



Using Program Shaping to Parallelise an Erlang Multi-Agent System

Adam D. Barwell, **Chris Brown**, Kevin Hammond University of St Andrews

Aleksander Byrski, Wojciech Turek
AGH University of Science and Technology

Lambda Days 2016, 18 February 2016







RePhrase Project: Refactoring Parallel Heterogeneous Software – a Software Engineering Approach (ICT-644235), 2015-2018, €3.6M budget

8 Partners, 6 European countries
UK, Spain, Italy, Austria, Hungary, Israel

Coordinated by St Andrews

















What are we trying to achieve?







Parallelism and Concurrency





Key Software Engineering Challenges



- Testing, Verification and Debugging
 - Automatic Test Generation, race condition detection, ...
- Software Quality Assurance
 - New Standards are needed
 - Cross-Platform Approaches
- Deployment on heterogeneous platforms
 - e.g. CPU/GPU, APU, manycore, FPGA
 - efficient scheduling of multiple applications
- Maintainability and Software Evolution
 - Change parallelism structure
 - Adapt to varying numbers of cores and processor types





The RePhrase Approach



Software development phase	RePhrase tools	
Requirements capture	Parallel requirements capturing methods	
Design	Refactoring tool, patterns	
Implementation and debugging	Refactoring tool, pattern implementations, adaptivity tool	
Testing and verification	Parallel testing framework, parallel verification tool, failure detection tool,	
	property violation detection tool	
Deployment	Adaptivity tools, refactoring tool	
Maintenance and evolution	Refactoring tool, adaptivity tools, patterns, pattern implementations, quality	
	assurance tool	





Program Shaping



- Restructuring (legacy) programs to enable the introduction of parallelism
- Might include:
 - Removing certain types of side effects
 - Encapsulating computations into components
 - Eliminating unnecessary dependencies
- Currently ad hoc
 - Often non-trivial
 - Requires intimate knowledge of code, language, and parallelism
- Refactoring techniques can be used to automate the process



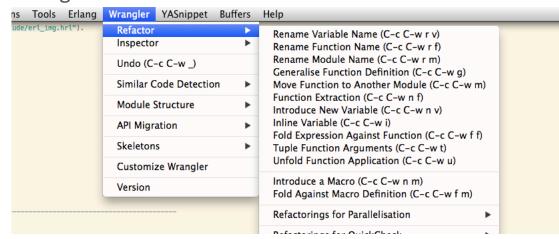




Refactoring for Parallelism



- Conditional, source-to-source transformation that preserves functional correctness
- Refactoring tools to help automate this process
 - Semi-automatic transformation avoids introducing errors
 - Developer input allows a wider range of possible transformations
- Wrangler
 - Extensible refactoring tool for Erlang
 - Built-in collection of refactorings
 - API allows user creation of refactorings
 - Originally created by the University of Kent







Algorithmic Skeletons



- High-level abstraction of some common pattern of parallelism
- Composable and nestable
- Language independent
- Need only problem-specific sequential code
 - plus any skeletal parameters
- Implemented and collected in algorithmic skeleton libraries

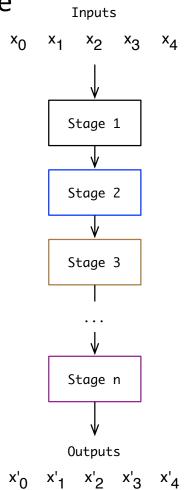


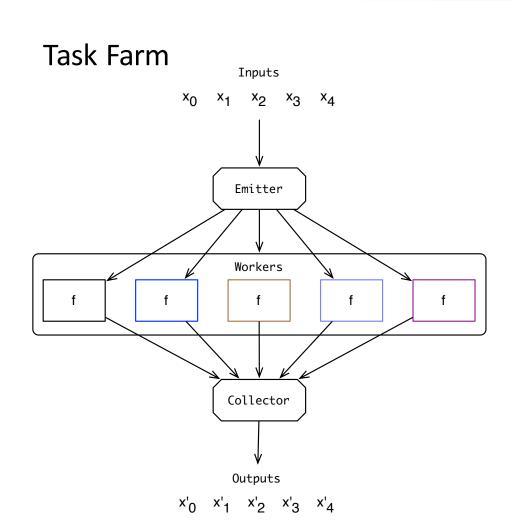


Example Skeletons



Pipeline



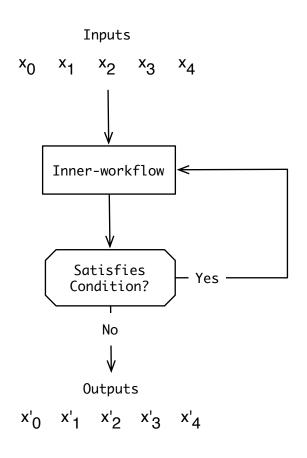




Example Skeletons



Feedback





Skel



- An algorithmic skeleton library for Erlang
 - Pipeline, Task Farm, Feedback, and more
 - Hybrid skeletons for both CPU and GPU targets
 - http://skel.weebly.com







Skel Example



```
lists:map(fun worker/1, Input)
```

Introduce Farm

skel:do({farm, fun worker/1, NW}, Input)





Multi-Agent Systems



- An agent is an intelligent, autonomous entity that solves some problem or subtask
- A Multi-Agent System (MAS) brings two or more agents together to solve some complex problem, e.g. flood forecasting
- An Evolutionary Multi-Agent System (EMAS) combines multiagent systems with evolutionary algorithms
- Highly parallel: agents are independent

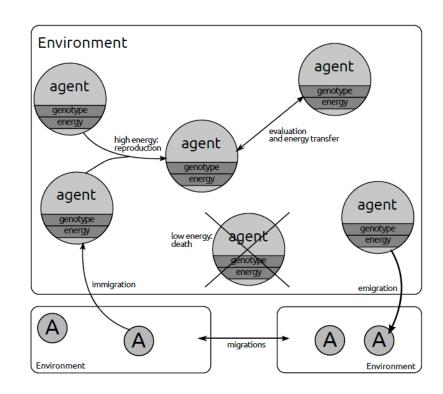




Evolutionary Multi-Agent Systems (EMAS)

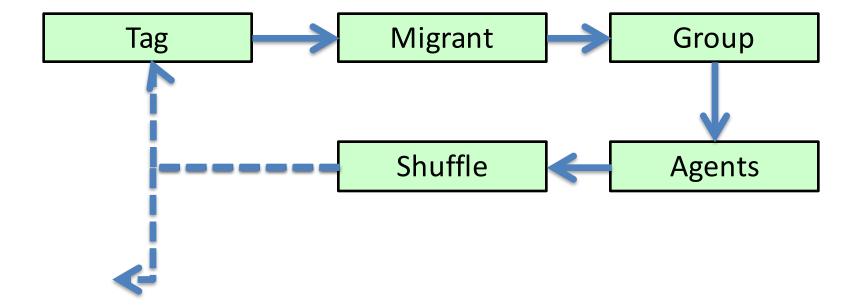


- Meta-heuristic approach for optimization
 - universal optimization algorithm (formally proven)
- Explicit hybridisation of agentoriented and evolutionary computing
- Agents
 - contain genotypes and energy as a means for distributed selection
 - located on evolutionary islands
 - perform actions (death, reproduction, migration, fight)



EMAS – Basic Structure









EMAS Code

```
loop(Islands, Time, SP, Cf) ->
    Tag = fun(Island) ->
                   [{
                     mas_misc_util:behaviour_proxy(
                       Agent, SP, Cf),
                     Agent | | Agent <- Island]
          end.
    Groups = [mas_misc_util:group_by(Tag(I)) || I <- Islands],</pre>
    Migrants = [seq_migrate(lists:keyfind(migration, 1, Island), Nr)
                 || {Island, Nr} <-
                         lists:zip(Groups,
                                   lists:seq(
                                     length(Groups)))],
    NewGroups = [[mas_misc_util:meeting_proxy(
                    Activity,
                    mas_sequential,
                    SP,
                    Cf) || Activity <- I]
                 || I <- Groups],
    WithMigrants = append(
                     lists:flatten(Migrants),
                     NewGroups),
    NewIslands = [mas_misc_util:shuffle(lists:flatten(I))
                  || I <- WithMigrants],
    case os:timestamp() < Time of</pre>
        true ->
            loop(NewIslands, Time, SP, Cf);
        false ->
            NewIslands
    end.
```

Parallelising EMAS



- We could introduce a task farm for each list comprehension...
- But this is inefficient:
 - Farm creation creates overhead
 - Not all tasks in the system are large enough for parallelism
 - The function loops until some condition is met, compounding the above issues
- It is better to express the parallel behaviour in a single operation
 - Knowing the full structure allows minimisation of overheads
- However, as the code stands, we cannot introduce this operation





Program Shaping Refactorings



- First divide up the sequential code into atomic components
 - which we can then rearrange
- We will use the following refactorings:
 - Extract Composition Function
 - Compose Maps
 - Intro Func
 - Intro Farm
 - Intro Feedback
 - Intro Skel

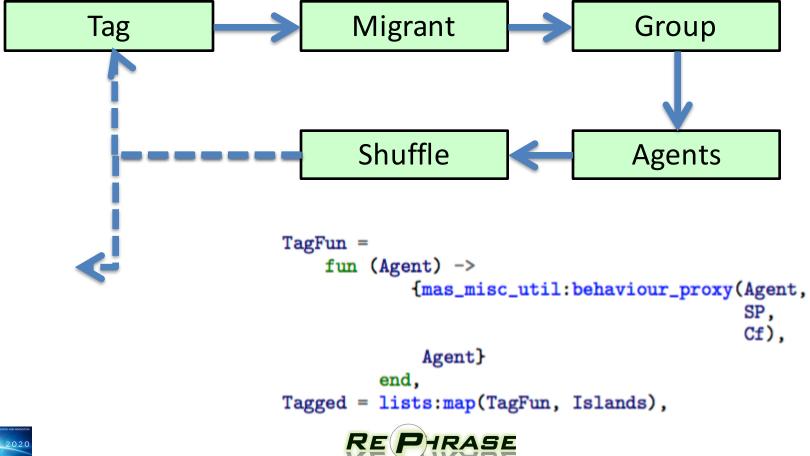




EMAS – Program Shaping



First, encapsulate code for stages into blocks using Extract **Composition Function**





Stage 1

Split and format the tagging, grouping, and migrating stages into components using Extract Composition **Function**

```
TagFun =
    fun (Agent) ->
            {mas_misc_util:behaviour_proxy(Agent,
                                            SP,
                                            Cf),
             Agent}
         end.
Tagged = lists:map(TagFun, Islands),
GroupFun = fun (I) -> mas_misc_util:group_by(I) end,
Groups = lists:map(GroupFun, Tagged),
MigrantFun =
    fun ({Island, Nr}) ->
            seq_migrate(lists:keyfind(migration,
                                       1, Island),
                        Nr)
             end.
Migrants = lists:map(MigrantFun,
                     lists:zip(Groups,
                                lists:seq(
                                  length(Groups)))),
```

Stage 2

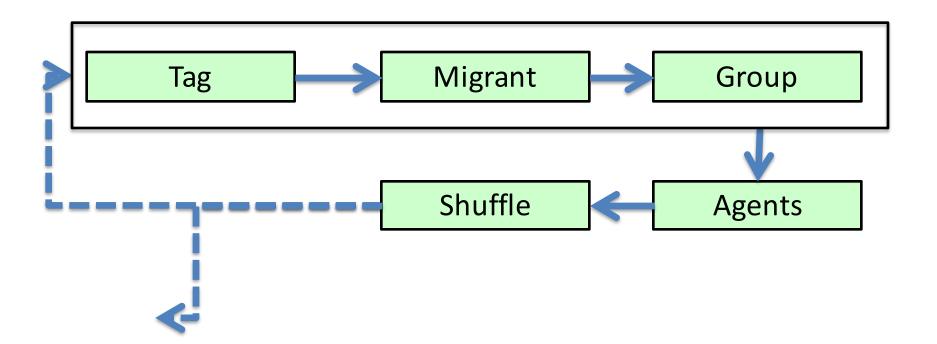
Since these stages can be easily composed, using the classical Inline Method refactoring, we inline the migration function.

```
MigrantFun =
    fun ({{migration, Agents}, From}) ->
            Destinations =
                [{mas_topology:getDestination(From),
                  Agent } | | Agent <-Agents],
            mas_misc_util:group_by(Destinations);
        (OtherAgent) -> OtherAgent
    end,
Migrants = lists:map(MigrantFun,
                     lists:zip(Groups,
                                lists:seq(
                                  length(Groups)))),
```

EMAS – Program Shaping



Next, group together stages and remove dependencies using Compose Maps







Stage 3

We now compose the tagging, grouping, and migrating stages using Compose Maps

```
TagFun =
    fun (Agent) ->
            {mas_misc_util:behaviour_proxy(Agent,
                                            SP.
                                            Cf),
             Agent}
    end.
GroupFun = fun (I) -> mas_misc_util:group_by(I) end,
MigrantFun =
    fun ({{migration, Agents}, From}) ->
            Destinations =
                [{mas_topology:getDestination(From),
                  Agent } | | Agent <-Agents],
            mas_misc_util:group_by(Destinations);
        (OtherAgent) -> OtherAgent
    end.
TGM = fun(Agents) ->
              Tagged = lists:map(TagFun, Agents),
              Migrants = lists:map(MigrantFun, Tagged),
              GroupFun(Migrants)
      end.
TGMs = lists:map(TGM, Islands),
```

Stage 4

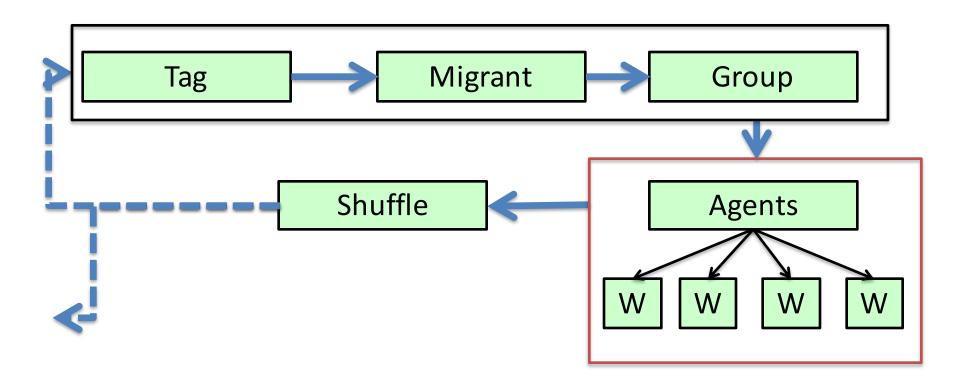
```
We expose
functions as
components
for new
groups and
new islands
stages using
Extract
Composition
Function
```

```
NewGroupsFunInnerFun =
    fun (Activity) ->
            mas_misc_util:meeting_proxy(Activity,
                                        mas_sequential,
                                        SP.
                                        Cf)
                       end.
NewGroupsFun =
    fun (I) ->
            lists:map(NewGroupsFunInnerFun, I)
    end.
NewGroups = lists:map(NewGroupsFun, TGMs),
NewIslandsFun =
    fun (I) ->
             mas_misc_util:shuffle(lists:flatten(I))
    end,
NewIslands = lists:map(NewIslandsFun, NewGroups),
```

EMAS – Program Shaping



Next, create a farm of agents using Intro Farm







Stage 5

We now start to arrange these individual components, ready to be passed to Skel. We apply Intro Func over TGM and NewGroupsInnerFun expressions. We also introduce a farm over NewGroupsFun

TGMs = {func, TGM},

```
Work =
    {func.
     fun (Activity) ->
             mas_misc_util:meeting_proxy(Activity,
                                          mas_farm,
                                          SP.
                                          Cf)
     end}.
Map = {farm, [Work], Cf#config.skel_workers},
NewGroups = lists:map(NewGroupsFun, TGMs),
Shuffle =
    fun (I) ->
            mas_misc_util:shuffle(lists:flatten(I))
    end.
NewIslands = lists:map(Shuffle, NewGroups),
```

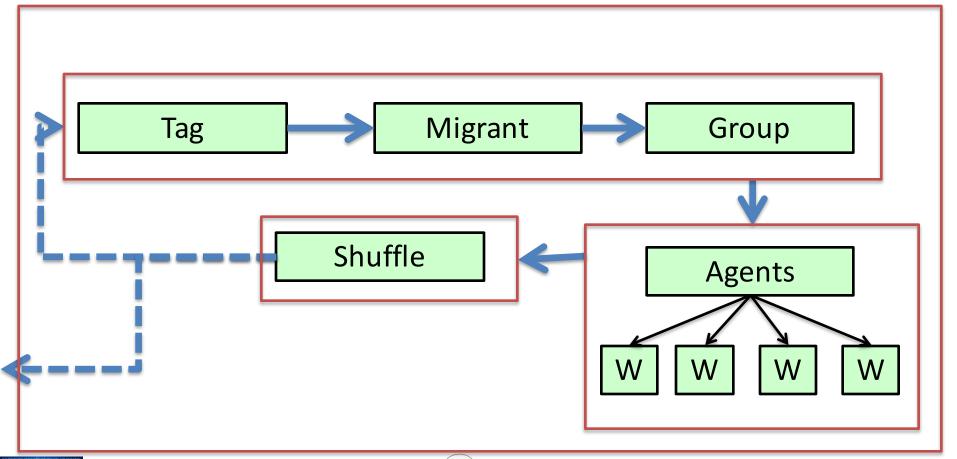
Stage 6

We use Intro Func on Shuffle, completing the skeletons needed for Skel, and use Intro Skel over NewIslands and NewGroups

EMAS – Program Shaping



Finally, use Intro Feedback using the **pipeline** and **farm** as components



Stage 7 (End)

We use Intro Feedback to fold the outer loop into the Skel invocation, improving efficiency.

This completes the shaping and parallelisation process.

```
loop(Islands, Time, SP, Cf) ->
    EndTime =
        mas_misc_util:add_miliseconds(os:timestamp(), Time),
    TagFun =
        fun (Agent) ->
                {mas_misc_util:behaviour_proxy(Agent,
                                                Cf), Agent}
        end.
    GroupFun = fun (I) -> mas_misc_util:group_by(I) end,
    MigrantFun =
        fun ({{migration, Agents}, From}) ->
                Destinations =
                    [{mas_topology:getDestination(From),
                      Agent} || Agent <-Agents],
                mas_misc_util:group_by(Destinations);
            (OtherAgent) -> OtherAgent
        end.
    TGM = tgm(TagFun, GroupFun, MigrantFun),
    TGMs = {func, TGM}.
    Work = {func.
            fun (Activity) ->
                    mas_misc_util:meeting_proxy(
                      Activity,
                      mas_farm,
                      SP.
                      Cf)
            end}.
    Map = {farm, [Work], Cf#config.skel_workers},
    Shuffle = {func,
               fun (I) ->
                       mas_misc_util:shuffle(lists:flatten(I))
               end}.
    Pipe = {pipe, [TGMs, Map, Shuffle]},
    Constraint = fun (_) -> os:timestamp() < Time end,
    FinalIslands = skel:do([{farm,
                              [{feedback, [Pipe], Constraint}],
                             Cf#config.skel_workers}],
                              [Islands]).
```

EMAS Code (Shaped)

```
loop(Islands, Time, SP, Cf) ->
    EndTime =
        mas_misc_util:add_miliseconds(os:timestamp(), Time),
    TagFun =
        fun (Agent) ->
                {mas_misc_util:behaviour_proxy(Agent,
                                                Cf), Agent}
        end,
    GroupFun = fun (I) -> mas_misc_util:group_by(I) end,
    MigrantFun =
        fun ({{migration, Agents}, From}) ->
                Destinations =
                    [{mas_topology:getDestination(From),
                      Agent} || Agent <-Agents],
                mas_misc_util:group_by(Destinations);
            (OtherAgent) -> OtherAgent
        end.
   TGM = tgm(TagFun, GroupFun, MigrantFun),
    TGMs = \{func, TGM\},\
    Work = {func,
            fun (Activity) ->
                    mas_misc_util:meeting_proxy(
                      Activity,
                      mas_farm,
                      SP.
                      Cf)
            end}.
    Map = {farm, [Work], Cf#config.skel_workers},
    Shuffle = {func,
               fun (I) ->
                       mas_misc_util:shuffle(lists:flatten(I))
               end}.
    Pipe = {pipe, [TGMs, Map, Shuffle]},
    Constraint = fun (_) -> os:timestamp() < Time end,
    FinalIslands = skel:do([{farm,
                             [{feedback, [Pipe], Constraint}],
                             Cf#config.skel_workers}],
                             [Islands]).
```

Results



- We compare our shaped EMAS to two other versions:
 - Concurrent: follows good Erlang practice for writing concurrent code;
 - 2. **Hybrid**: designed and manually tuned to give the best possible performance for the EMAS algorithm
- Two different benchmarks
 - continuous (Rastrigin)
 - discrete (Low Autocorrelation Binary Sequences)
- Tested on a 64-core machine at AGH, Poland (ZEUS)
 - 4 x AMD Opteron 6276, 16 2.3GHz cores

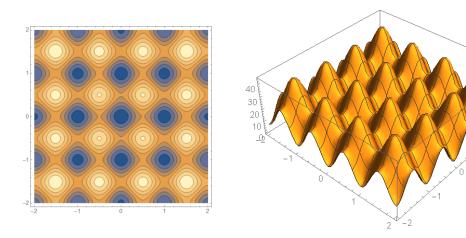




Optimization Benchmark



- Find optimum of Rastrigin function in dimensions n=100
 - $f(x) = 10n + \sum_{i=1}^{n} (x_i^2 10\cos(2\pi x_i))$
 - One of classic global optimization benchmark functions
- Example: Rastrigin function in two dimensions

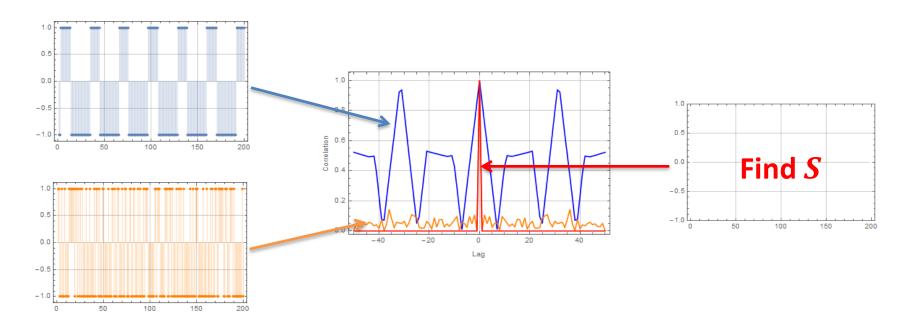


LABS

Low-Autocorrelation Binary Sequences

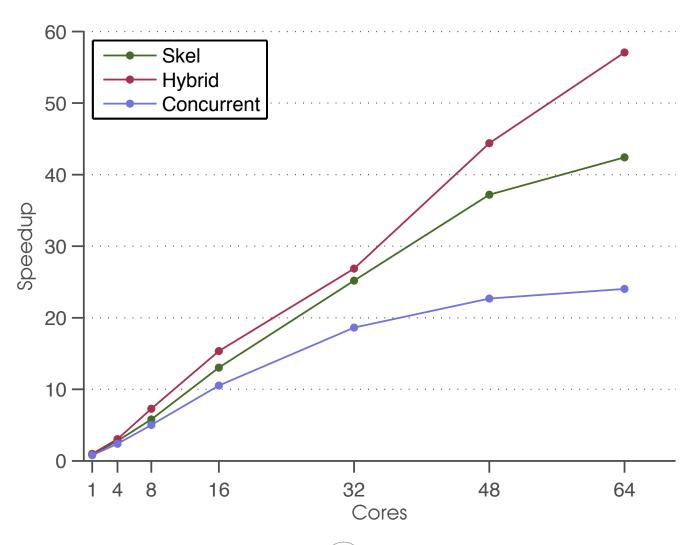


- $S = s_1 s_2 \dots s_L$: binary sequence of length L and $s_i \in \{-1, +1\}$
- Aperiodic Autocorrelation with lag $k:C_k(S)=\sum_{i=1}^{L-k}s_is_{i+k}$
- Minimize $E(S) = \sum_{k=1}^{L-1} C_k^2(S)$ with respect to S



Speedups for Rastrigin Function



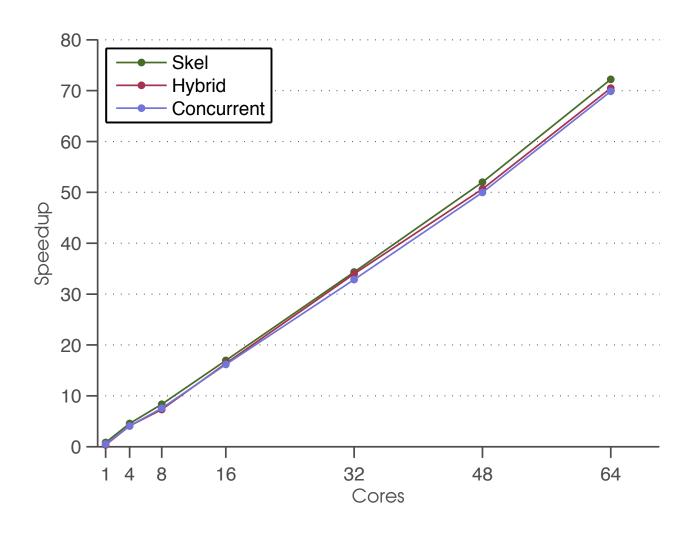






Speedups for LABS









EMAS: Coding Efficiency



Effort for implementing the generic EMAS backends

	Lines of Code	Effort in Days
Sequential	85	10
Hybrid	129	2
Concurrent	353	7
SKEL	100	1

Conclusions



- We have introduced novel program shaping techniques
 - applied to an Erlang implementation of an Evolutionary Multi-Agent System, a real-world use case
- Obtain speedups of 45x for Rastrigrin and 70x for LABS
 - at minimal programmer effort
- Applicable to other languages, e.g. C++, Java, ...





Future Work



- Other use cases, and further evaluate the effectiveness of the approach; e.g. the Dialyzer
- Expansion of our library of program shaping techniques
- Incorporate static analysis techniques to further automate the program shaping process, at the same time reducing the burden on the programmer
- Demonstrate the applicability of this approach to use cases in languages other than Erlang







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