto Fronts using Each Technique



#### Summary of the Performance Metrics Results

Technique	$N$	$P$	$T$ (hours)	$T_s$ (%)	$HV$	$\Delta HV$ (%)	$IS$ (%)
IR	100	---	7.20	---	7,004	---	38
CRN	10	---	1.37	80.97	7,036	0.47	26
AV	---	5	1.33	81.53	7,160	2.24	13
CM	---	5	1.33	81.53	7,368	5.21	7
CRN <sub>n<sub>s</sub></sub>	10	---	1.27	82.36	7,265	3.74	20
AV <sub>n<sub>s</sub></sub>	---	5	1.28	82.22	7,038	0.50	18
CM <sub>n<sub>s</sub></sub>	---	5	1.28	82.22	7,162	2.27	0

$N$  = Number of replications;  $P$  = Number of pair runs

### CONCLUSION

Proposed method seems promising in improving overall performance of simulation optimization models. Time reduced by 81.81%, quality improved by 2.4%, and the presence of inferior solutions reduced by 63.15%.

### FUTURE WORK / REFERENCES

Further investigation of the proposed method.