



# Profiling and Autotuning for Energy-Aware Approximate Programming

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#### Motivation



- Approximation has well-known benefits
  - Energy saving, performance, etc.
  - Thus this workshop
- But, as a developer, how do we write an approximate application?
  - How do we understand/manage tradeoffs between energy/performance and quality/precision?
- Key to adoption: easy-to-use, configurable tools that assist developers





# This talk: Prototyping tools



- Development often starts with prototyping
- What should an approximation profiler look like?
  - What tools are needed?
- We propose a three layered system
  - Approximation layer: Provide simple, coarse-grained approximate semantics.
  - Profiling layer: Determine quality (QoR) impacts, and energy/performance benefits
    - Allow customization of approximate semantics, benefits
  - Autotuning layer: Suggest refinements to approximation that may improve tradeoffs





#### **EnerCaml**



- EnerCaml: our implementation of this design
- Built on top of OCaml
  - An ML variant with object-oriented extensions
  - Often used for prototyping
  - Functional style great for coarse-grained approximation
- Contains the three layers described earlier
  - Code-centric approximation via primitive call
  - Profiling with customizable quality metrics
  - Autotuning by searching for alternate preciseapproximate decompostions





#### **Approximation Layer**



- Key primitive for code-centric approximation
  - EnerCaml.approximate: (unit->'a)->'a approx
    - Takes a (thunked) code block (think C++ functor), executes it approximately, and returns an approximately-typed result.
    - Also provide endorsement, precise, continue primitives.
- Convenient model for prototyping just specify approximate kernels
- Natural fit for a functional language
  - Everything is a function
- Simulation: simply create precise and approximate versions of each function
  - Approximate versions executes approximate operations
  - Call sites in approximate functions call approximate versions





# Ray Tracer Approximation Example



```
(* Compute a pixel by sending rays in
   every direction *)
for dx = 0 to ss - 1 do
  for dy = 0 to ss - 1 do
    (* Compute direction vector *)
    (* Trace ray *)
    let next ray = ray trace dir scene in
    g := !g +. next ray;
  done;
done;
```



# Ray Tracer Approximation Example



```
(* Compute a pixel by sending rays in
   every direction *)
for dx = 0 to ss - 1 do
  for dy = 0 to ss - 1 do
    (* Compute direction vector *)
    (* Trace ray approximately *)
    let next ray = EnerCaml.approximate (
      fun () -> ray trace dir scene)
    in
    g := !g +. EnerCaml.endorse(next ray);
  done;
done;
```



# Next layer: Profiling



- Profiling layer lets users investigate the effects of approximation on their code
- Two key features:
  - Measure the quality of service and efficiency impacts of approximation.
  - Let users customize (defaults provided):
    - How operations are approximated (via custom error functions).
    - Relative energy savings of approximate operations (via custom scoring function)





### Measuring QoR impacts



- Profiling layer lets users define a quality function that compares data from precise and approximate executions.
- User also specifies data to collect to use as input to the QoR function.
  - Stored as a temporally ordered list.
- Profiler executes the application precisely and then approximately, and compare the data lists collected in the two executions using the QoR function.





# Example: Ray Tracer Profiling



```
(* loop over pixels *)
for (...)
  (* compute brightness g of current pixel *)
  (* add g to list of profile output for
     current execution *)
  EnerCaml.record profile output g;
done;
let psnr prec lst app list =
  (* compute PSNR of pixels in app list
     relative to pixels in prec list *)
in EnerCaml.eval qor psnr
```



# **EnerCaml Autotuning Layer**



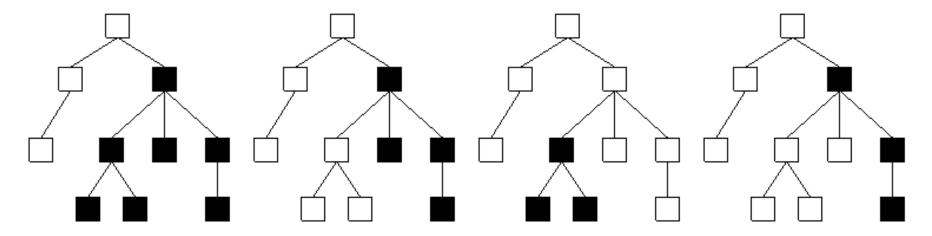
- Searches for alternative precise/approximate decompositions of programs that improve the quality and/or energy efficiency.
- Starts with the original approximation specified by the programmer.
  - Idea: specify coarsely, let autotuner refine
- Performs additional runs that remove part of the approximation.
  - Varies which function call sites call the precise versus the approximate version of the function.
- Never add approximation unsafe





# **Autotuning Search Strategies**





- Can't try every possible combination: exponential
- So, use some heuristics:
  - Remove approximation at a single static call site
  - Narrow approximation to a single static call site
  - Remove approximation from two "sibling" static call sites (call sites in the same calling function).





#### **Autotuning Output**



- Autotuner outputs the QoR and approximate operation counts for every trial.
- A trial dominates another trial if it has better QoR and more approximated operations.
- Non-dominated trials form quality-efficiency Pareto curve.
  - We output these trials with the code changes that produce them.
  - And plot these results.





#### Case Studies



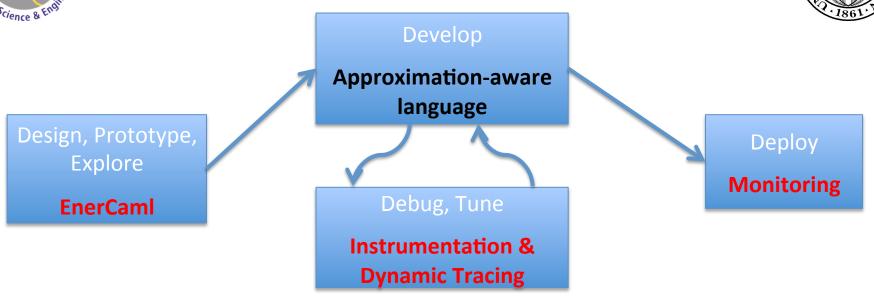
- Ray tracer:
  - Improved PSNR from 26.9 to 33.6, while maintaining nearly half of energy savings
- N-body simulation:
  - Improved QoR (average error<sup>-1</sup>) from 0.01 to nearly 4000, and maintained over half of the energy savings.
- Collision detection:
  - Reduced errors by 51% at expense of 30% approximation reduction.





# Part of a Larger Ecosystem





- Part of suite of dynamic tools we've designed for managing QoR of approximate applications
- Aimed at different phases of the software lifecycle:
  - EnerCaml for design and prototyping
  - Instrumentation & Tracing for debugging and tuning
  - Monitoring for real-time, post-deployment response to QoR issues
     Sallipa





#### Questions?







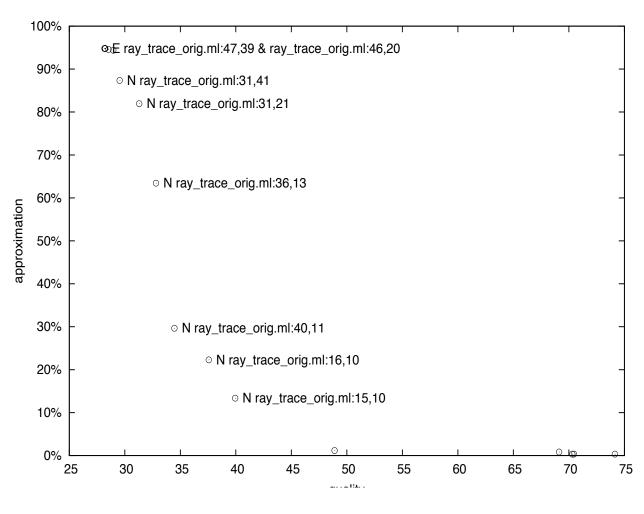
## Backup





## **Autotuning Example**









## **Tracking Approximation**



- To track our two function versions, the compiler creates dual function closures
  - Closures typically used to represent functions in languages where they are first-class values. Contain pointer to function and an environment.
  - Our dual closures replace the single function pointer in the closure with two: one for a precise version, and one for the approximate version.
  - Call the approximate version of lambda passed to approximate primitive call (and precise version in precise primitive)
  - All other calls are determined statically by context
    - If we are in a precise caller, calls go to the precise callee.
    - If we are in an approximate caller, call the approximate callee.





## **Specifying Approximation**



- EnerCaml's approximable operations:
  - Integer arithmetic
  - Floating point arithmetic
  - Integer and floating point array loads
- Approximation function for each of these: replaces result of the operation with another (possibly identical) result of the same type.
  - E.g., introduce a bit flip 0.1% of the time.
  - set float approximation : (float->float) -> unit
  - set integer approximation : (int->int) -> unit
  - set load approximation : (int->int) -> unit
  - set load float approximation : (float->float) -> unit
- Also, log approximate and precise operations, and let users create a customized energy score.
  - Default scorer is just percentage approximated.

