



## Introduction

Serving the ever increasing demands for computational power, expenses of HPC centers increase in terms of acquisition, energy, and programming. Thus, a quantifiable productivity metric is needed in HPC procurements to make an informed decision on how to invest available budgets.

$$\text{productivity} = \frac{\text{value}}{\text{cost}[\$]}$$

$$\text{\$} = \text{HW} + \text{energy} + \text{development cost} + \dots$$

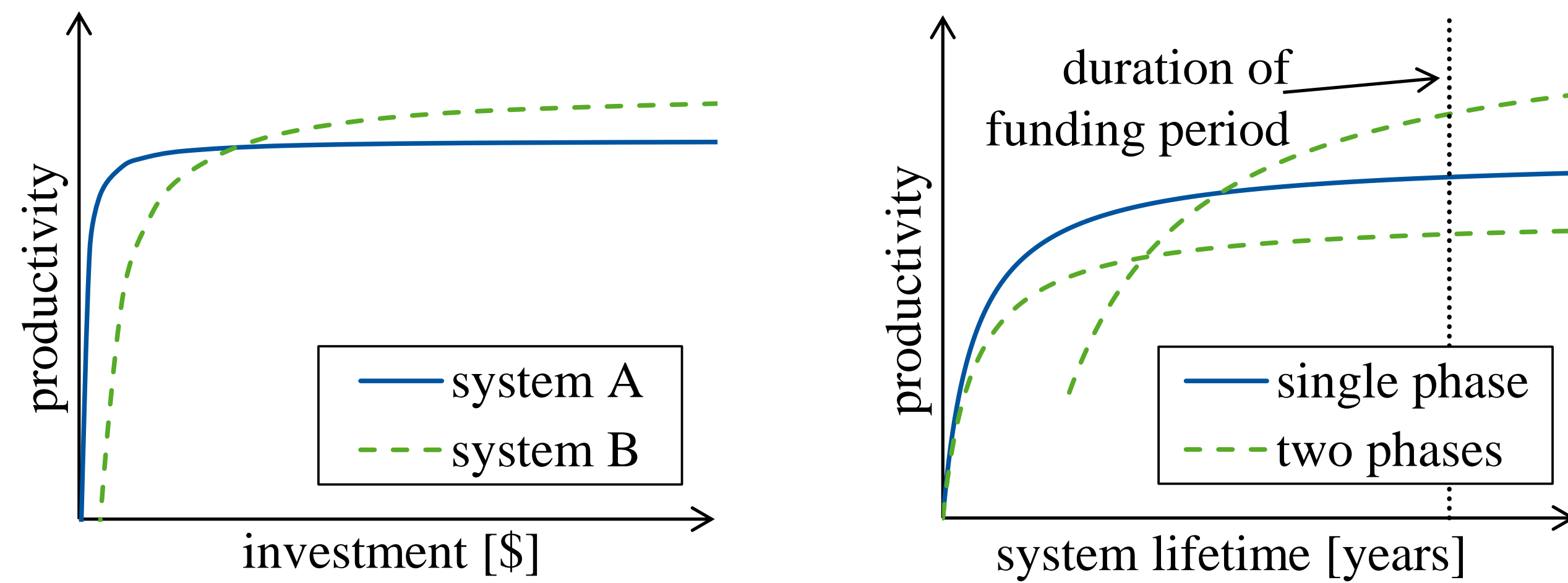
$$\text{dev. effort} [\text{day}] * \text{salary} \left[ \frac{\$}{\text{day}} \right]$$

### Challenges

- Real-world applicability w/ multi-job setups
- Quantification of value
- Prediction of all productivity parameters
- Especially, estimation of development effort
- Intangible character of various impact factors
- Data collection through human-subject research

## Productivity

I introduce a productivity model with predictive power to foster informed decision making in HPC procurements.



For instance, it can be used to compare various systems, or to argue for single- or two-phase procurements.

$$\text{productivity } \psi(n, \tau) = \frac{\sum r_{app,i}(n, \tau)}{\text{TCO}(n, \tau)}$$

$n$ : system size [#nodes]  
 $\tau$ : system lifetime [years]

### Value: Number of Application Runs

- Number of runs of (relevant) simulation applications
- Focus on application runtime  $t_{app,i}$  for application  $i$  running on  $n_i$  compute nodes
- Overarching metric for real-world job mix setups

$$r_{app,i}(n, \tau) = p_i \cdot \frac{q_{app,i}(n_i) \cdot \alpha \cdot n \cdot \tau}{n_i \cdot t_{app,i}(n_i)}$$

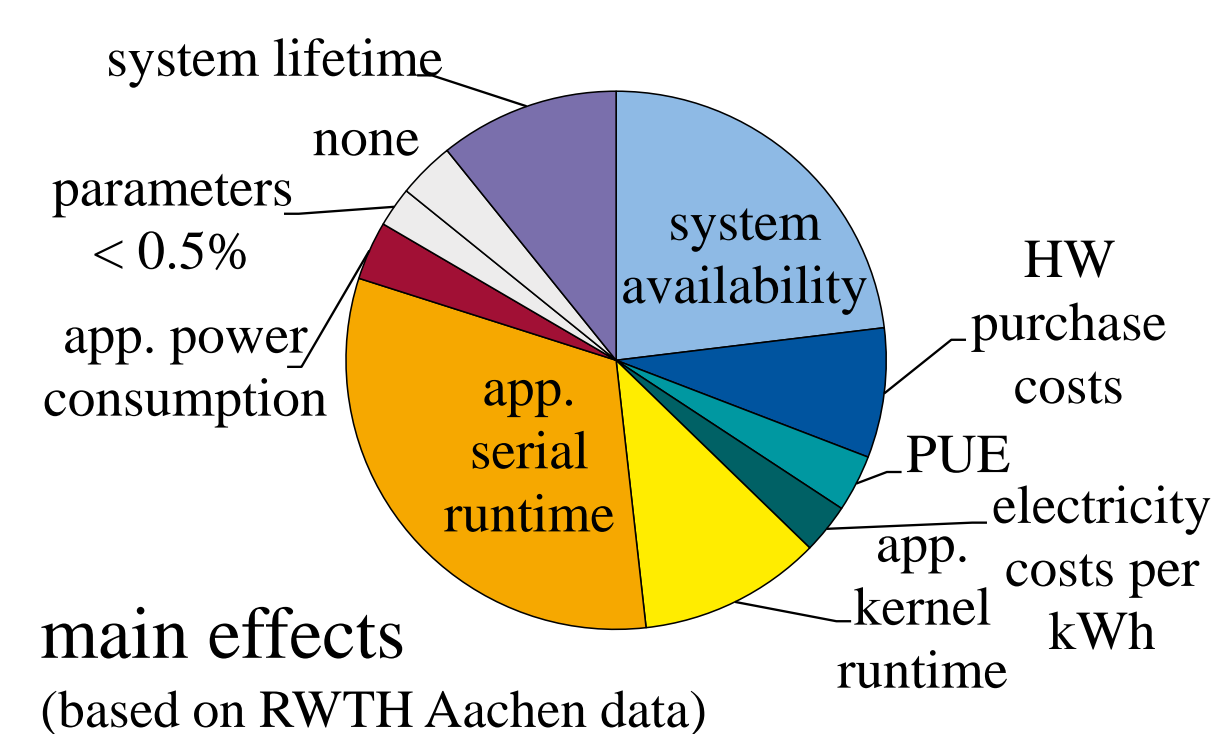
$\alpha$ : system availability  
 $q_{app,i}$ : quality weighting factor  
 $p_i$ : capacity weighting factor

### Cost: Total Cost of Ownership (TCO)

- One-time costs  $C^{ot}$  and annual costs  $C^{pa}$
- Categorization: costs per node and per node type
- Inclusion of costs for, e.g., programming, energy, hardware

$$\text{TCO}(n, \tau) = C^{ot}(n) + C^{pa}(n) \cdot \tau$$

### Sensitivity Analysis



- Variances in productivity due to errors in assumptions
- Only few (well-understood) parameters must be accurately predicted → robust

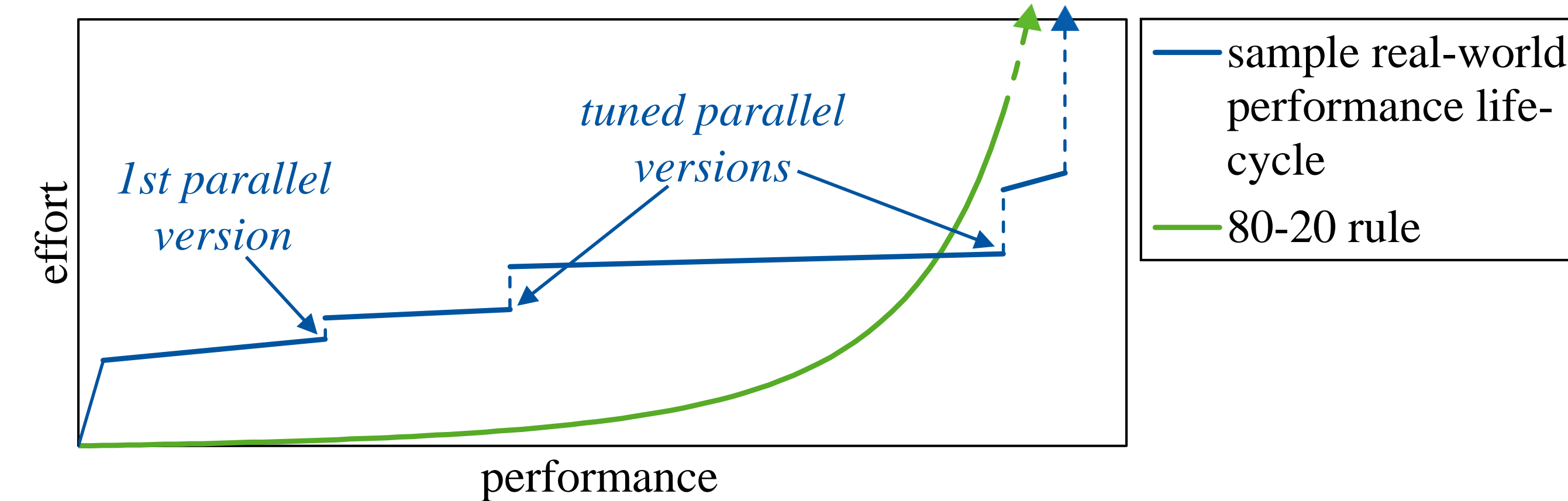
## Development Effort

As part of a sound TCO and productivity model, I introduce a methodology to estimate development efforts needed to parallelize, port and tune simulation codes to efficiently exploit (novel) HPC hardware.

### Performance Life-Cycle

- Model of relationship of performance and effort needed to achieve this performance

$$\text{effort} = S \cdot f(\text{performance})^R + Q$$



- Numerous factors impact effort (captured in  $S, R, Q$ )
- Regression analysis from collected data

### Impact Factors on Effort

Pre-knowledge on HW & parallel prog. model	Code work
Pre-knowledge on numerical algorithm used	Parallel prog. model & compiler/ runtime system
Performance	Architecture/ hardware
Tools	Kind of algorithm
Code size	Portability & maintainability over code's lifetime
Energy efficiency	

### Identification of key drivers

- Focus on most influencing factors
- Factor ranking based on surveys

### Quantification of “pre-knowledge”

- Knowledge surveys (confidence rating)

### Quantification of “parallel programming model & numerical algorithm”

- Pattern-based suitability mapping

### Data Collection

- Basis of statistically reliable results are sufficient data sets
- Targeting at community effort: tools and material online, e.g., effort-performance pairs by our tool *EffortLog*

## Conclusion

My research covers a productivity figure of merit for HPC procurements that is applicable to real-world multi-job setups. To embrace an estimation of software development effort into the productivity model, I provide a methodology that is based on performance life-cycles and statistical analysis of data collected in human-subject research.

In future, I will continue to collect corresponding data sets and investigate conditional refinements of the productivity model.

## References

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