# A Framework for Decomposition in Integer Programming

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Column Generation Workshop 2008 Aussois, France

## Outline

Traditional Decomposition Methods

Integrated Decomposition Methods

DECOMP Framework

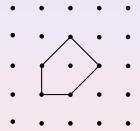
Basic Idea: By leveraging our ability to solve the optimization/separation problem for a relaxation, we can improve the bound yielded by the LP relaxation.

$$z_{IP} = \min_{x \in \mathbb{Z}^n} \{ c^\top x \mid A' x \ge b', A'' x \ge b'' \}$$

 $z_{LP} = \min_{x \in \mathbb{R}^n} \{ c^\top x \mid A' x \ge b', A'' x \ge b'' \}$ 

 $z_D = \min_{x \in \mathbb{Z}^d} \{ c^{\top} x \mid A'' x \ge b'' \}$ 

$$z_{IP} \ge z_D \ge z_{LP}$$



$$\mathcal{P} = \operatorname{conv}\{x \in \mathbb{Z}^n \mid A'x \ge b', A''x \ge b''\}$$

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ullet  $OPT(c,\mathcal{P})$  and  $SEP(x,\mathcal{P})$  are "hard"

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 $\bullet$   $\mathcal{Q}''$  can be represented explicitly (description has polynomial size

 $\circ$   $\mathcal{P}'$  must be represented implicitly (description has exponential size)

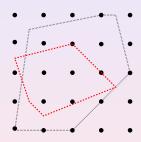
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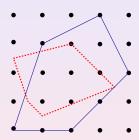
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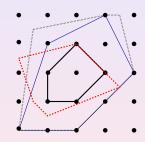
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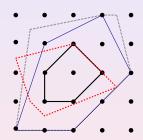
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# Example - Traveling Salesman Problem

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$$\begin{array}{lcl} x(\delta(\{u\})) & = & 2 & \forall u \in V \\ x(E(S)) & \leq & |S|-1 & \forall S \subset V, \ 3 \leq |S| \leq |V|-1 \\ x_e \in \{0,1\} & \forall e \in E \end{array}$$



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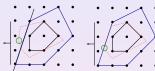
#### 2-Matching

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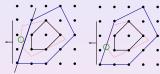
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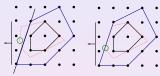


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The Lagrangian Method (LD) iteratively solves a Lagrangian relaxation subproblem.

### Common Threads

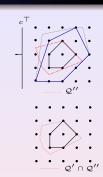
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- Traditional decomposition-based bounding methods contain two primary steps
  - Master Problem: Update the primal/dual solution information
  - Subproblem: Update the approximation of  $\mathcal{P}'$ :  $SEP(x,\mathcal{P}')$  or  $OPT(c,\mathcal{P}')$
- Integrated decomposition methods further improve the bound by considering two implicitly defined polyhedra whose descriptions are iteratively refined.
  - Price and Cut (PC)
  - Relax and Cut (RC)
  - Decompose and Cut (DC)



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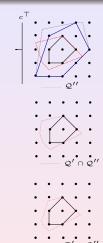
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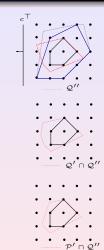
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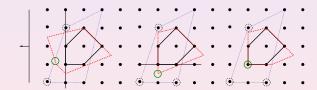


### Price and Cut

Price and Cut: Use DW as the bounding method. If we let  $\mathcal{F}' = \mathcal{P}' \cap \mathbb{Z}^n$ , then

$$z_{DW} = \min_{\lambda \in \mathbb{R}_+^{\mathcal{F}'}} \{ c^\top (\sum_{s \in \mathcal{F}'} s \lambda_s) : A''(\sum_{s \in \mathcal{F}'} s \lambda_s) \ge b'', \sum_{s \in \mathcal{F}'} \lambda_s = 1 \}$$

- As in the cutting plane method, separate  $\hat{x} = \sum_{s \in \mathcal{F}'} s \hat{\lambda}_s$  from  $\mathcal{P}$  and add cuts to [A'', b''].
- Advantage: Cut generation takes place in the space of the compact formulation (the original space), maintaining the structure of the column generation subproblem.

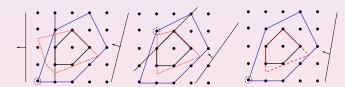


### Relax and Cut

Relax and Cut: Use LD as the bounding method.

$$z_{LD} = \max_{u \in \mathbb{R}^n_+} \min_{s \in \mathcal{F}'} \{ (\boldsymbol{c}^\top - \boldsymbol{u}^\top \boldsymbol{A}'') s + \boldsymbol{u}^\top \boldsymbol{b}'' \}$$

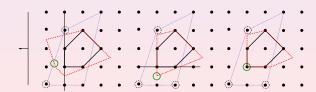
- In each iteration, separate  $\hat{s} \in \operatorname{argmin}_{s \in \mathcal{F}'} \{ (c^\top u^\top A'') s + u^\top b'' \}$ , a solution to the Lagrangian relaxation.
- Advantage: It is often much easier to separate a member of  $\mathcal{F}'$  from  $\mathcal{P}$  than an arbitrary real vector, such as  $\hat{x}$ .



# Decompose and Cut

Decompose and Cut: As in price and cut, use DW as the bounding method, but use the decomposition obtained in each iteration to generate improving inequalities as in RC.

- Rather than (or in addition to) separating  $\hat{x}$ , separate each member of  $D = \{s \in \mathcal{F}' \mid \hat{\lambda}_s > 0\}.$
- As with RC, it is often much easier to separate a member of  $\mathcal{F}'$  from  $\mathcal{P}$  than an arbitrary real vector, such as  $\hat{x}$ .
- RC only gives us one member of  $\mathcal{F}'$  to separate, while PC gives us a set, one of which must be violated by any inequality violated by  $\hat{x}$ .
- We can also use CP and decompose the fractional solution obtained in each iteration into a convex combination of members of  $\mathcal{F}'$  and apply the same technique.
- In case this decomposition fails, we still get a Farkas cut for free.

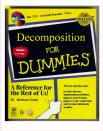


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**DECOMP** is a software framework that provides a virtual sandbox for testing and comparing various decomposition-based bounding methods.

- It's very difficult to compare the variants discussed here in a controlled w.
- The method for separation/optimization over  $\mathcal{P}'$  is the primary application-dependent component of any of these algorithms.
- DECOMP abstracts the common, generic elements of these methods.
- Key: The user defines application-specific components in the space of the compact formulation
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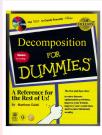


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- Built on top of the OSI interface, so easy to swap solvers (simplex to interior point).
- Can utilize CGL cuts in all algorithms (since cut generation is always done in the origina space).
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- Flexible parameter interface: command line, param file, direct call overrides.
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- One interface to all default algorithms: CP/DC, DW, LD, PC, RC.
- Automatic reformulation allows users to deal with variables and constraints in the original space.
- Built on top of the OSI interface, so easy to swap solvers (simplex to interior point).
- Can utilize CGL cuts in all algorithms (since cut generation is always done in the original space).
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- The base class DecompApp provides an interface for the user to define the application-specific components of their algorithm.
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    DecompApp::APPcreateModel(). Define [A", b"] and [A', b'] (optional).
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    TSP 2-Match: [A", b"] define trivial subtour constraints.
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    TSP: do we have a feasible tour?
    DecompApp::APPsolveRelaxed(). Provide a subroutine for OPT(c, P').
    This is optional as well, if [A', b'] is defined (it will call the built in IP solver, currently CBC).
    TSP 1-Tree: provide a solver for 1-tree.
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- All other methods have appropriate defaults but are virtual and may be overridden.
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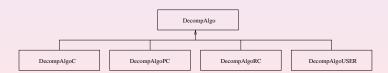
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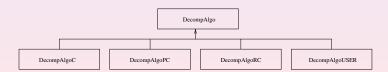
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  - Alternative methods for solving the master LP in DW, such as interior point methods or ACCPN
    - Add stabilization to the dual updates in LD, as in bundle methods.
    - For LD, replace subgradient with Volume, providing an approximate primal solution
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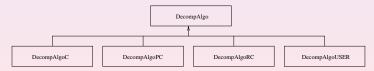
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# DECOMP - TSP Example

#### TSP\_Main

```
int main(int argc, char ** argv){
  //create the utility class for parsing parameters
  UtilParameters utilParam(argc, argv);
  //create the user application (a DecompApp)
  TSP_DecompApp tsp(utilParam);
  tsp.createModel();
  //create the algorithm(s) (a DecompAlgo)
  DecompAlgoC * cut = new DecompAlgoC(&tsp , &utilParam );
  DecompAlgoPC * pcOneTree = new DecompAlgoPC(&tsp , &utilParam ,
                                                TSP_DecompApp::MODEL_ONETREE);
  DecompAlgoPC * pcTwoMatch = new DecompAlgoPC(&tsp , &utilParam ,
                                                TSP_DecompApp::MODEL_TWOMATCH);
  DecompAlgoRC * rcOneTree = new DecompAlgoRC(&tsp , &utilParam ,
                                                TSP_DecompApp::MODEL_ONETREE);
  DecompAlgoRC * rcTwoMatch = new DecompAlgoRC(&tsp , &utilParam ,
                                                TSP_DecompApp::MODEL_TWOMATCH);
  //create the driver AlpsDecomp model
  AlpsDecompModel alpsModel(utilParam):
  //install the algorithms
  //alpsModel.addDecompAlgo(cut):
  alpsModel.addDecompAlgo(pcOneTree):
  //solve
  alpsModel.solve():
```

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- DECOMP provides an easy-to-use framework for comparing and developing various decomposition-based bounding methods
  - The user only needs to define the components based on the compact formulation (irrespective or algorithm).
- The interface to ALPS allows us to investigate large-scale problems on distributed networks
- The code is open-source, currently released under CPL and will soon be available through the COIN-OR project repository www.coin-or.org.
- Related publications
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