A Framework for Decomposition in Integer Programming

Ted Ralphs¹ Matthew Galati²

¹COR@L Lab, Department of Industrial and Systems Engineering, Lehigh University

 2 SAS Institute, Advanced Analytics, Operations Research R & D



University of Newcastle, 1 June 2009 Newcastle, Australia

Outline

Traditional Decomposition Methods

Integrated Decomposition Methods

DECOMP Framework

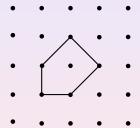
Basic Idea: By leveraging our ability to solve the optimization/separation problem for a relaxation, we can improve the bound yielded by the LP relaxation.

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

 $z_D = \min_{c \in \mathcal{D}'} \{ c^\top x \mid A'' x \ge b'' \}$

$$z_{IP} > z_D > z_{IP}$$



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

Assumptions

- \bullet $OPT(c, \mathcal{P})$ and $SEP(x, \mathcal{P})$ are "hard".
- ullet $OPT(c, \mathcal{P}')$ and $SEP(x, \mathcal{P}')$ are "easy"
- \bullet \mathcal{Q}'' can be represented explicitly (description has polynomial size
- \mathcal{P}' must be represented implicitly (description has exponential size)

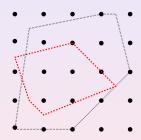
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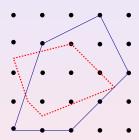
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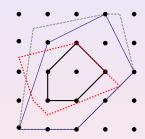
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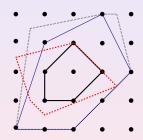
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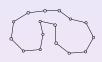
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Example - Traveling Salesman Problem

Classical Formulation

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



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2-Matching

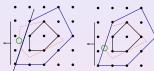
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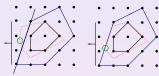
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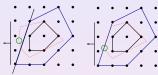


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The Dantzig-Wolfe Method (DW) iteratively builds an *inner* approximation of \mathcal{P}' by solving a column generation subproblem.



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



Common Threads

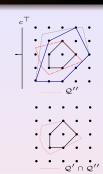
 The LP bound is obtained by optimizing over the intersection of two explicitly defined polyhedra.

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$$z_{CP} = z_{DW} = z_{LD} = z_D = \min_{x \in \mathbb{R}^n} \{c^\top x \mid x \in \mathcal{P}' \cap \mathcal{Q}''\} \ge z_{LP}$$

- 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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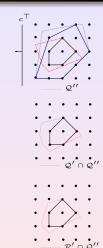
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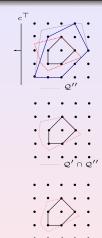
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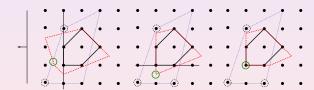


Price and Cut

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

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

- 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.

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

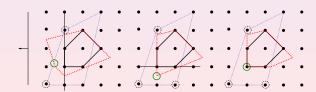
- In each iteration, separate $\hat{s} \in \operatorname{argmin}_{s \in \mathcal{F}'}\{(c^\top u^\top A'')s + u^\top b''\}$, a solution to the Lagrangian relaxation.
- Advantage: It is often much easier to separate a member of \mathcal{F}' from \mathcal{P} than an arbitrary real vector, such as \hat{x}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



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.



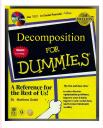
DECOMP Framework: Motivation

DECOMP Framework

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 wa
- The method for separation/optimization over P' is the primary application-dependen 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.

 The framework takes care of reformulation and implementation for all variants described here.

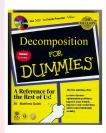


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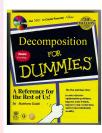


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DECOMP Framework: Implementation

COmputational INfrastructure for Operations Research



DECOMP was built around data structures and interfaces provided by COIN-OR.



Algorithms Interface: DecompAlgo

DECOMP provides the bounding method for branch and bound

• AlpsDecompModel : public AlpsModel

• AlpsDecompTreeNode : public AlpsTreeNode

• a wrapper class that calls (algorithmic) methods from DecompAlgo

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 - Applications Interface: DecompApp
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- 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).
- Column generation based on multiple algorithms can be easily defined and employed.
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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 original space).
- Column generation based on multiple algorithms can be easily defined and employed.
- Can derive bounds based on *multiple model/algorithm* combinations.
- Provides default (naive) branching rules in the original space.
- Active LP compression, variable and cut pool management.
- Flexible parameter interface: command line, param file, direct call overrides.
- Visualization tools for graph problems (linked to graphviz).

- The base class DecompApp provides an interface for the user to define the application-specific components of their algorithm.
- In order to develop an application, the user must derive the following methods/objects.

```
    DecompApp::APPcreateModel(). Define [A", b"] and [A', b'] (optional).
    TSP 1-Tree: [A", b"] define the 2-matching constraints.
    TSP 2-Match: [A", b"] define trivial subtour constraints.
    DecompApp::isUserFeasible(). Does x* define a feasible solution?
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    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.
 - DecompApp::APPheuristics()
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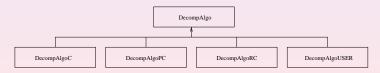
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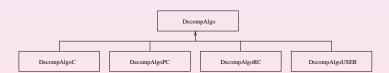
DECOMP - Algorithms

- The base class DecompAlgo provides the shell (init / master / subproblem / update).
- Each of the methods described have derived default implementations DecompAlgoX : public DecompAlgo which are accessible by any application class, allowing full flexibilit
- subroutines, which are called from the base class. For example,
 - 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.
 - Hybrid methods like using LD to initialize the columns of the DW master
 - During PC, adding cuts to both inner and outer approximations. simultaneously (Vanderbeck
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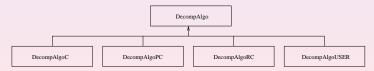
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DECOMP - Algorithms

- The base class DecompAlgo provides the shell (init / master / subproblem / update).
- Each of the methods described have derived default implementations DecompAlgoX : public DecompAlgo which are accessible by any application class, allowing full flexibility.
- New, hybrid or extended methods can be easily derived by overriding the various subroutines, which are called from the base class. For example,
 - Alternative methods for solving the master LP in DW, such as interior point methods or ACCPM.
 - Add stabilization to the dual updates in LD, as in bundle methods.
 - For LD, replace subgradient with Volume, providing an approximate primal solution.
 - Hybrid methods like using LD to initialize the columns of the DW master.
 - During PC, adding cuts to both inner and outer approximations. simultaneously (Vanderbeck).
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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():
```

- Traditional Decomposition Methods approximate \mathcal{P} as $\mathcal{P}' \cap \mathcal{Q}''$.
 - $\mathcal{P}' \supset \mathcal{P}$ may have a *large* description.
- Integrated Decomposition Methods approximate \mathcal{P} as $\mathcal{P}_I \cap \mathcal{P}_O$.
 - Both $\mathcal{P}_I \subset \mathcal{P}'$ and $\mathcal{P}_O \supset \mathcal{P}$ may have a *large* description.
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 - The user only needs to define the components based on the compact formulation (irrespective o 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.
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