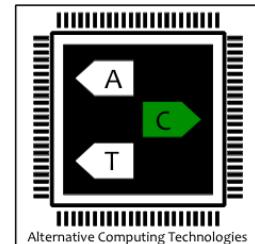


Expectation-Oriented Framework for Automating Approximate Programming

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Approximate Programming

Programmer's
manual/explicit specification

[EnerJ PLDI'11, Rely OOPSLA'13]



AUTOMATE
approximate programming

Approximate Programming

Programmer's
manual/explicit specification

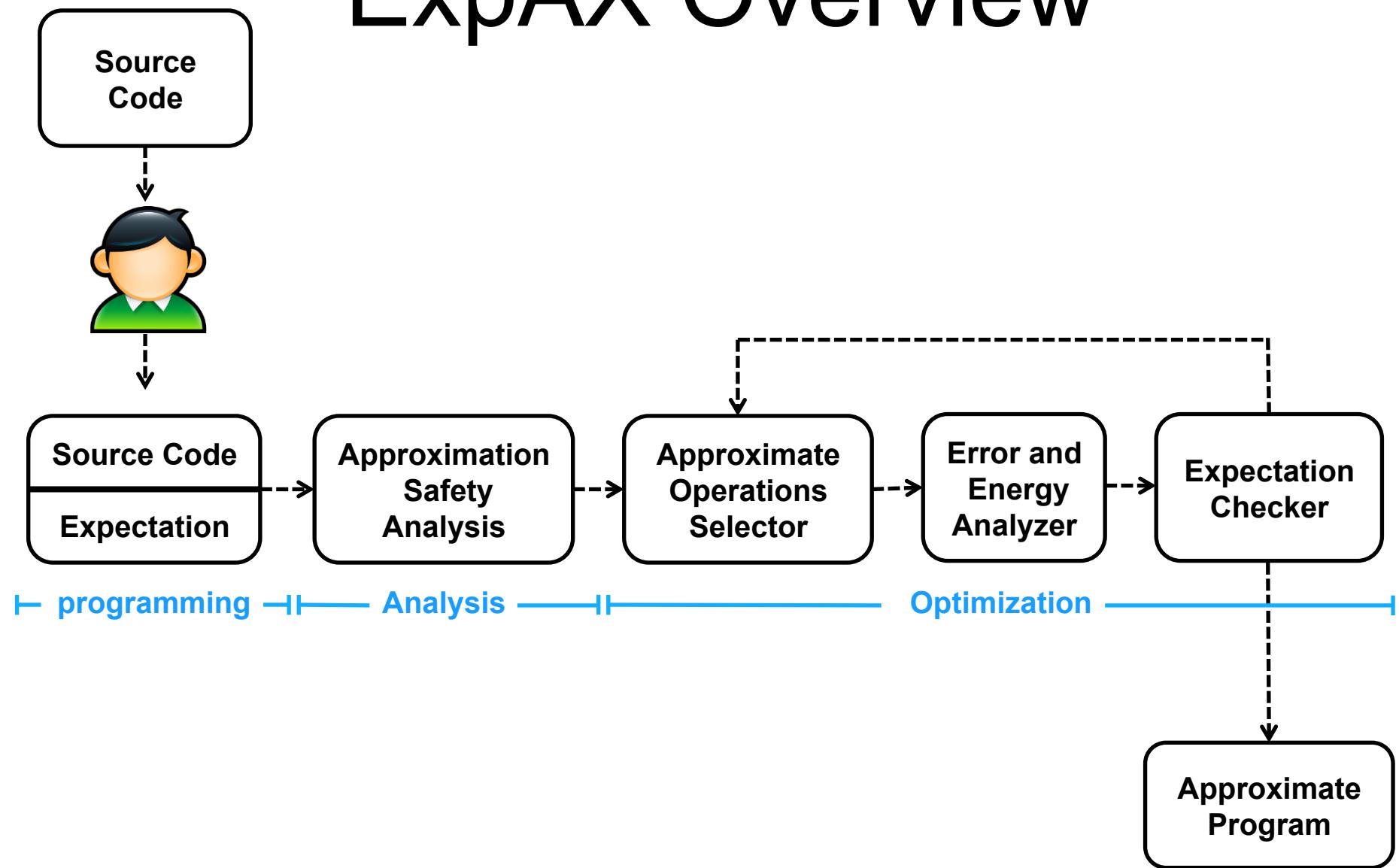
[EnerJ PLDI'11, Rely OOPSLA'13]



AUTOMATE
approximate programming

Where? How much?

ExpAX Overview



Programming Model

Programmer's Annotations with **Expectation**

1. **accept rate(v) < c**
e.g. accept rate(v) < 0.2
2. **accept magnitude(v) < c using f**
e.g. accept magnitude(v) < 0.1
3. **accept magnitude(v) > c using f with rate < c'**
e.g. accept magnitude(v) > 0.9 with rate < 0.3

Approximation Safety Analysis

Find possible **safe-to-approximate variables**

Unsafe-to-approximate variables

1. Variables violating memory safety
2. Variables violating functional correctness

Approximation Safety Analysis

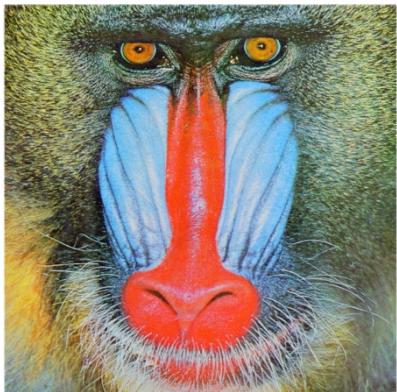
Backslicing Analysis

For each variable **v** in program,
find all variables **contributing** to the variable **v**

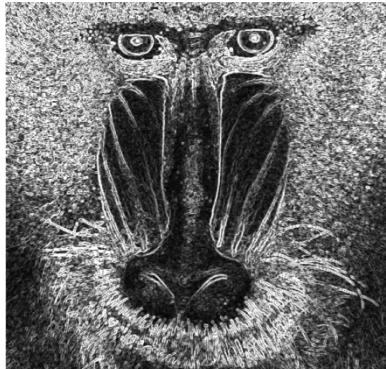
unsafe-to-approximate variables
should be **precise**

Everything else should be **precise** variables

Example



edgeDetection



```
Float sobel (float[3][3] p) {  
    float x, y, gradient;  
    x = (p[0][0] + 2 * p[0][1] + p[0][2]);  
    x += (p[2][0] + 2 * p[0][1] + p[2][2]);  
    y = (p[0][2] + 2 * p[1][2] + p[2][2]);  
    y += (p[0][0] + 2 * p[1][1] + p[2][0]);  
    gradient = sqrt(x * x + y * y);  
    ...  
    return gradient;  
}
```

```
void edgeDetection(Image &src, Image &dst) {  
    grayscale(src);  
  
    for (int y = ...)  
        for (int x = ...)  
            dst[x][y] = sobel(window(src, x, y));  
  
    accept rate(dst) < 0.1;  
}
```

Optimization

Find a **subset** of safe-to-approximate operations

- Minimize error
- Maximize energy saving

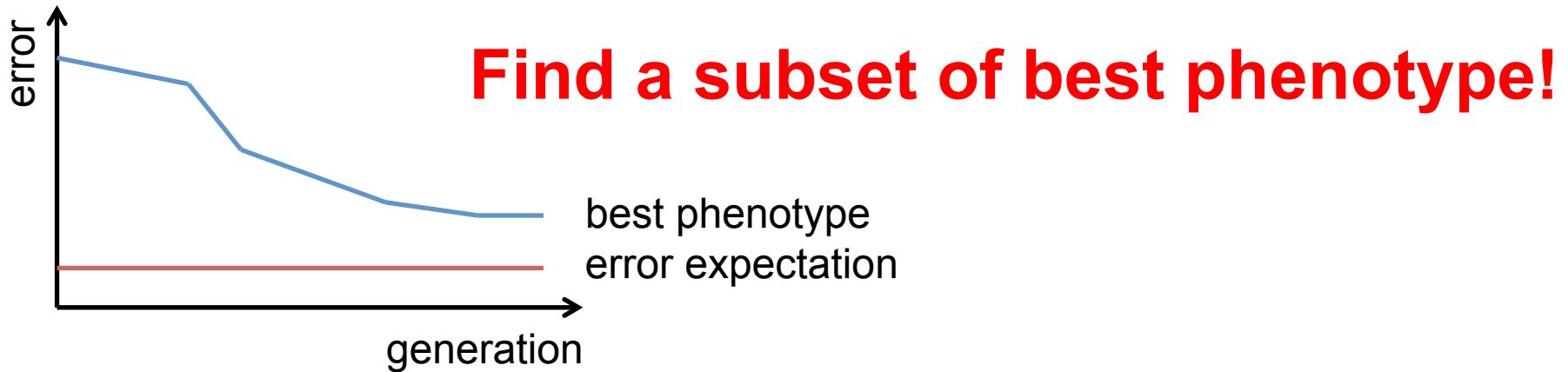
Objective function

$$f(\text{subset}) = (\alpha \times \text{error} + \beta \times \text{energy})^{-1}$$

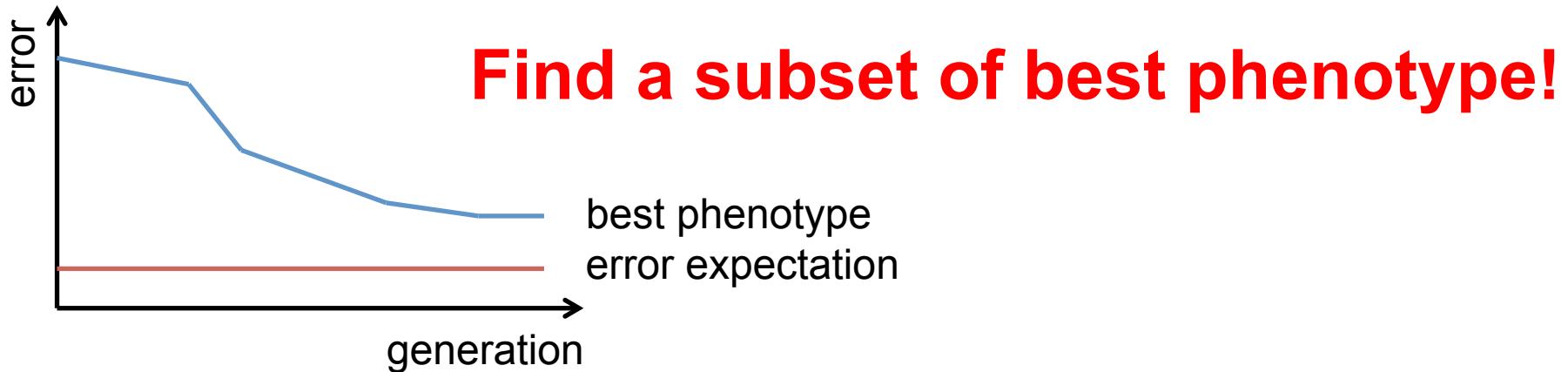
Genetic algorithm

phenotype: a bitvector representing a subset
(approximate('0') or precise('1'))

Statistical Guarantee



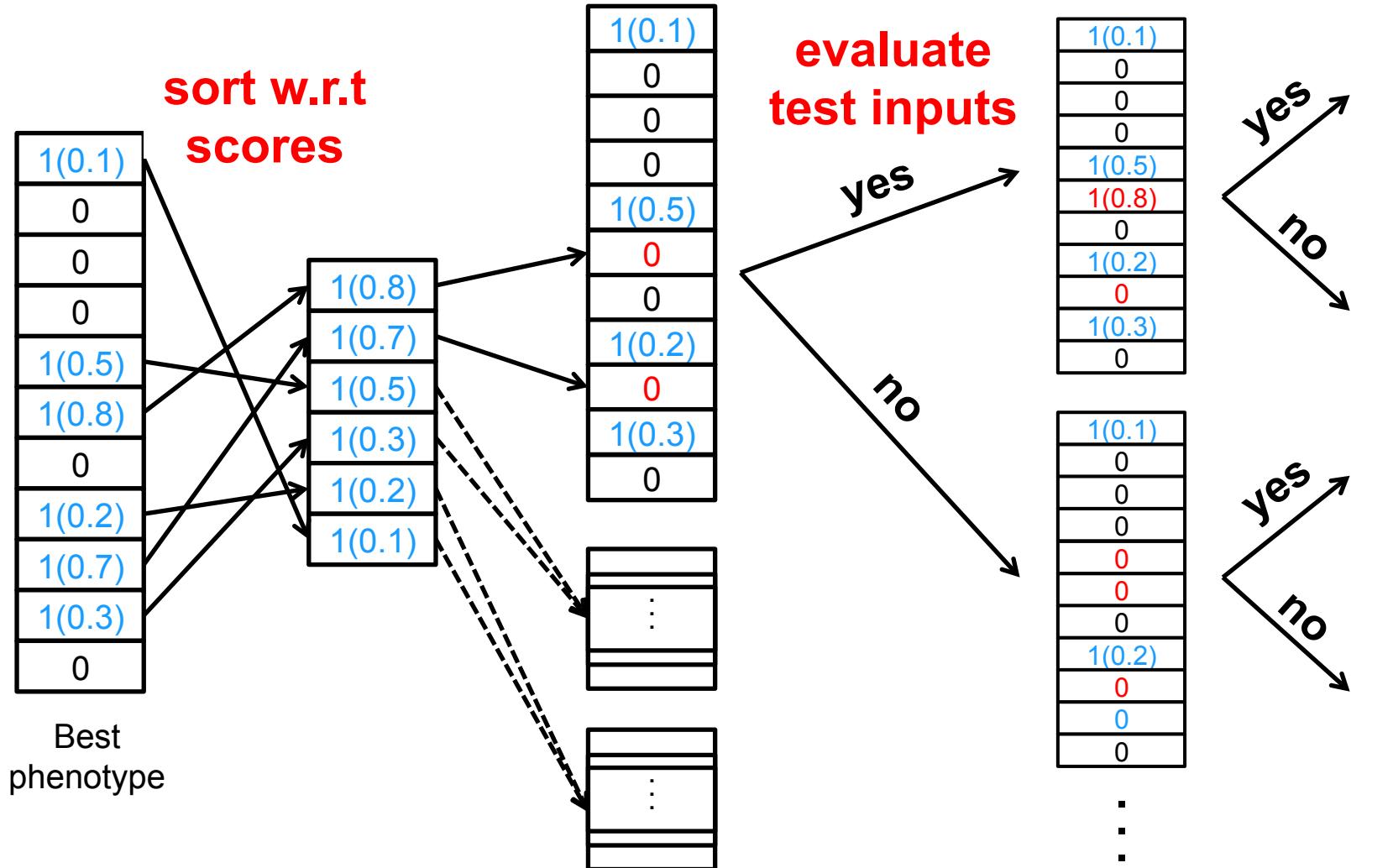
Statistical Guarantee



For each **eval** in genetic algorithm:
calculate a **score** for each **operation**

$$f(operation) = \sum_{eval \in Eval} \left(\frac{\alpha \times error + \beta \times energy}{n(approx)} \right) / n(Eval)$$

Space Exploration with Transformed Best Phenotype



Evaluation

Benchmarks:

scimark2 – FFT, LU, SOR, MonteCarlo, SMM
Imagefill, raytracer, jmeint, zxing

Analysis tool:

Jchord – Open source programming analysis platform for Java

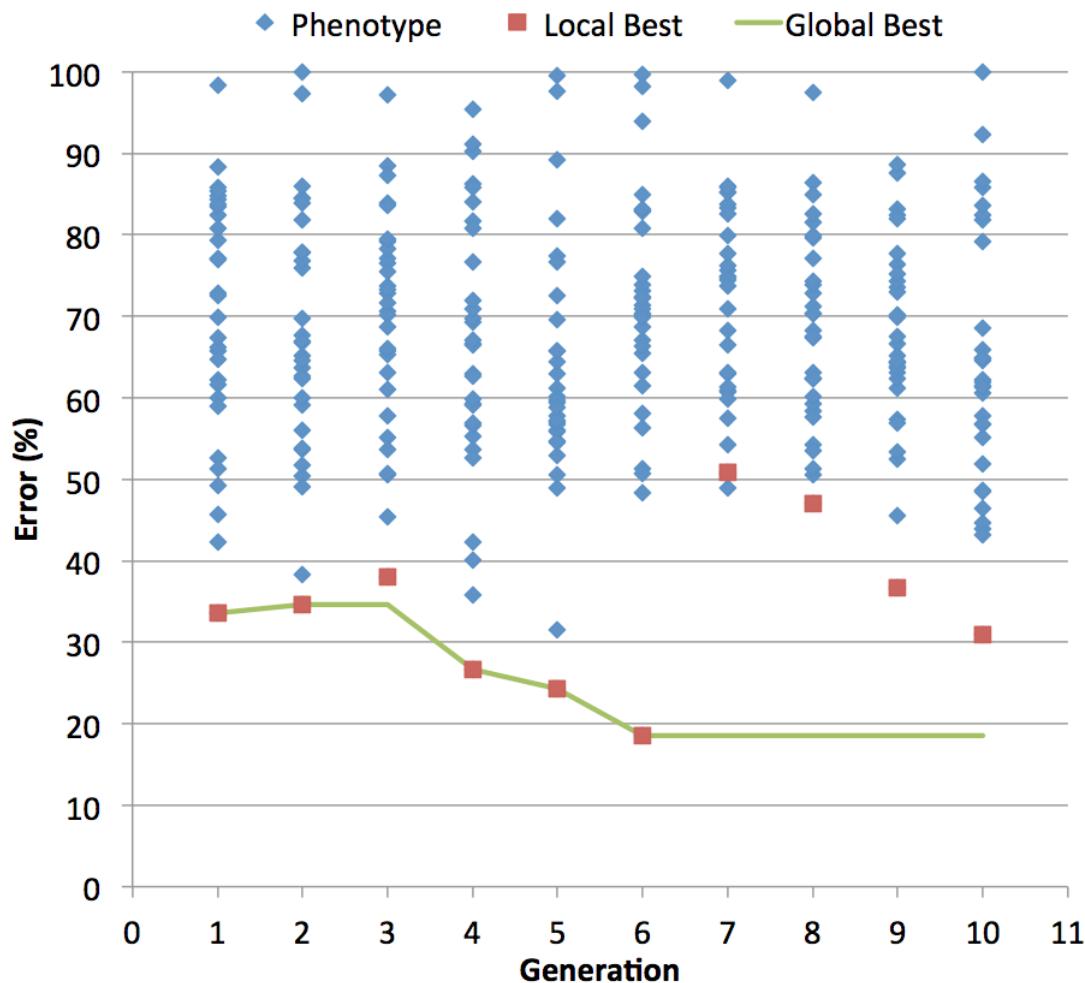
Simulator:

Open-source simulator provided by EnerJ

Analysis Result

BenchName	Enerj: # of Annotations	ExpAX: # of Expectations
FFT	27	1
LU	20	1
SOR	9	1
MonteCarlo	3	1
SMM	8	1
imagefill	28	7
RayTracer	27	2
jmeint	113	1
zxing	172	15

Genetic Algorithm Results



LU

Conclusion

Expx:

an expectation-oriented framework for
automating approximate programming

1. Programming model with a new program specification
2. Approximation safety analysis
3. Optimization framework with heuristics for statistical guarantee