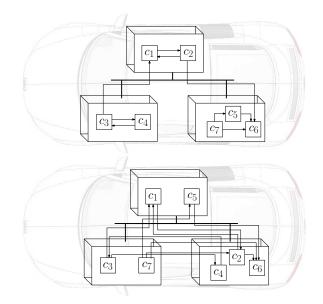
# Adaptive Neighbourhood Search for the Component Deployment Problem

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# Software architecture design



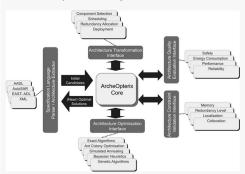
#### Software quality optimisation



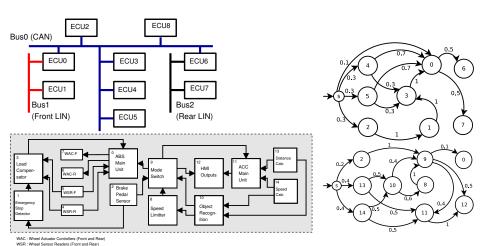
The design of embedded systems involves several important decisions, such as choosing the software components to use, and deciding how to deploy them into the hardware platform. These decisions affect the quality attributes of the system, such as reliability and safety. Embedded systems are becoming more complex with many design options to choose from. We have automated this task.

#### **ArcheOpterix**

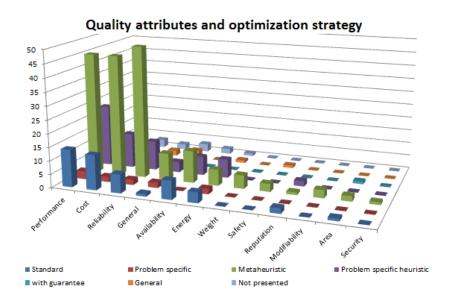
ArcheOpterix is a generic platform for modelling, evaluating and optimising embedded systems. The main modules of ArcheOpterix are shown in the figure below.



## Component Deployment Problem



## **Problem Difficulty**



## Variables in Component Deployment optimization

**Software components**  $\mathbb{C} = \{c_1, c_2, \dots, c_n\}$  (discrete fix number)

- memory size szi of the i-th component (KiloBytes)
- workload  $w_i \rightarrow$  million instructions
- ▶ initialization probability *q<sub>i</sub>* to start from the *i*-th component

#### Interaction between software components

- ▶ data size ds<sub>ij</sub> → the amount of data sent between c<sub>i</sub> and c<sub>j</sub> in a single communication event
- $lackbox{ execution probability } p_{ij} 
  ightarrow ext{the probability that calling } c_i ext{ ends}$  with a call of  $c_j$

**Hardware hosts**  $\mathbb{H} = \{h_1, h_2, \dots, h_m\}$  (discrete fix number)

- memory capacity cp of each host (KiloBytes)
- ▶ processing speed ps → instruction process capacity of hardware unit
- ightharpoonup failure rate fr o the probability of a single hardware failure

#### Hardware links $\mathbb{N} = \{n_1, n_2, \dots, n_s\}$

- ▶ data rate  $dr_{ii}$  → data transmission rate of a bus
- lacktriangledown failure rate  $fr_{ii} 
  ightarrow$  data communication error at each link



## Component Deployment optimization

 $D=\{d\mid d:\mathbb{C}\to\mathbb{H}\}\to$  the set of all function assignment components to hardware resources

#### Objective function

- ▶ reliability of a component i is  $R_i = e^{-\operatorname{fr}_{d(c_i)} \cdot \frac{\operatorname{wl}_i}{\operatorname{ps}_{d(c_i)}}}$
- ▶ reliability of communication between components c<sub>i</sub> and c<sub>j</sub>

$$R_{ij} = e^{-\mathsf{fr}_{d(c_i)d(c_j)} \cdot \frac{\mathsf{ds}_{ij}}{\mathsf{dr}_{d(c_i)d(c_j)}}}$$

 $\blacktriangleright$  expected number of visits for each component  $v:C\to\mathbb{R}_{\geq 0}$ 

$$v_i = q_i + \sum_{j \in \mathcal{I}} v_j \cdot p_{ji}$$

• expected number of visits of network links  $v: C \times C \to \mathbb{R}_{\geq 0}$ 

$$v_{ij} = v_i \cdot p_{ij}$$

▶ reliability of a deployment architecture  $d \in D$  is  $R = \prod_{i=1}^{n} R_i^{v_i} \prod_{i \in I} R_{ii}^{v_{ij}}$ 



## Representation and Pertubator operators

### Solution representation ← string of (component,host) pairs

$c_1$	$c_2$	$c_3$	$c_4$	$c_5$	$c_6$	$c_7$	$c_8$
$h_1$	$h_3$	$h_3$	$h_2$	$h_4$	$h_1$	$h_2$	$h_4$

#### OneFlip operator

$c_1$	$c_2$	$c_3$	$c_4$	$c_5$	$c_6$	$c_7$	$c_8$
$h_1$	$h_3$	$h_3$	$h_2$	$h_4$	$h_1$	$h_2$	$h_4$

## **kOpt** operator, where k = 2

$c_1$	$c_2$	$c_3$	$c_4$	$c_5$	$c_6$	$c_7$	$c_8$
$h_1$	$h_3$	$h_3$	$h_2$	$h_4$	$h_1$	$h_2$	$h_4$

**Perturb**  $\leftarrow$  **kOpt** with random value for  $k \in \{1, ..., n\}$ 

## Adaptive Neighbourhood Search

```
procedure AN
   S = \text{RANDOMLYALLOCATE}(C, H)
   N = \text{SelectNeighbourhoodOperator}(P(N))
   if N == OneFlip then
      Q(N) = OneFlip(S)
   end if
   if N == KOpt then
      Q(N) = KOpt(S, k)
   end if
   if N == Perturb then
      Q(N)=Perturb(S)
   end if
   ReportFeedback(Q(N))
   RETURN(S)
end procedure
```

The selection of the neighbourhood operator is a fitness proportionate method  $P(N) = \alpha P(N) + (1 - \alpha)Q(N)$ .

## Neighbourhood operators for the Component Deployment

```
procedure ONEFLIP(S)
2.
      S^* = S
      localOptimum = True
      for all c < C do
 4:
          h = \text{RANDOMLYSELECTHOST}(H)
          S' = AssignComponentToHost(S, c, h)
6:
          EVALUATE(S')
         if S' > S^* then
8:
             S^* = S'
         end if
10:
      end for
      improvement = FITNESSDIFFERENCE(S, S^*)
12:
      S = S^*
      RETURN(improvement)
14:
   end procedure
```

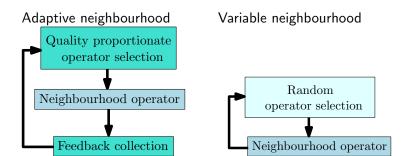
## Deterministic neighbourhood search

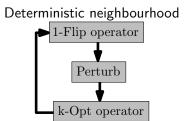
#### procedure DN

- 2: S = RandomlyAllocate(C, H)OneFlip(S)
- 4: PERTURB(S) KOPT(S, k)
- 6: RETURN(S) end procedure

## Variable neighbourhood search

```
procedure VN
      S = \text{RANDOMLYALLOCATE}(C, H)
      r = RANDOM([0,1])
   if r > 0.5 then
 4:
          ONEFLIP(S)
      end if
6:
      if r < 0.5 then
         KOpt(S, k)
8:
      end if
10:
      p = RANDOM([0,1])
      if p < 0.01 then
          Perturb(S)
12:
      end if
      RETURN(S)
14:
   end procedure
```





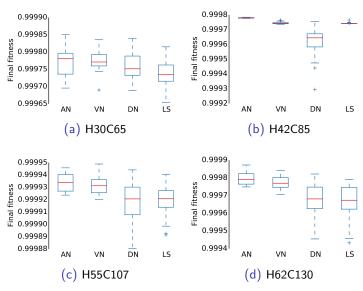
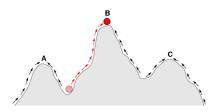


Figure: 30 trials (KS-test p-value<0.05).

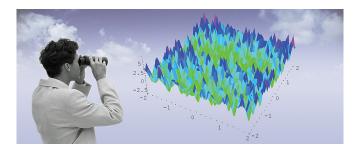
## Alternative optimization methods

- Genetic algorithms
- Simulated annealing
- Steepest descent
- Ant colony optimisation
- ► Hill-climbing
- ▶ and the list goes on



#### Remarks

- ► The nature of the search-space is the key factor determining the performance of the optimisation algorithm,
- Define/characterise the search-space,
- Analyse what makes problems difficult,
- Guide the optimisation process
  - Select the right search strategy



# Future work: difficulty of local search

