

Auto-Tuning Framework (ATF)

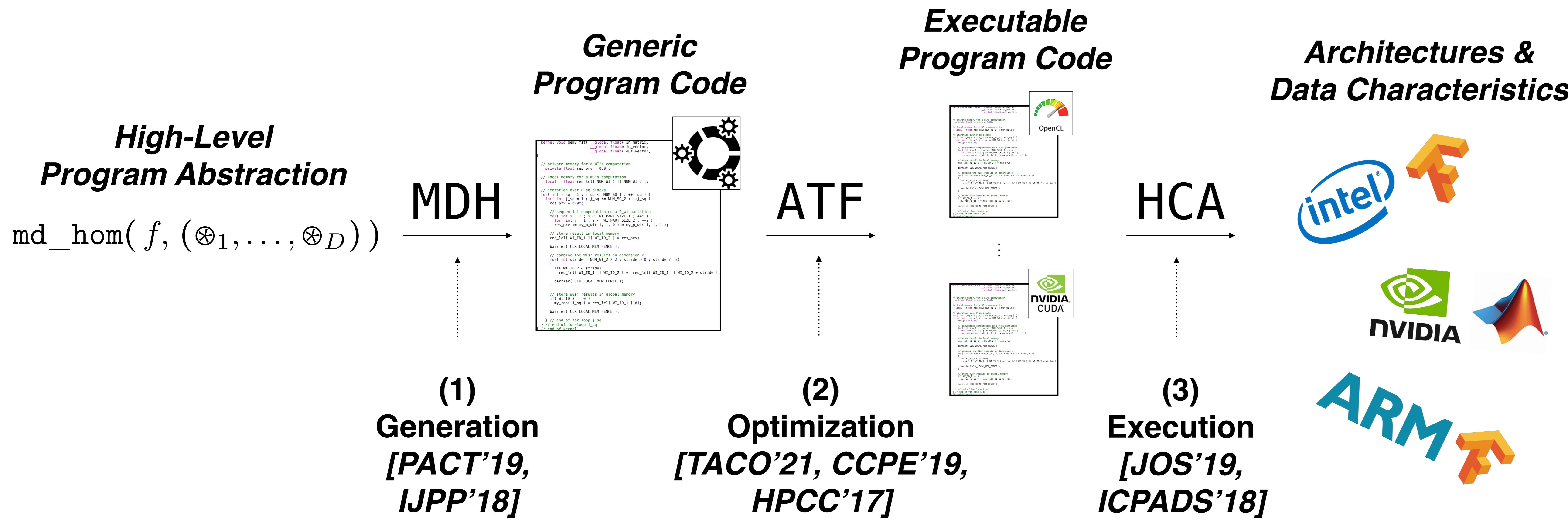
Ari Rasch, Richard Schulze, ...



THE UNIVERSITY
of EDINBURGH

Who are we?

We are the developers of the MDH+ATF+HCA approach:



A holistic approach to code *generation* (MDH) & *optimization* (ATF) & *execution* (HCA):

- (1) MDH (*Multi-Dimensional Homomorphisms*): How to generate auto-tunable code?
 - (2) ATF (*Auto-Tuning Framework*): How to auto-tune code?
 - (3) HCA (*Host Code Abstraction*): How to execute code on (distr.) multi-dev. systems?



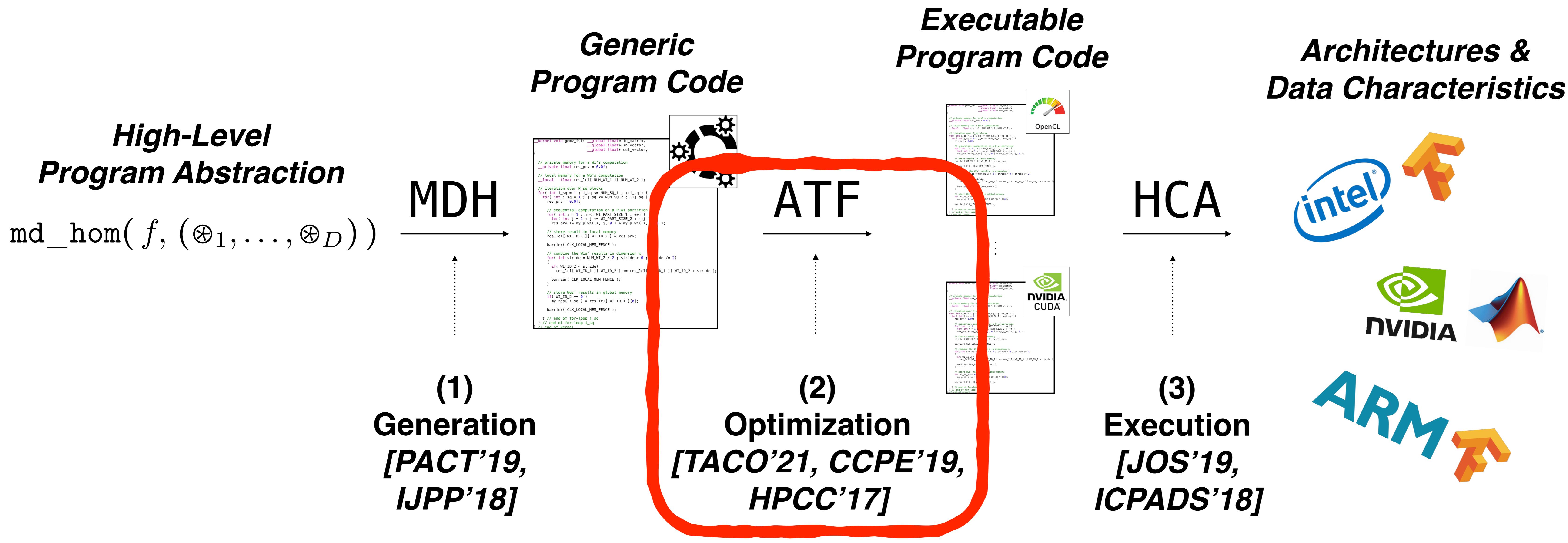
Richard Schulze



Ari Rasch

Who are we?

We are the developers of the **MDH+ATF+HCA** approach:



Richard Schulze



Ari Rasch

A holistic approach to code generation (MDH) & optimization (ATF) & execution (HCA):

- (1) MDH (Multi-Dimensional H) **our focus this week** generate auto-tunable code?
- (2) ATF (Auto-Tuning Framev **our focus this week** generate auto-tunable code?
- (3) HCA (Host Code Abstraction): How to execute code?

ATF – Yet another Auto-Tuning Framework?

We have seen *KernelTuner (KT)* & *KernelTuningToolkit (KTT)* — why do we need ATF?

ATF [1] efficiently handles
interdependencies among tuning parameters
via optimized processes to
generating & storing & exploring
the spaces of interdependent parameters.

→ We illustrate ATF by comparing it to ***CLTune*** [MCSOC'15] – the foundation of KT & KTT.

[1] Rasch, Schulze, Steuwer, Gorlatch, “Efficient Auto-Tuning of Parallel Programs with Interdependent Tuning Parameters via Auto-Tuning Framework (ATF)”, TACO’21

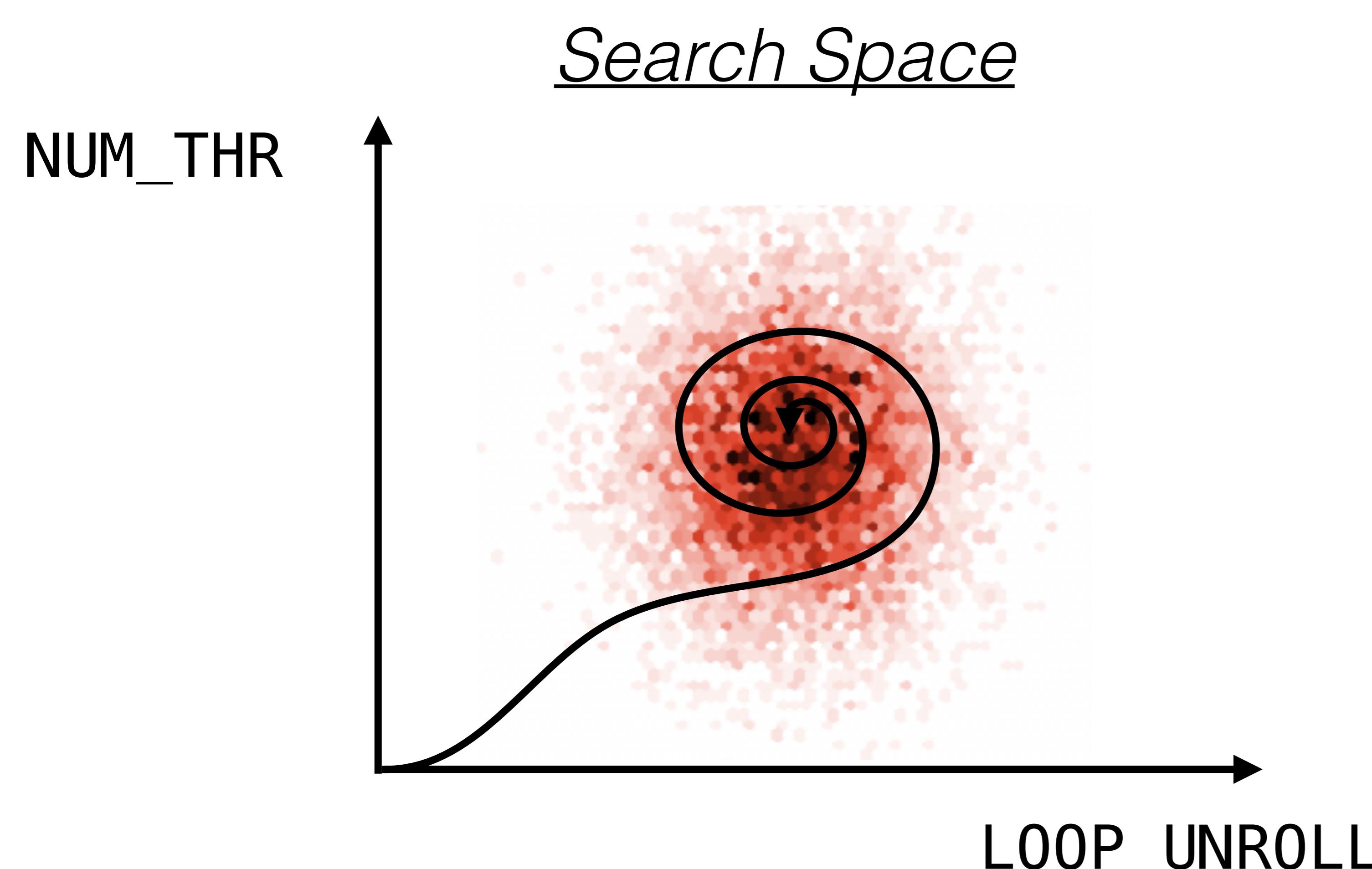
ATF – Yet another Auto-Tuning Framework?

Excursion: *Interdependent Tuning Parameters* (in a nutshell)

- Classic approaches (like OpenTuner [PACT'14]) are inherently designed toward parameters without interdependencies among them, e.g.:
 - $\text{NUM_THR} \in \text{int}$
 - $\text{LOOP_UNROLL} \in \text{int}$
- Parameter configurations: $(\text{NUM_THR}, \text{LOOP_UNROLL}) \in \text{int} \times \text{int}$
- Each configuration is considered as valid by the classic approaches!

Problem:
Applications for modern architectures often have interdependencies among their tuning parameters

What happens inside the classic tuners:



The space is explored using a search technique.

Main challenge in classic approaches

ATF – Yet another Auto-Tuning Framework?

Excursion: *Interdependent Tuning Parameters* (in a nutshell)

- We briefly illustrate *interdependent tuning parameters* using two simple, example parameters from OpenCL — a *modern* programming approach:
 - $GS \in \text{int}$ (*global size*): total numbers of threads
 - $LS \in \text{int}$ (*local size*): number of threads per group
- Parameter configurations are of the form: $(GS, LS) \in \text{int} \times \text{int}$
- Configurations are valid iff(!) they satisfy the following two constraints:
 1. LS divides GS
 2. GS is smaller than or equal to the input size N
- For example, for $N=8$, configuration $(4, 2)$ is valid, but $(4, 3)$ and $(10, 5)$ are not valid.

Modern approaches like CLTune (KT, KTT, ...) allow interdependencies among tuning parameters

ATF – Yet another Auto-Tuning Framework?

Excursion: *Interdependent Tuning Parameters* (in a nutshell)

As CLTune (pseudo)code:

```
int N = /* input size */ ;  
  
tuner.addParameter( "GS", int );  
tuner.addParameter( "LS", int );  
  
tuner.addConstraint(  
    (int GS, int LS)  
{  
    return (GS % LS == 0) && (GS <= N);  
}  
);  
  
tuner.tune( application );
```

Summary (slide 6)

- $GS \in \text{int}$
- $LS \in \text{int}$
- LS divides GS
- GS is smaller/equal N

addConstraint()
not available in classic
auto-tuning approaches
(like OpenTuner)

ATF – Yet another Auto-Tuning Framework?

Excursion: *Interdependent Tuning Parameters* (in a nutshell)

Generation

Configurations are only added to the space
iff the constraint is satisfied

```
for GS ∈ int
  for LS ∈ int
  {
    if( check_constraint() )
      add_config( GS,LS );
  }
```

Storing

Valid configurations are stored in an array

```
SP := [ (1,1) | (2,1) | (2,2) | ... ]
```

Search Space (1D Array)

Exploration

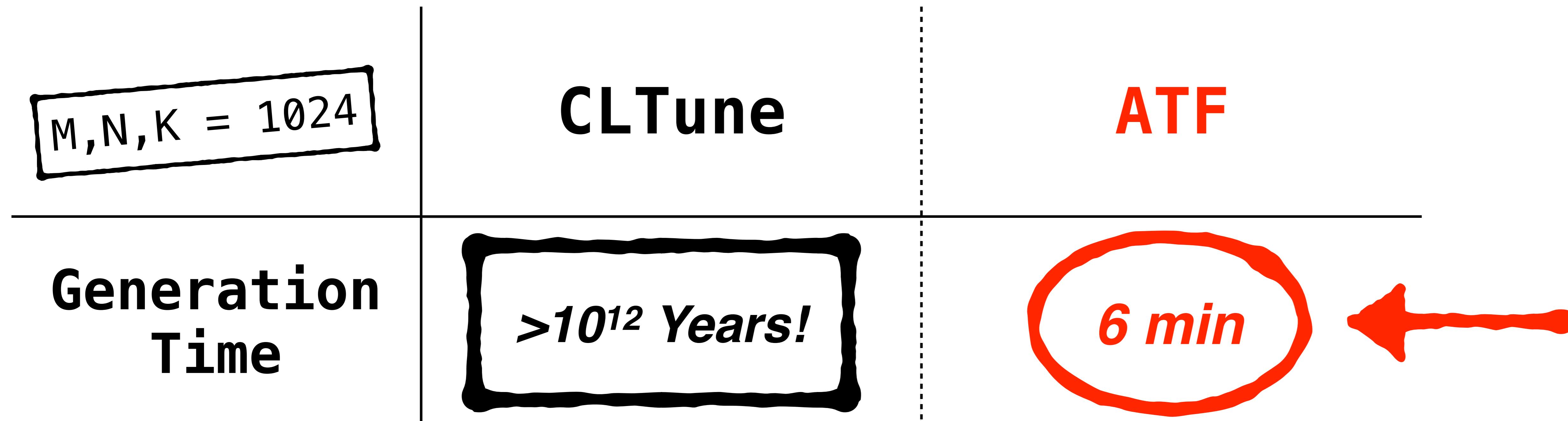
Space is explored via a search technique
(e.g., simulated annealing)

```
best_config = explore_1d_space(
  search_technique,
  SP,
  application
);
```

CLTune:
**Behind the
Scenes**

ATF – Yet another Auto-Tuning Framework?

Why ATF when we have CLTune?



Analysis of CLTune's process to *generating* search spaces, using CLTune's own GEMM example (14 tuning parameters with various interdependencies)

→ CLTune requires from user **hand-pruning** parameter ranges, e.g., using $\{16, 32, 64, 128\}$ instead of $\{1, \dots, N\}$ for tile sizes.

need to perform well for all potential target architectures and input sizes

CLTune requires **hand pruning** also for *Storing* and *Exploration* processes [TACO'21]:

CLTune's Storing Process memory intensive:
CLTune stores parameter configurations in 1D array
→ many redundancies

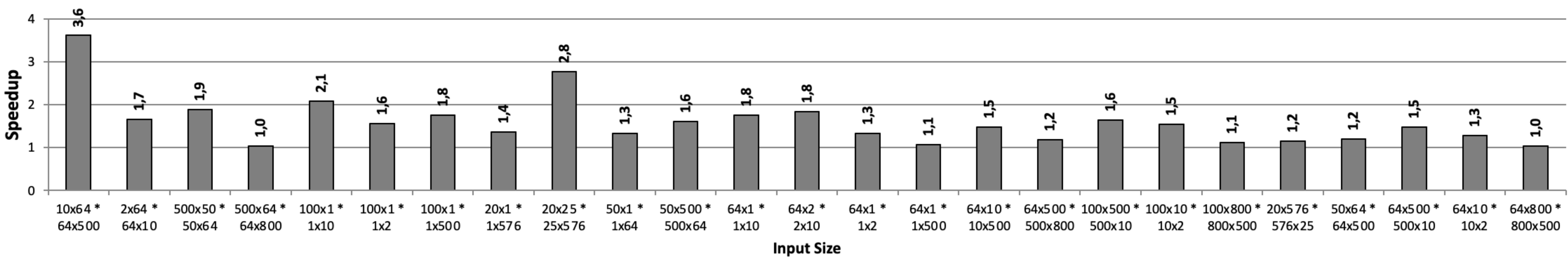
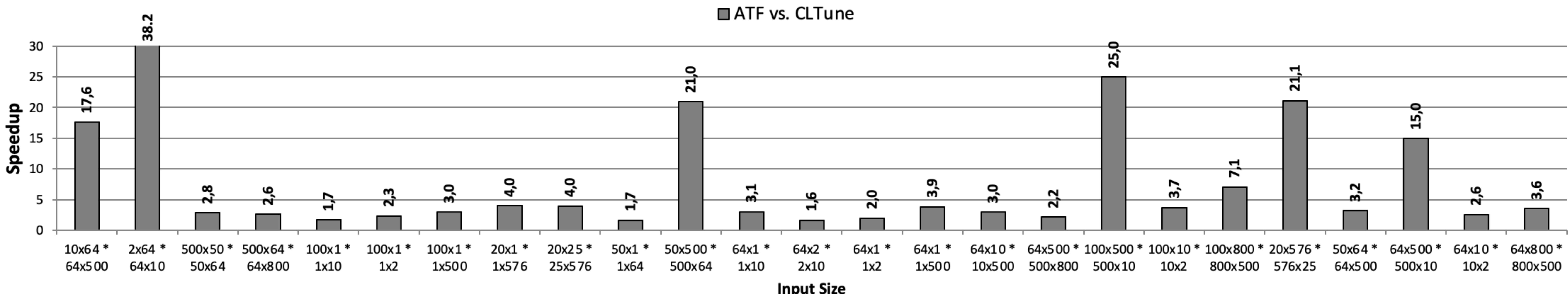
CLTune's Exploration Process hampered:
CLTune explores 1D index range of array only,
→ loses space's multi-dimensional locality information

Problem: Hand pruning search spaces is complex
→ well performing parameter configurations are often not intuitive

ATF – Yet another Auto-Tuning Framework?

Severeness of CLTune's hand pruning for its GEMM example:

Real-World
Input Sizes from DL



Speedup (higher is better) of CLTune's GEMM auto-tuned via ATF over auto-tuning via CLTune on Intel CPU (top part of figure) and NVIDIA GPU (bottom part).

By avoiding hand-pruned parameter ranges,
ATF achieves up to 38x better performance than CLTune for CLTune's GEMM example.

ATF – Yet another Auto-Tuning Framework?

In a nutshell

How does ATF achieve its efficiency for interdependent tuning parameters:

ATF introduces
parameter constraints

```
tuner.addParameter( "tp_1", T1 );
tuner.addParameter( "tp_2", T2 );
// ...
tuner.addConstraint(
    [](T1 tp_1, T2 tp_2, ... ) -> bool
    { /* ... */ }
)
```

CLTune constraints

Defined on:
search space (CLTune)
vs. ***parameters (ATF)***

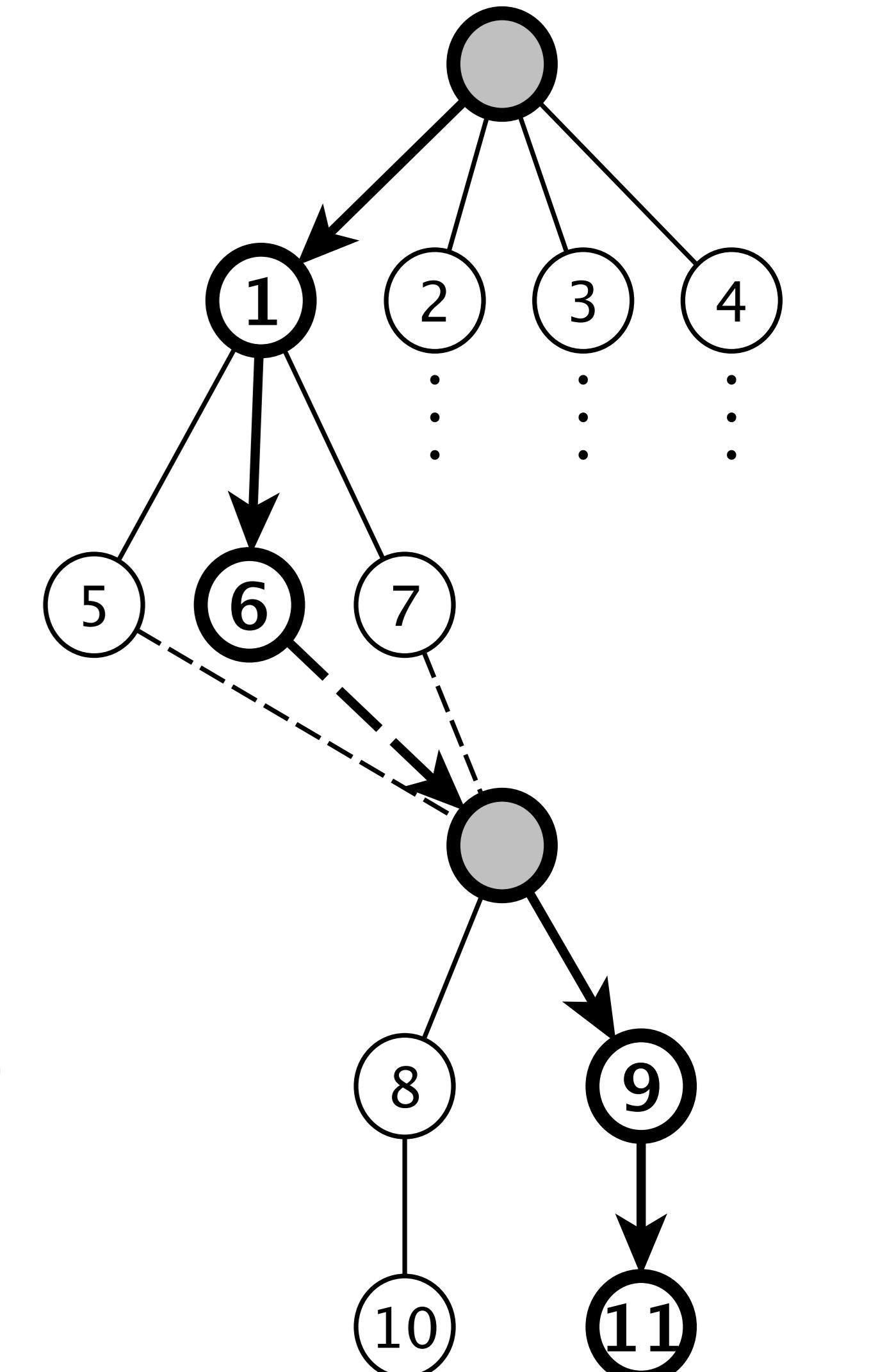
$SP := [(1,1) \mid (2,1) \mid (2,2) \mid \dots]$

CLTune search space

```
tuner.addParameter( "tp_1", T1, [](T1 tp_1) -> bool { /* ... */ } );
tuner.addParameter( "tp_2", T2, [](T2 tp_2) -> bool { /* ... */ } );
```

ATF constraints

verbose & 1D (CLTune)
vs. ***compact & nD (ATF)***



ATF introduces novel processes to
generating & ***storing*** & ***exploring***
the spaces of interdependent tuning parameters, based on its
constraint design and ***search space structure***.

In the Following

We *briefly* outline the following agenda points:

1. ATF's processes to search space:

- i. generation
- ii. storing
- iii. exploration

2. Experimental Results

3. ATF's User Interface

- i. DSL-Based (offline tuning)
- ii. GPL-based (online tuning)

3. Summary

4. Current State & Future Work

We are happy to discuss details
throughout the week

1. Generation

Basic Idea

ATF relies on *parameter constraints (PC)*, rather than *search space constraints (SC)*:

```
for ( v1 : r1 )  
  ...  
  for ( vk : rk )  
    if( sc(v1,...,vk) )  
      add_config( v1,...,vk );
```

CLTune's Approach

PCs enable generating groups of interdependent parameters independently & in parallel

```
parallel_for ( G : {G1,...,Gn} )
```

```
{
```

```
  parallel_for ( v1G : r1G )
```

Parameter Constraint (PC)

```
    if( pc1G(v1G) )
```

```
    parallel_for ( vtgG : rtgG )
```

```
      if( pctgG(vtgG) )
```

```
        for ( vtg+1G : rtg+1G )
```

```
          if( pc(vtg+1G) )
```

```
          for( vkG : rkG )
```

```
            if( pc(vkG) )
```

```
              add_config( v1G,...,vkG );
```

PCs enable generating individual groups in parallel

PCs enable checking constraints early in the loop nest

ATF's parameter constraints contain additional semantic information, enabling:

1. Generating groups of interdependent parameters *independently & in parallel*

2. Generating individual groups in *parallel*

3. Checking constraints *early*

ATF's Approach

2. Storing

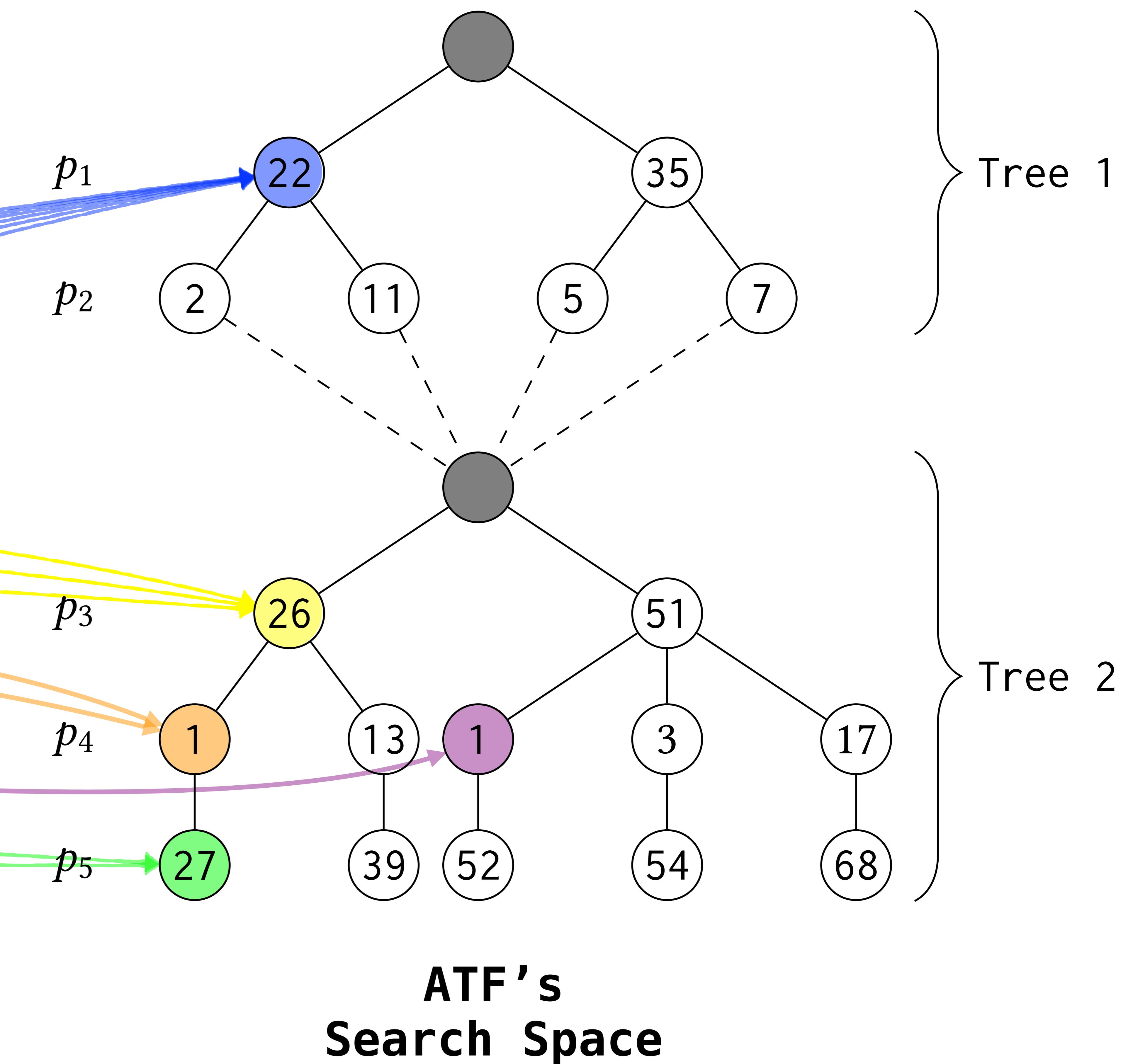
ATF relies on a new *chain-of-trees* search space structure & parameter constraints:

$p_1 := (n_1,$	{22, 35},	-)
$p_2 := (n_2,$	{2, 5, 7, 11},	- divides(n_1))
$p_3 := (n_3,$	{26, 51},	-)
$p_4 := (n_4,$	{1, 3, 13, 17},	- divides(n_3))
$p_5 := (n_5,$	{27, 39, 52, 54, 68},	equals($n_3 + n_4$))

Example Parameters

22	22	22	22	22	22
2	2	2	2	2	11
26	26	51	51	51	26
1	13	1	3	17	1
27	39	52	54	68	27

CLTune
Search Space

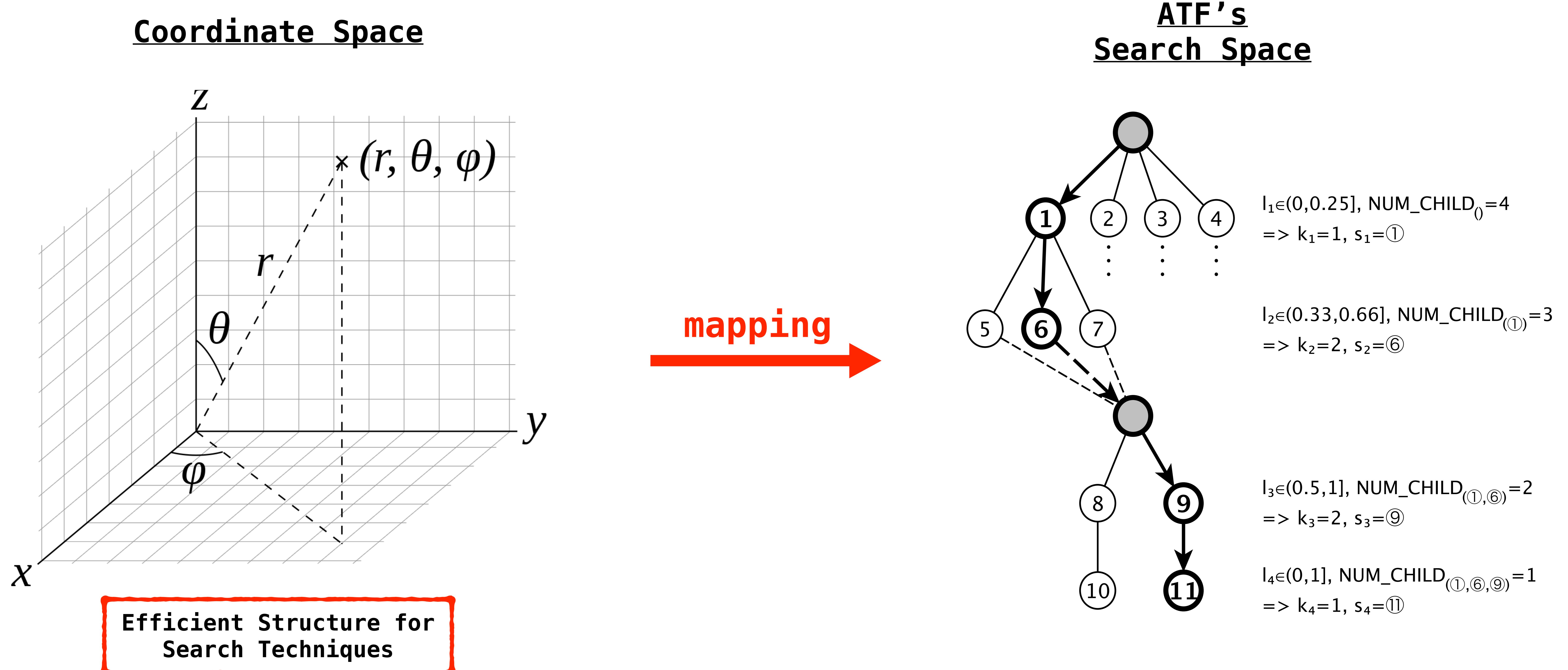


ATF's search space structure
avoids memory-intensive redundancies

3. Exploration

Basic Idea

ATF exploits its *chain-of-trees* search space structure for a multi-dimensional search:



ATF's search space structure enables
reducing the complexity of exploration to exploring a *Coordinate Space*

Experimental Evaluation

ATF is able to auto-tune important applications for **CPU & GPU** on **real-world data sets** to high performance:

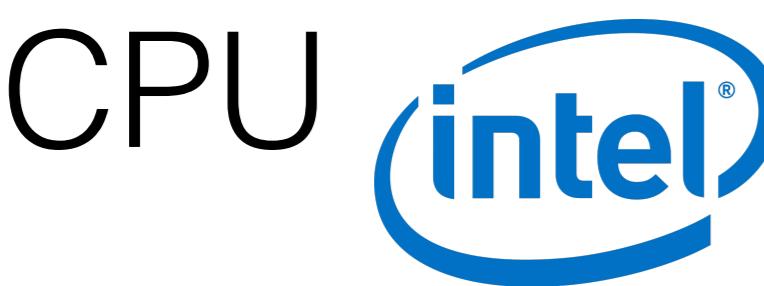
Stencil

ATF is able to auto-tune the CONV implementation in [2] to:

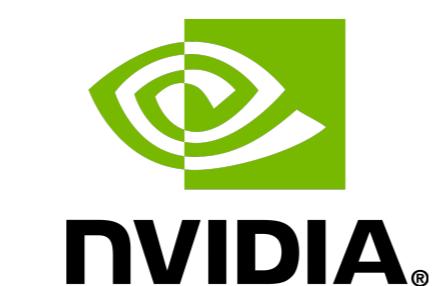
>40x higher performance
than CONV+CLTune
on CPU

>10⁴x higher performance
than CONV+CLTune
on GPU

>3x higher performance
than Intel MKL-DNN
on CPU



>15x higher performance
than NVIDIA cuDNN
on GPU



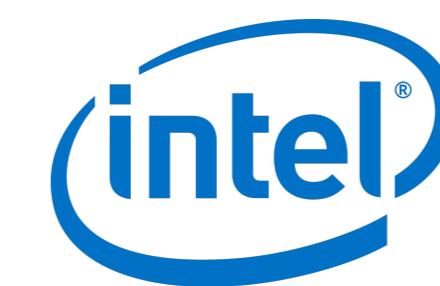
Linear Algebra

ATF is able to auto-tune the GEMM implementation in [2] to:

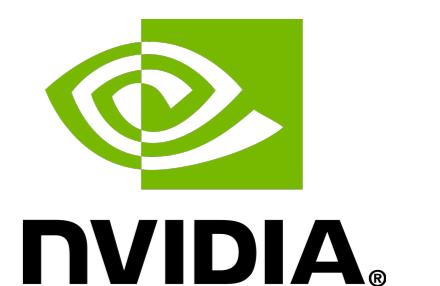
>2x higher performance
than GEMM+CLTune
on CPU

>120x higher performance
than GEMM+CLTune
on GPU

>2x higher performance
than Intel MKL
on CPU



>2x higher performance
than NVIDIA cuBLAS
on GPU



Quantum Chemistry

ATF is able to auto-tune the CCSD(T) implementation in [2] to:

>2x higher performance
than TensorComprehensions
on GPU

CLTune **fails!**
(too high search space
generation time)

Data Mining

ATF is able to auto-tune the PRL implementation in [2] to:

>1.66x higher performance
than PRL+CLTune
on CPU

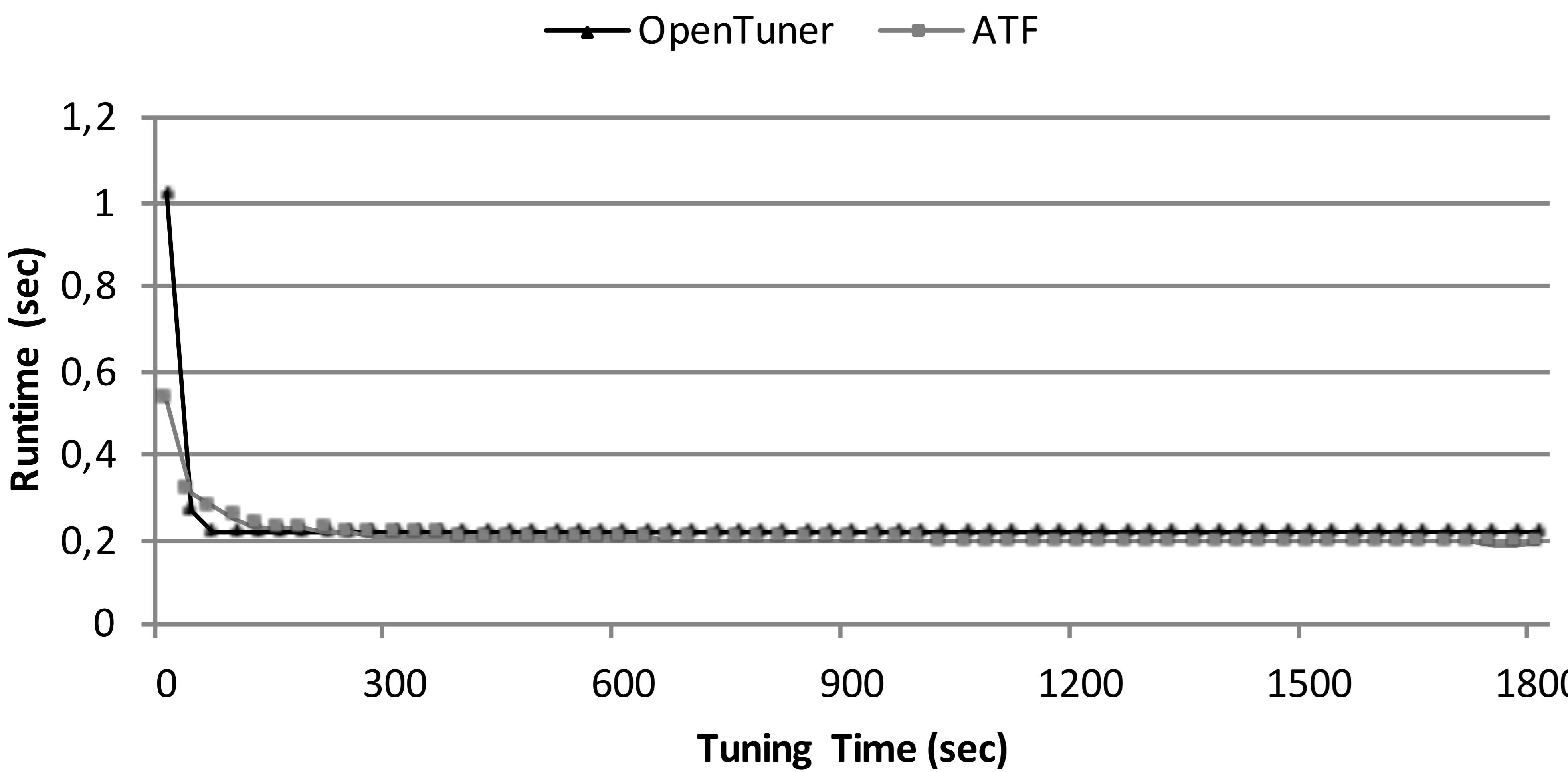
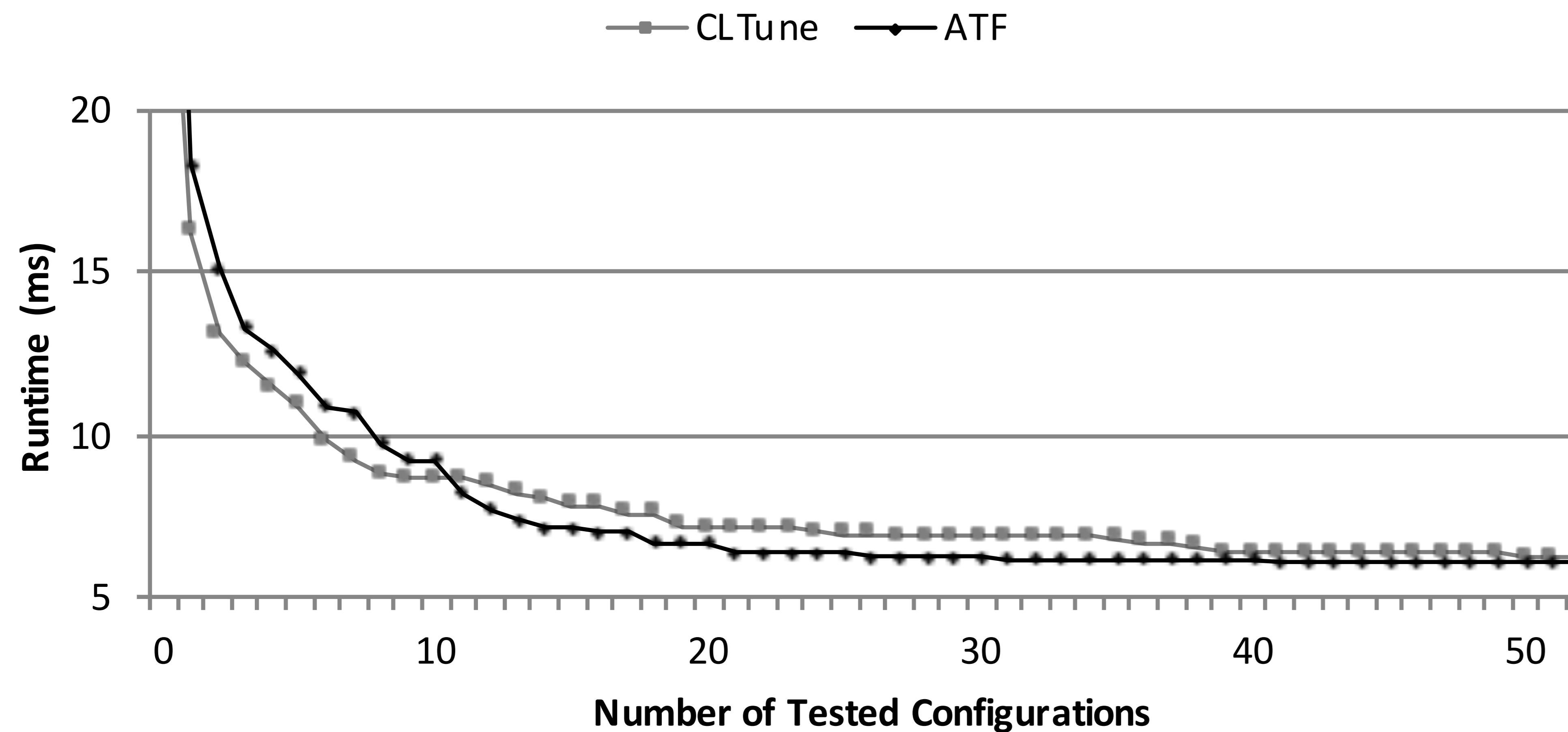
>1.07x higher performance
than PRL+CLTune
on GPU

*Auto-tunable implementations generated via our **MDH code generation approach** [2]*

Experimental Evaluation

Highlights
from CCPE'19

ATF for the favorable application classes of CLTune and OpenTuner:



**2D Convolution
(CLTune)**

**GCC Flags
(OpenTuner)**

ATF achieves tuning results of the same high quality as CLTune and OpenTuner for their favorable application classes.

Experimental Evaluation

ATF has also been successfully used for further application classes:

	Domains	Applications
1	Compiler Optimizations	GCC Flags [CCPE'19], SIMD Vectorization [ICS'19]
2	Data Mining	Probabilistic Record Linkage [PACT'19]
3	Quantum Chemistry	CCSD(T) [PACT'19]
4	Deep Learning	MCC [SC'21], BLAS [PACT'19]
5	DSL Compiler Optimizations	Lift [CGO'18], MDH [PACT'19]
6	Polyhedral Compilation	PPCG [CGO'18], Pluto [WIP]
7	Signal Processing	FFT [FHPNC'19@ICFP]
8	Stencil Computations	Conv2D, Jacobi 2D/3D, ... [CGO'18]

ATF – User Interface

ATF's user interface compared to popular auto-tuning approaches:

	Domain-specific approaches	OpenTuner	CLTune	ATF
Arbitrary Programming Language	✓			✓
Arbitrary Application Domain		✓	✓	✓
Arbitrary Tuning Objective	(✓)	✓		✓
Arbitrary Search Technique	(✓)	✓	✓	✓
Interdependent Parameters	✓		✓	✓
Automatic Cost Function Generation	✓		✓	✓

ATF's user interface combines major advantages
of existing auto-tuning approaches

ATF – User Interface

ATF's interface types:

**Offline
Auto-Tuning**

**Online
Auto-Tuning**

```
#atf::tp name      NUM_WG_1
           range     interval<int>( 1,N_1 )

#atf::tp name      NUM WI_1
           range     interval<int>( 1,N_1 )

#atf::tp name      LM_SIZE_1
           range     interval<int>( 1,N_1 )
           constraint LM_SIZE_1 <= N_1

#atf::tp name      PM_SIZE_1
           range     interval<int>( 1,N_1 )
           constraint PM_SIZE_1 <= LM_SIZE_1

#atf::run_script   "./run.sh"
#atf::compile_script "./compile.sh"

#atf::search_technique auc_bandit
#atf::search_technique duration<minutes>( 10 )

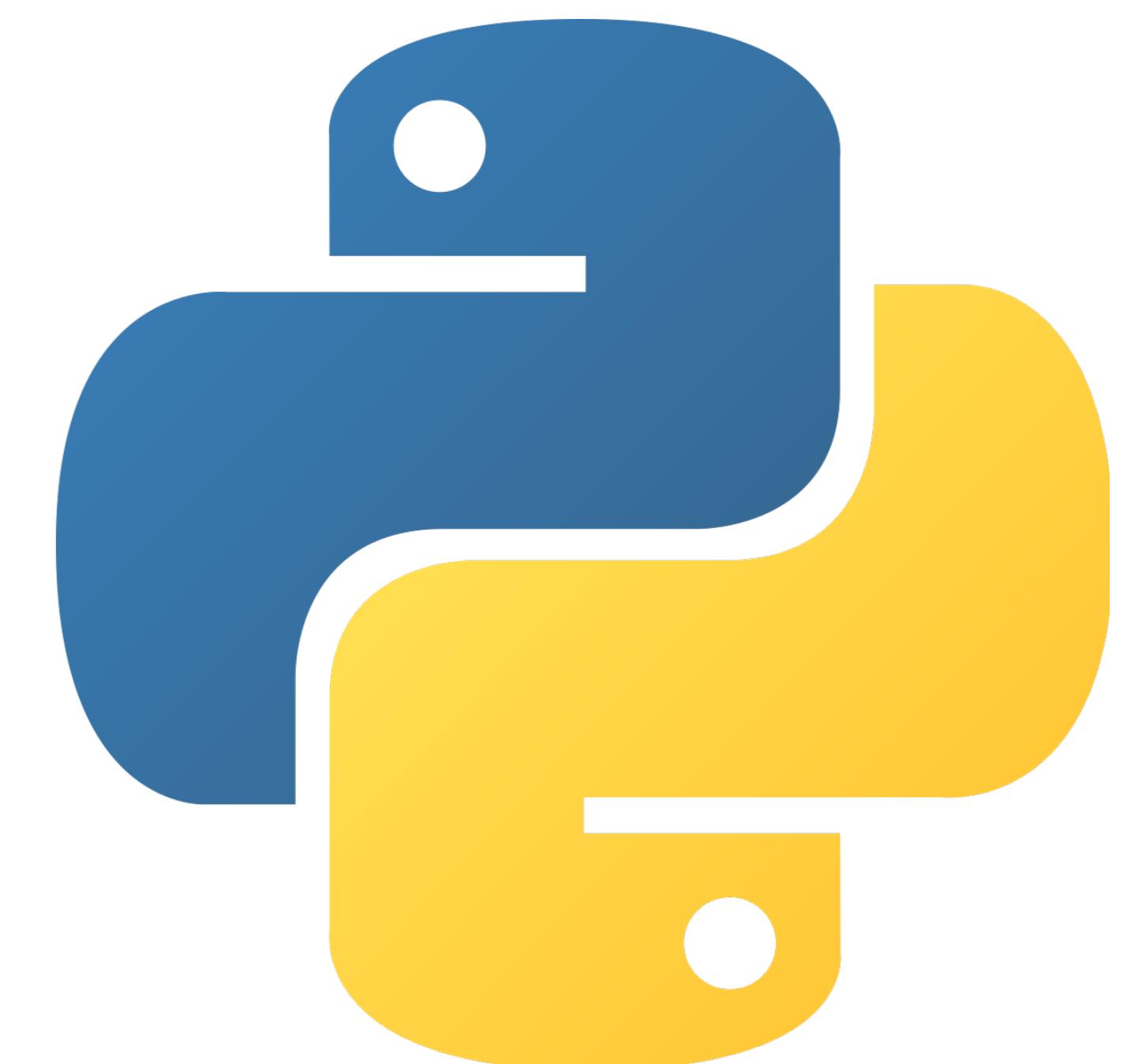
// program code
```

*ATF
Tuning Annotations*

DSL-based Interface



*ATF
C++ Interface*




*ATF
Python Interface
(WIP)*

...

GPL-based Interfaces

ATF – User Interface

ATF's **DSL-based** user interface [3] (**offline tuning**):



ATF – User Interface

ATF supports different *search techniques & abort conditions [3]*:

**ATF provides pre-implemented
search techniques:**

1. exhaustive
2. random_search
3. simulated_annealing
4. differential_evolution
5. particle_swarm
6. pattern_search
7. torczon

**Basic
Techniques**

1. round_robin
2. auc_bandit

**Meta
Techniques**

**ATF provides various
abort conditions, e.g.:**

- `duration<D>(t)`: stops tuning after time interval t ; here, D is an `std::chrono::duration` (sec, min, etc.)
- `cost(c)`: stops tuning when a configuration with a cost lower or equal than c has been found;
- `speedup<D>(s, t)`: stops tuning when within last time interval t cost could not be lowered by a factor $\geq s$;

- ...

**Further search techniques
and abort conditions
can be easily added to ATF**

ATF – User Interface

ATF automatically generates the *cost function* for the user [3]:

```
#atf::tp name      NUM_WG_1
range          interval<int>( 1,N_1 )

#atf::tp name      NUM_WI_1
range          interval<int>( 1,N_1 )

#atf::tp name      LM_SIZE_1
range          interval<int>( 1,N_1 )
constraint     LM_SIZE_1 <= N_1

#atf::tp name      PM_SIZE_1
range          interval<int>( 1,N_1 )
constraint     PM_SIZE_1 <= LM_SIZE_1

#atf::run_script   "./run.sh"
#atf::compile_script "./compile.sh"
#atf::cost_file    "./cost.txt"

#atf::search_technique auc_bandit
#atf::search_technique duration<minutes>( 10 )

// program code
```

ATF program
(from slide 22)

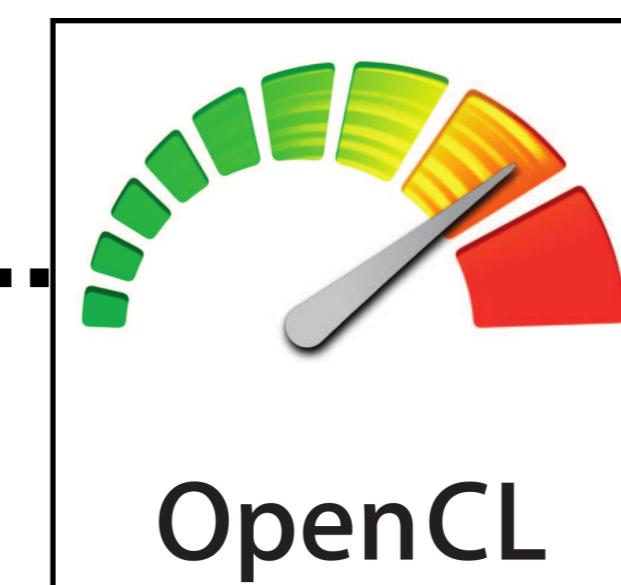
```
#atf::run_script   "./run.sh"
#atf::compile_script "./compile.sh"
#atf::cost_file    "./cost.txt"
```

Script for running program to tune
Script for compiling program (optional)
Cost File (optional)

**ATF automatically generates the cost function for
compiling & running & measuring
the program to tune**

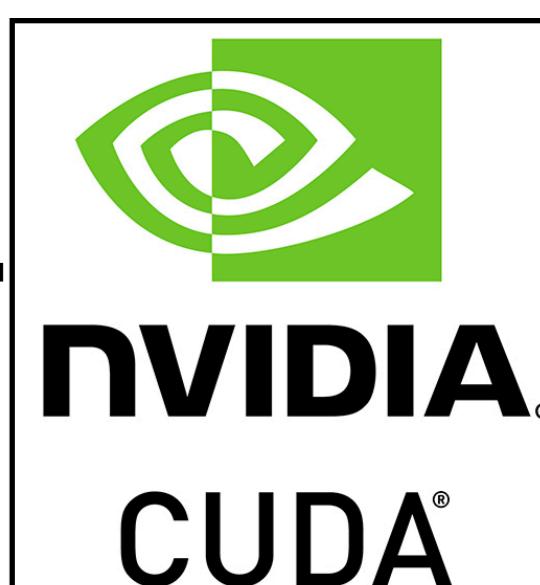
```
#atf::opencl::platform_id 0
#atf::opencl::device_id 1

#atf::opencl::input scalar<int>( N )
#atf::opencl::input scalar<float>()
#atf::opencl::input buffer<float>( N )
...
#atf::opencl::global_size N/WPT
#atf::opencl::local_size LS
```



```
#atf::cuda::device_id 1

#atf::cuda::input scalar<int>( N )
#atf::cuda::input scalar<float>()
#atf::cuda::input buffer<float>( N )
...
#atf::cuda::grid_dim (N/WPT)/LS
#atf::cuda::block_dim LS
```



**ATF provides convenience features
for automatically generating
OpenCL and CUDA host code**

ATF – User Interface

Side note — observation:

```
global_size ( ((1 + ((kSizeM - 1) / MWG))*MWG * MDIMCD) / MWG ,  
              ((1 + ((kSizeN - 1) / NWG))*NWG * NDIMCD) / NWG )  
  
local_size ( MDIMCD , NDIMCD )
```

ATF: Global & Local size

**ATF more expressive
for OpenCL than CLTune**

*Exactly as originally intended
for CLTune's GEMM in [Matsumoto et al., 2014]*

```
// Default values for global & local size  
auto id = tuner.AddKernel(gemm_fast, "gemm_fast", {kSizeM, kSizeN}, {1, 1});  
  
// ...  
  
// Sets constraints  
tuner.AddConstraint(id, DividesM, {"MWG"}); // MWG has to divide M  
tuner.AddConstraint(id, DividesN, {"NWG"}); // NWG has to divide N  
  
// ...  
  
// Modifies the thread-sizes (both global and local) based on the parameters  
tuner.MulLocalSize(id, {"MDIMC", "NDIMC"});  
tuner.MulGlobalSize(id, {"MDIMC", "NDIMC"});  
tuner.DivGlobalSize(id, {"MWG", "NWG"});
```


CLTune: Global & Local size

workaround
(not required in ATF)

**ATF is able to express exactly CLTune GEMM's global & local sizes,
thereby avoiding additional constraints (speedups >1,37x for DL sizes)**

ATF – User Interface

ATF's **DSL-based** user interface [3] (*offline tuning*):



ATF – User Interface



ATF's ***GPL-based*** user interface [4] (***online tuning***), for **C++**:

```
int main()
{
    const std::string saxpy = /* path to kernel of Listing */;
    const int      N      = /* an arbitrary input size */;

    // Step 1: Generate the Search Space
    auto WPT = atf::tp( "WPT"
                        , atf::interval<size_t>( 1,N )
                        , atf::divides( N ) );
    auto LS  = atf::tp( "LS"
                        , atf::interval<size_t>( 1,N )
                        , atf::divides( N/WPT ) );

    // Step 2: Implement a Cost Function
    auto saxpy_kernel = atf::opencl::kernel< atf::scalar<int> , // N
                                                atf::scalar<float> , // a
                                                atf::buffer<float> , // x
                                                atf::buffer<float> > // y
        ( saxpy_kernel_as_string, "saxpy" ); // kernel code & name

    auto cf_saxpy = atf::opencl::cost_function( saxpy_kernel ).platform_id( 0 ) // OpenCL platform id
                                                                .device_id( 0 ) // OpenCL device id
                                                                .inputs( atf::scalar<int>( N ) , // N
                             atf::scalar<float>() , // a
                             atf::buffer<float>( N ) , // x
                             atf::buffer<float>( N ) ) // y
                                                                .global_size( N/WPT ) // OpenCL global size
                                                                .local_size( LS ) // OpenCL local size

    // Step 3: Explore the Search Space
    auto tuning_result = atf::tuner().tuning_parameters( WPT,LS )
                            .search_technique( atf::auc_bandit() )
                            .tune( cf_saxpy , atf::evaluations(50) );
}
```

Analogous to
ATF's DSL-based interface

ATF – User Interface



ATF's ***GPL-based*** user interface (***online tuning***), for **Python**:

```
import atf

# kernel code as string
saxpy_kernel_as_string = """
kernel void saxpy( const int N, const float a, const __global float* x, __global float* y )
{
    for( int w = 0 ; w < WPT ; ++w )
    {
        const int index = w * get_global_size(0) + get_global_id(0);
        y[ index ] += a * x[ index ];
    }
}"""

# input size
N = 1000

# Step 1: Generate the Search Space
WPT = atf.tp( "WPT", atf.interval( atf.size_t , 1,N ), atf.divides( N ) )
LS = atf.tp( "LS" , atf.interval( atf.size_t , 1,N ), atf.divides( N/WPT ) )

# Step 2: Implement a Cost Function
saxpy_kernel = atf.opencl.kernel( types=[ atf.scalar(atf.int) ,      # N
                                             atf.scalar(atf.float) ; # a
                                             atf.buffer(atf.float) ; # x
                                             atf.buffer(atf.float) ], # y
                                 kernel_code=saxpy_kernel_as_string,
                                 kernel_name="saxpy" )

cf_saxpy = atf.opencl.cost_function(
    kernel=saxpy_kernel,
    platform_id=0,                      # OpenCL platform id
    device_id=0,                         # OpenCL device id
    inputs=[ atf.scalar( atf.int , N ) , # N
             atf.scalar( atf.float ) ; # a
             atf.buffer( atf.float , N ) , # x
             atf.buffer( atf.float , N ) ], # y
    global_size=N/WPT,                   # OpenCL global size
    local_size=LS                        # OpenCL local size
)

# Step 3: Explore the Search Space
tuning_result = atf.tuner(
    tuning_parameters=[ WPT, LS ],
    search_technique=atf.auc_bandit,
).tune( cf_saxpy, atf.evaluations(50) )
```

Analogous to
ATF's **DSL-based interface**

Work-in-Progress results from
Bachelor Thesis:
“Migration of Auto-Tuning Framework
from C++ to Python”,
Waldemar Gorous, 2019

Summary

ATF's **main contribution** is efficiently handling **tuning parameters with interdependencies** among them:

- ▶ ATF introduces novel processes to *generating & storing & exploring* the search spaces of interdependent parameters, based on its *parameter constraints* and the *chain-of-trees search space representation*;
- ▶ ATF's user interface is arguably simpler to use than CLTune & OpenTuner [CCPE'19, HPCC'17].
- ▶ *ATF does not contribution to search techniques (in contrast to: **KT**, **KTT**, **HyperMapper – next talk**, ...)*

GitLab

Auto-Tuning Framework

Project information

Repository

Issues

Merge requests

CI/CD

Security & Compliance

Deployments

Monitor

Infrastructure

Packages & Registries

Analytics

Wiki

Snippets

Settings

mdh-project > Auto-Tuning Framework

Auto-Tuning Framework

Project ID: 33318174

150 Commits 3 Branches 0 Tags 2.4 MB Files 2.4 MB Storage

dev atf / +

log timestamp, check coordinates for validity
Richard Schulze authored 6 hours ago

Upload File README Add LICENSE Add CHANGELOG Add CONTRIBUTING Add Kubernetes cluster

Set up CI/CD Configure Integrations

Name	Last commit	Last update
benchmark	add plot script for benchmarks	1 day ago
doc/images	Replace atf_experiments.png	3 weeks ago
examples	Update examples/feature_demonstration/...	1 day ago
include	log timestamp, check coordinates for valid...	6 hours ago
Readme.md	Update Readme.md	22 hours ago
atf.hpp	log timestamp, check coordinates for valid...	6 hours ago

Readme.md

Auto-Tuning Framework (ATF)

Auto-Tuning Framework (ATF) is a generic, general-purpose auto-tuning approach that automatically finds well-performing values of performance-critical parameters (a.k.a. tuning parameters), like the sizes of tiles and numbers of threads. ATF works for programs written in arbitrary programming languages and belonging to arbitrary application domains, and it allows tuning for arbitrary objectives (e.g., high runtime performance and/or low energy consumption).

A major feature of ATF is that it supports tuning parameters with *interdependencies* among them, e.g., the value of one tuning parameter has to be a multiple of the value of another tuning parameter. For this, ATF introduces novel process to *generating, storing, and exploring*.

ATF available on GitLab
for all workshop attendees:

<https://gitlab.com/mdh-project/atf>

ATF – Current State & Future Work

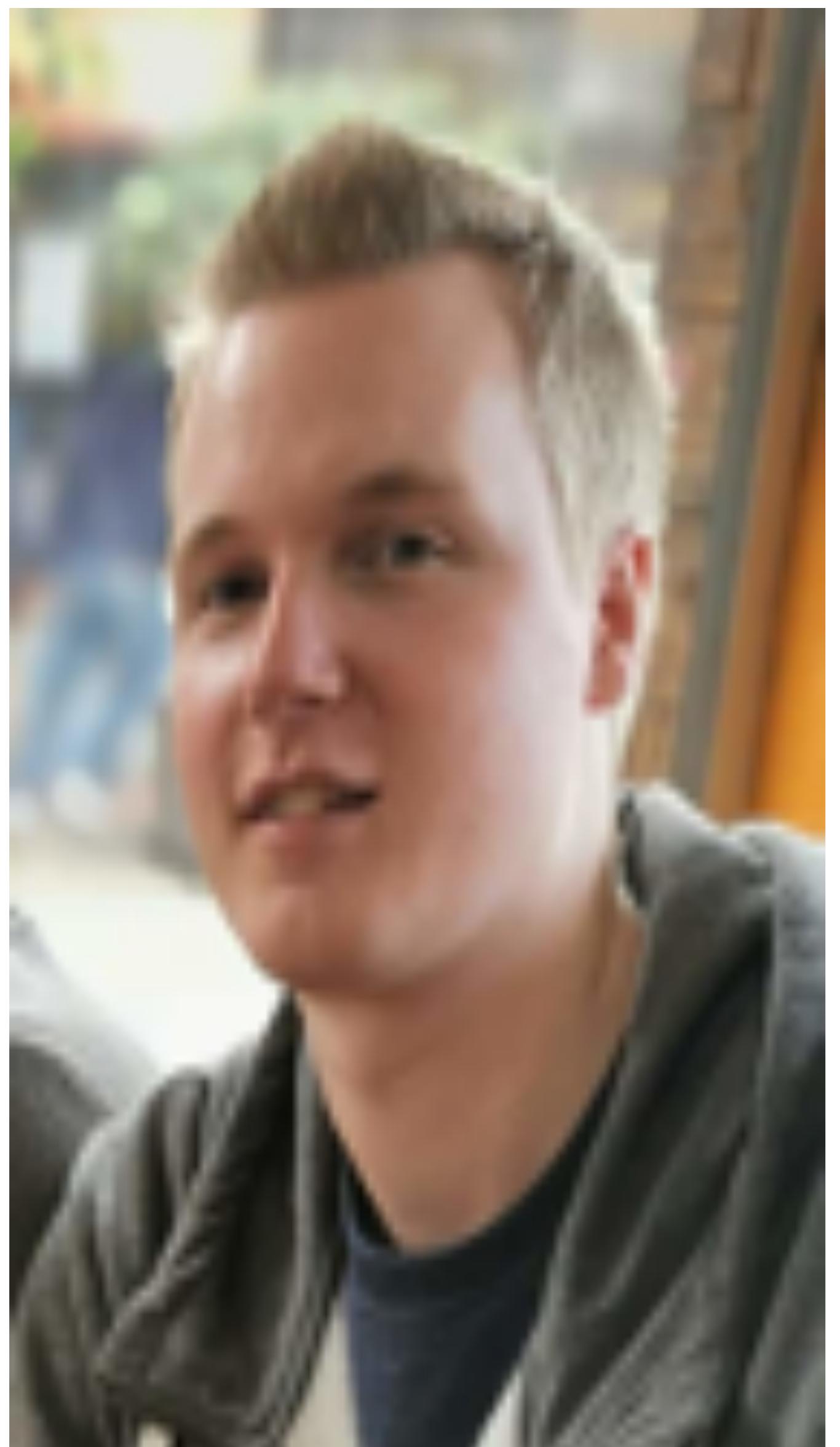
Current State:

- ATF's DSL-based user interface currently not maintained (we use its C++ interface)
- ATF's implementation is a proof of concept (error handling can be improved, etc)
- ...

Future Work:

- getting the best of both worlds:
 1. *efficient search techniques*, as in **KT**, **KTT**, **HyperMapper**, etc;
 2. *efficient search space generation & storing & exploring*, as in **ATF**.
- using different search techniques for search space parts with different characteristics (as in Pfaffe et. al [ICS'19], **HyperMapper**, etc): *categorical*, *ordinal*, *real*, etc.
- improving multi-objective auto-tuning toward *Pareto optimality* (as in **HyperMapper**)
- ...

Thanks for listening!



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Questions?