

# ASK: Adaptive Sampling Kit

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Exascale Computing Research

2012/08/29

# Outline

- 1 Building Empirical Performance Models
- 2 Adaptive Sampling Kit
- 3 Hierarchical Variance Sampling
- 4 Evaluation

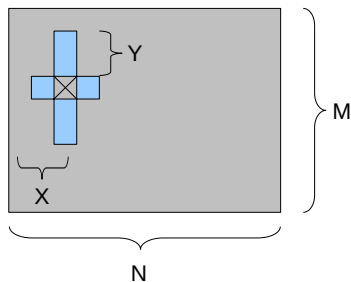
# Motivation: Building Performance Models

- Building performance models is important to
  - ▶ Understand performance bottlenecks
  - ▶ Optimize applications
  - ▶ Find best architecture for a given application (co-design)

# Motivation: Building Performance Models

- How to model performance ?
  - ▶ Using simulators or analytical models
    - ★ Architectures are complex and many factors interact (memory hierarchy, amount of parallelism, mapping, access patterns)
    - ★ Often models are too complex or costly
  - ▶ Black-box approach:
    - ★ Measure performance for different hardware or software configurations (the design space)
    - ★ Build an empirical model

# Design Space example: Jacobi Stencil code



- $T$ , number of OpenMP Threads, between 1 and 32
- $N$  and  $M$  between 64 and 2048
- $X, Y \in \{1, 2, 4, 8, 16\}$
- Design space size around  $31 \cdot 10^8$
- What is the performance for any combination of factors ?

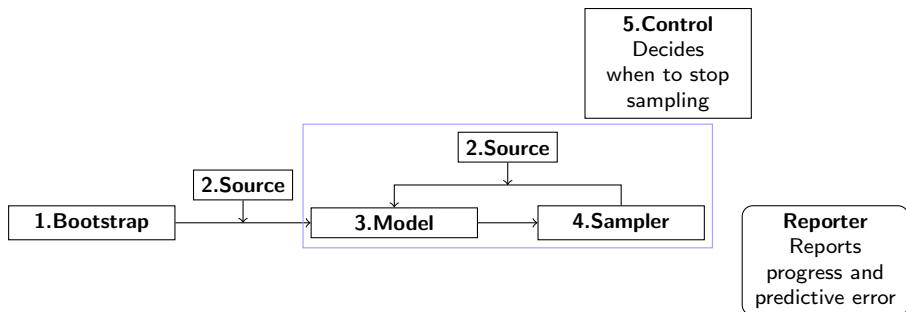
# Building empirical models

- Exhaustively measuring large design spaces is prohibitive.
- Build an accurate performance model with as few samples as possible
- Sampling method to select which points to measure
  - ▶ Samples must be chosen with care or the model will be biased.
- Regression model to estimate the missing samples
  - ▶ Linear, polynomial, SVM, Gaussian Process, Regression Trees, etc.
- No *one size fits all* strategy:
  - ▶ Depending on the design space response some models and sampling methods will work better than others
  - ▶ Important to try different strategies

- The contributions of this work are:
  - ▶ ASK open-source toolkit to build empirical models
    - ★ Easy to try different sampling strategies
  - ▶ A novel sampling strategy HVS
    - ★ Evaluated on different performance characterization problems

# ASK: Adaptive Sampling Kit

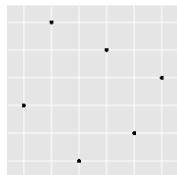
- Adaptive Sampling Kit (ASK) is a toolkit for building empirical models
- Modular architecture for conducting experiments:
  - ▶ Easy to combine different sampling strategies and models
  - ▶ Gathers state-of-the-art sampling methods
  - ▶ Provides visualization modules to supervise the sampling
  - ▶ Provides control modules to stop the sampling when its accurate enough



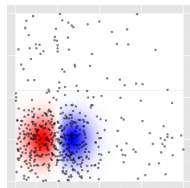


# Sampling methods included in ASK

- Sampling methods fall in two main categories
- Static methods: Space Filling Designs
  - ▶ Select a set of samples covering the design space
  - ▶ All points are measured in a single batch
    - ★ Latin Hyper Cube
    - ★ Maximin
    - ★ Low discrepancy
    - ★ Random
- Adaptive methods:
  - ▶ Sampling iteratively adapts to the design space complexity
    - ★ AMART [Li09]: a Query-By-Committee method
    - ★ TGP + ALC [Gramacy09]: an Error-reduction method
    - ★ HVS: a novel Error-reduction method that takes into account bias



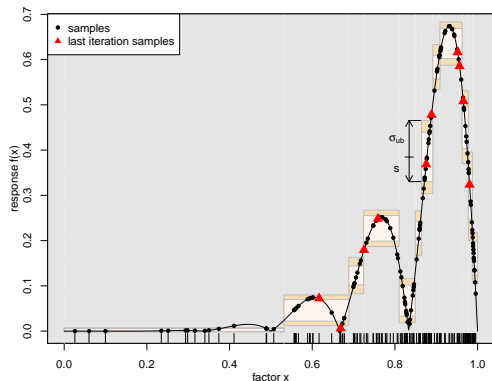
Latin Hyper Cube



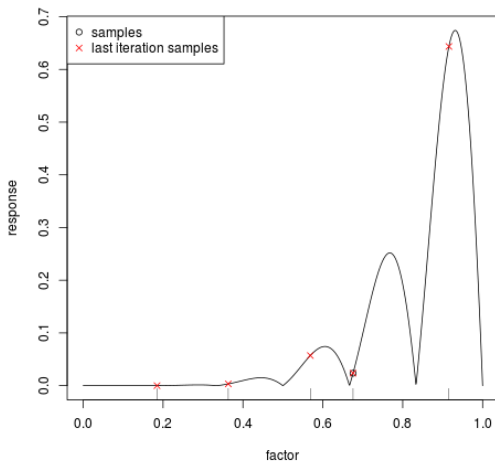
Adaptive Sampling

# Hierarchical Variance Sampling (1/2)

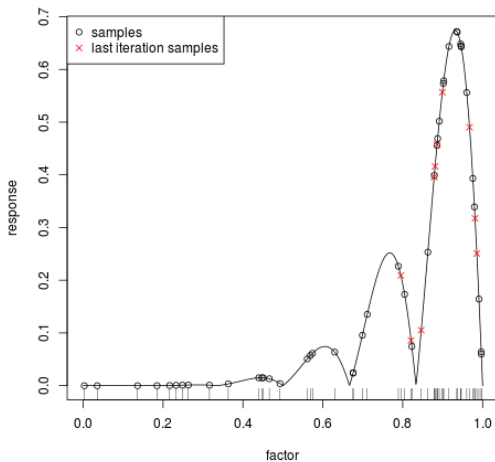
- Divide the space in regions using Regression Trees
- Compute the variance in each region
- Sample new points proportionally to: Variance upper bound  $\times$  size of the region



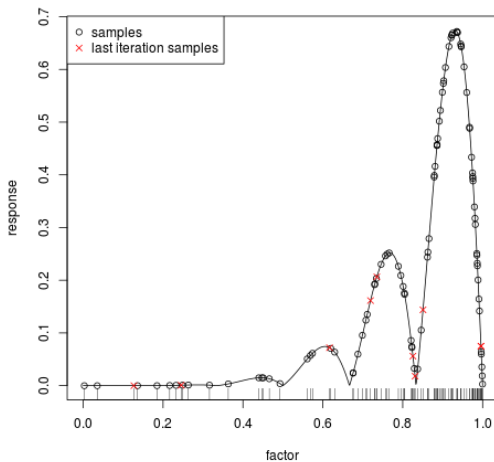
# Hierarchical Variance Sampling (2/2)



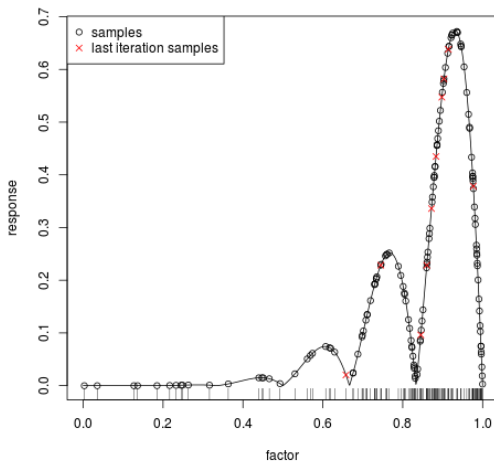
# Hierarchical Variance Sampling (2/2)



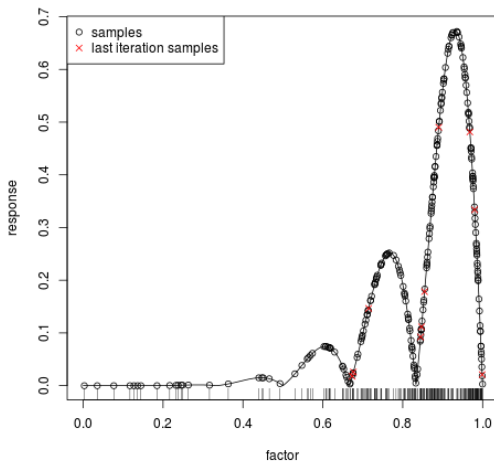
# Hierarchical Variance Sampling (2/2)



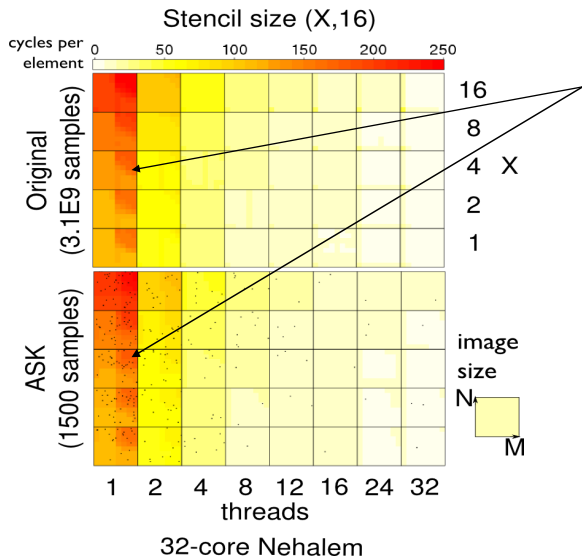
# Hierarchical Variance Sampling (2/2)



# Hierarchical Variance Sampling (2/2)



# ASK: Stencil code evaluation



- Despite using only 1500 points, HVS+GBM captures the performance features of the application.
- (25600 samples used as original response test set)
- 32 cores Xeon X7550 2.00GHz



# ASK: Evaluating estimation error

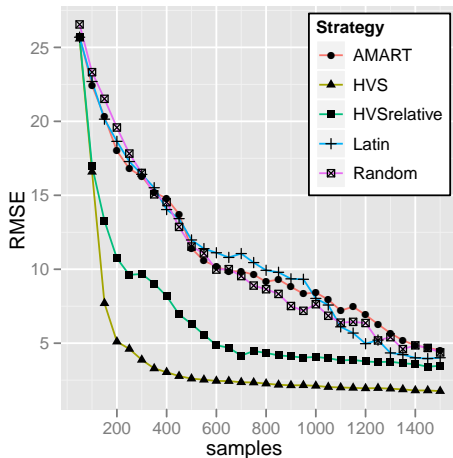


Figure: Stencils, Root Mean Square Error for different ASK sampling strategies

# Using the Model for prediction

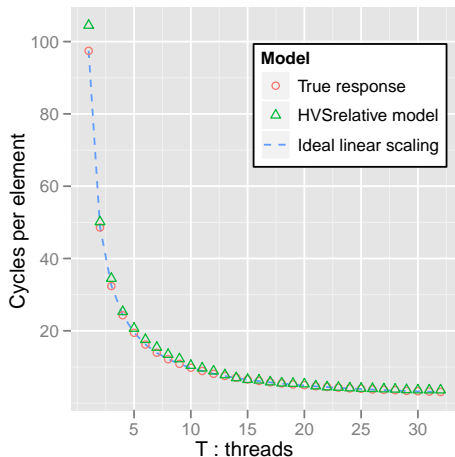
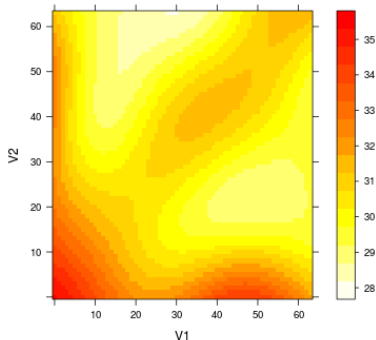


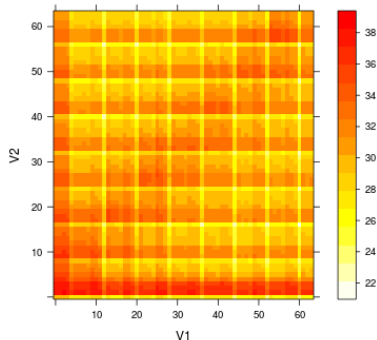
Figure: Scalability of the 8x8 stencil on a 1000x1000 matrix

# Importance of selecting a good model

- Influence of alignment stream benchmark
  - ▶ Three store streams hitting memory
  - ▶ Memory offsets:  $S(k)$ ,  $S(V1 + k)$ ,  $S(V2 + k)$
  - ▶ 4K aliasing
  - ▶ non aligned access overhead



SVM model



GBM model

# Alternatives to ASK

- SUrrogate MOdeling Lab (SUMO) [Gorissen2010]
  - ▶ Mature toolbox
  - ▶ Includes many models and sampling methods
  - ▶ Automatic tuning of model parameters
  - ▶ Supports modeling of multiple responses
  - ▶ ASK specifically targets performance characterization
    - ★ AMART [Li09] and HVS methods have been evaluated on performance problems
  - ▶ Only supports real-valued inputs
  - ▶ Depends on Matlab and is not open-source (but freely available for academic use)
- Caret R package [Kuhn2012]
  - ▶ Includes many models
  - ▶ Automatic tuning of model parameters
  - ▶ Does not handle sampling

# How to get ASK ?

- ASK is open-source and available at
  - ▶ <http://code.google.com/p/adaptive-sampling-kit/>
  
- The experimental data used in the paper is available at
  - ▶ <http://code.google.com/p/adaptive-sampling-kit/wiki/ExperimentalData>