

Constraint Integer Programming

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Constraint Integer Programming

SCIP's Design

The Solving Process of SCIP

<http://scipopt.org>

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What is a Constraint Integer Program?

Constraint Integer Program

Objective function:

- ▷ linear function

Feasible set:

- ▷ described by arbitrary constraints

Variable domains:

- ▷ real or integer values

Restriction:

- ▷ When all integer variables are fixed, remaining subproblem is LP or NLP

$$\begin{aligned} \min \quad & c^T x \\ \text{s.t.} \quad & x \in F \\ & (x_I, x_C) \in \mathbb{Z}^I \times \mathbb{R}^C \end{aligned}$$

Remark:

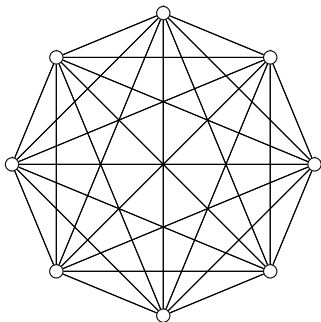
- ▶ arbitrary objective or variables modeled by constraints

An Example: the Traveling Salesman Problem

Definition (TSP)

Given a complete graph $G = (V, E)$
and distances d_e for all $e \in E$:

Find a **Hamiltonian cycle** (cycle
containing all nodes, tour) of minimum
length.



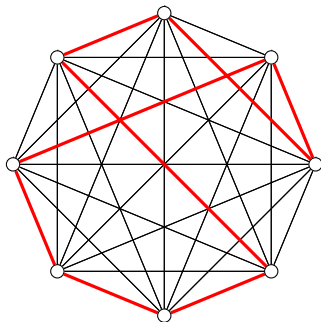
K_8

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K_8

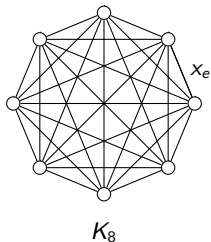
TSP – Integer Programming Formulation

Given

- ▶ complete graph $G = (V, E)$
- ▶ distances $d_e > 0$ for all $e \in E$

Binary variables

- ▶ $x_e = 1$ if edge e is used



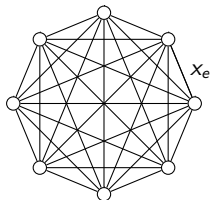
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K_8

$$\begin{array}{ll}
 \min & \sum_{e \in E} d_e x_e \\
 \text{subject to} & \sum_{e \in \delta(v)} x_e = 2 \quad \forall v \in V \\
 & \sum_{e \in \delta(S)} x_e \geq 2 \quad \forall S \subset V, S \neq \emptyset \\
 & x_e \in \{0, 1\} \quad \forall e \in E
 \end{array}$$

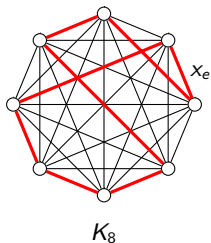
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$$\min \sum_{e \in E} d_e x_e$$

subject to	$\sum_{e \in \delta(v)} x_e = 2$	$\forall v \in V$	node degree
------------	----------------------------------	-------------------	-------------

$\sum_{e \in \delta(S)} x_e \geq 2$	$\forall S \subset V, S \neq \emptyset$
-------------------------------------	---

$x_e \in \{0, 1\}$	$\forall e \in E$
--------------------	-------------------

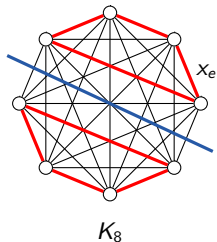
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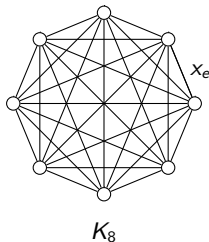
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Binary variables

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$$\min \sum_{e \in E} d_e x_e \quad \text{distance}$$

$$\text{subject to } \sum_{e \in \delta(v)} x_e = 2 \quad \forall v \in V$$

$$\sum_{e \in \delta(S)} x_e \geq 2 \quad \forall S \subset V, S \neq \emptyset$$

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 & \text{nosubtour}(x) \\
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 \end{aligned}$$

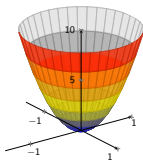
(CIP formulation of TSP)

Single nosubtour constraint rules out subtours (e.g. by domain propagation). It may also separate subtour elimination inequalities.

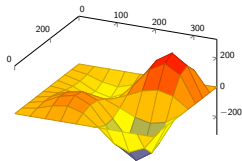
Mixed-Integer Nonlinear Programs (MINLPs)

$$\begin{aligned}
 & \min c^T x \\
 & \text{s.t. } g_k(x) \leq 0 && \forall k \in [m] \\
 & \quad x_i \in \mathbb{Z} && \forall i \in \mathcal{I} \subseteq [n] \\
 & \quad x_i \in [\ell_i, u_i] && \forall i \in [n]
 \end{aligned}$$

The functions $g_k \in C^1([\ell, u], \mathbb{R})$ can be



convex or

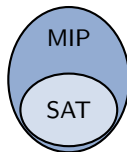


nonconvex

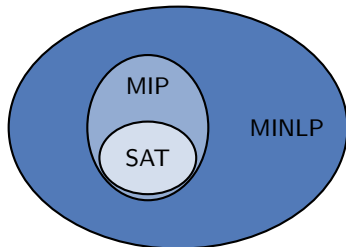
► Mixed Integer Programs



- ▶ Mixed Integer Programs
- ▶ SATisifiability problems

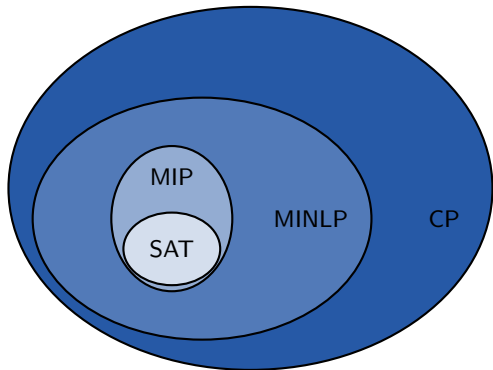


- ▶ Mixed Integer Programs
- ▶ SATisifiability problems
- ▶ Mixed Integer Nonlinear Programs

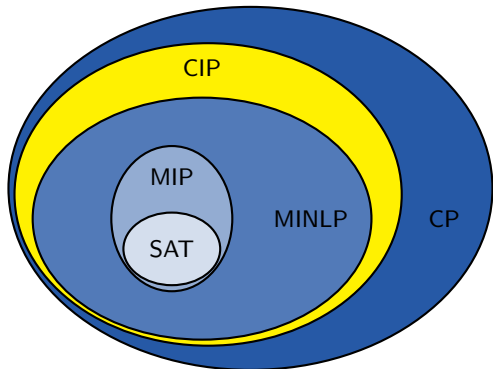


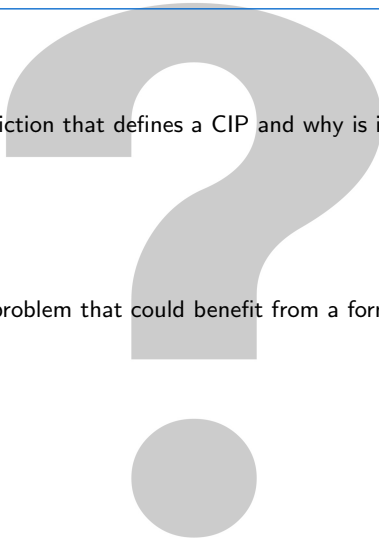
Constraint Integer Programming

- ▶ Mixed Integer Programs
- ▶ SAT is feasibility problems
- ▶ Mixed Integer Nonlinear Programs
- ▶ Constraint Programming



- ▶ Mixed Integer Programs
- ▶ SATisifiability problems
- ▶ Mixed Integer Nonlinear Programs
- ▶ Constraint Programming
- ▶ Constraint Integer Programming





- ▶ What is the key restriction that defines a CIP and why is it necessary?

- ▶ Think of a different problem that could benefit from a formulation as a CIP

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Possibilities: Linear ordering problem, Steiner tree problem, scheduling, ...

Constraint Integer Programming

SCIP's Design

The Solving Process of SCIP

<http://scipopt.org>

Different plugin classes are responsible of the following tasks.

1. Presolving and node propagation
 - ▶ **Constraint handlers**
 - ▶ Presolvers
 - ▶ Propagators
2. Separation
 - ▶ **Constraint handlers**
 - ▶ Separators
3. Improving solutions
 - ▶ Primal heuristics
4. Branching
 - ▶ **Constraint handlers**
 - ▶ Branching rules
5. Node selection
 - ▶ Node selectors

Plugin based design

SCIP core

- ▶ branching tree
- ▶ variables
- ▶ conflict analysis
- ▶ solution pool
- ▶ cut pool
- ▶ statistics
- ▶ clique table
- ▶ implication graph
- ▶ ...

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Plugins

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- ▶ interact with the framework through a **very detailed interface**

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Plugins

- ▶ **external callback objects**
 - ▶ interact with the framework through a **very detailed interface**
 - ▶ SCIP knows for each plugin type:
 - ▶ the number of available plugins
 - ▶ priority defining the calling order (usually)
 - ▶ SCIP does not know any structure behind a plugin
- ⇒ **plugins are black boxes** for the SCIP core

Plugin based design

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Plugins

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- ⇒ **plugins are black boxes** for the SCIP core
- ⇒ **Very flexible branch-and-bound based search algorithm**

Constraint Handlers

Constraint handlers

- ▶ most powerful plugins in SCIP
- ▶ define the feasible region
- ▶ a single constraint may represent a whole set of inequalities

Functions

- ▶ check and enforce feasibility of solutions
- ▶ can add linear representation to LP relaxation
- ▶ constraint-specific presolving, domain propagation, separation

Result

- ▶ SCIP is **constraint based**
 - ▶ Advantage: flexibility
 - ▶ Disadvantage: limited global view: **A constraint knows its variables but a variable does not know the constraints it appears in.**
 - ▶ LP relaxation is an add-in and handled by external software via LP interface.

Types of Plugins

- ▶ **Constraint handler:** assures feasibility, strengthens formulation
- ▶ **Separator:** adds cuts, improves dual bound
- ▶ **Pricer:** allows dynamic generation of variables
- ▶ **Heuristic:** searches solutions, improves primal bound
- ▶ **Branching rule:** how to divide the problem?
- ▶ **Node selection:** which subproblem should be regarded next?
- ▶ **Presolver:** simplifies the problem in advance, strengthens structure
- ▶ **Propagator:** simplifies problem, improves dual bound locally
- ▶ **Reader:** reads problems from different formats
- ▶ **Event handler:** catches events (e.g., bound changes, new solutions)
- ▶ **Display:** allows modification of output
- ▶ **Relaxation handlers:** custom relaxations
- ▶ ...

What does SCIP know about plugins?

- ▶ interactive shell shows the information SCIP has for a plugin type

```
SCIP> display {branching | conshdlrs | heuristics | ...}
```

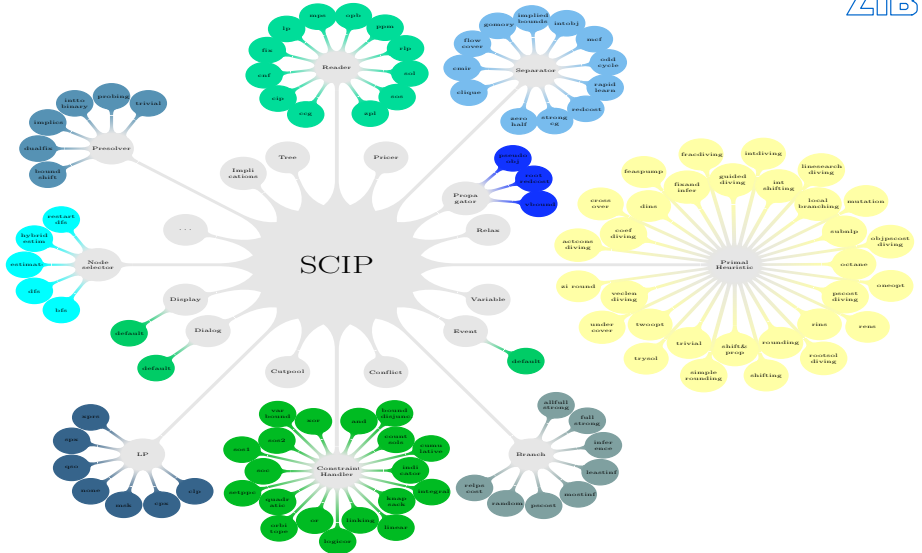
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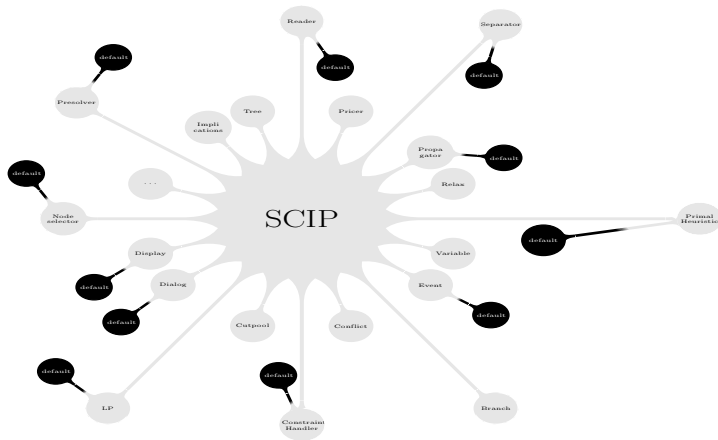
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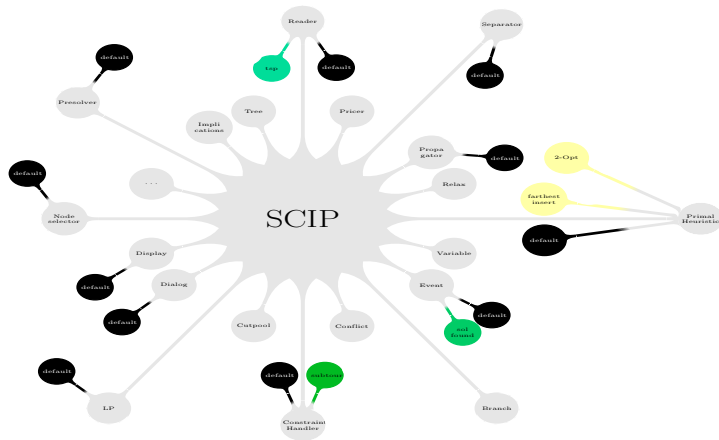
```
SCIP> display {branching | conshdlrs | heuristics | ...}
```

```
SCIP> display branching
```

branching	rule	priority	maxdepth	maxbddist	description
relpscost		10000	-1	100.0%	reliability branching on pseudo cost values
pscost		2000	-1	100.0%	branching on pseudo cost values
inference		1000	-1	100.0%	inference history branching
mostinf		100	-1	100.0%	most infeasible branching
leastinf		50	-1	100.0%	least infeasible branching
distribution		0	-1	100.0%	branching rule based on variable influence on cumu
fullstrong		0	-1	100.0%	full strong branching
cloud		0	-1	100.0%	branching rule that considers several alternative
allfullstrong		-1000	-1	100.0%	all variables full strong branching
random		-100000	-1	100.0%	random variable branching







Summary

Plugins for the SCIP core

- ▶ are black boxes for SCIP
- ▶ perform all problem specific actions
- ▶ interact through a **very detailed interface**

Plugins

- ▶ can have private data
- ▶ can access global information of the SCIP core
- ▶ should not access data of other plugins

Result

- ▶ SCIP is **constraint based**
- ▶ new types of constraints can easily be defined
- ▶ easy to add problem specific algorithms

Advantages and Disadvantages of SCIP

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- ▶ **Extendable in any direction since constraint based**
 - ▶ Adding global constraints
 - ▶ Adding problem specific plugins

Advantages and Disadvantages of SCIP

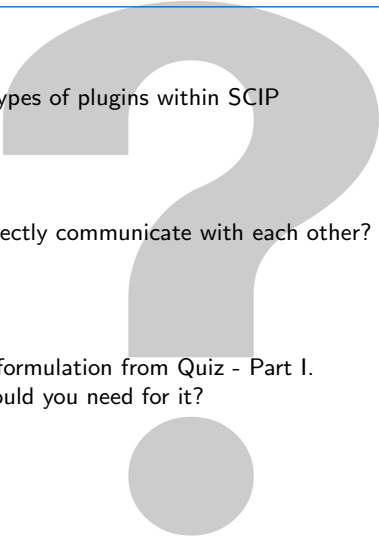
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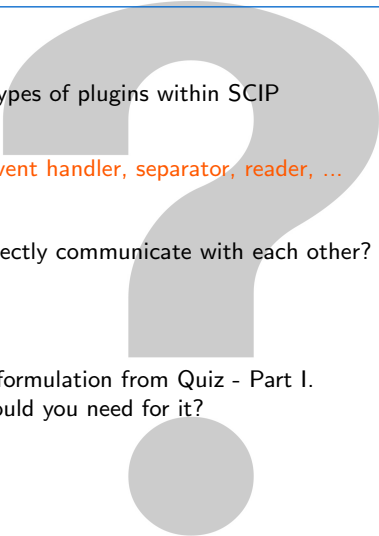
Disadvantages

- ▶ **Hundreds of parameters** to play with
- ▶ In general variables do not know in which constraint they appear since constraint based

Quiz - Part II

- 
- ▶ Think of 3 different types of plugins within SCIP
 - ▶ Why can't plugins directly communicate with each other?
 - ▶ Recall your own CIP formulation from Quiz - Part I. What plugin types would you need for it?

Quiz - Part II

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- ▶ Think of 3 different types of plugins within SCIP
 - Constraint handler, event handler, separator, reader, ...
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What plugin types would you need for it?

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Constraint handler, event handler, separator, reader, ...

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Keep the framework as flexible as possible, no interdependencies

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What plugin types would you need for it?

Very likely: Constraint handler, reader, heuristic

Constraint Integer Programming

SCIP's Design

The Solving Process of SCIP

<http://scipopt.org>

Basic Workflow

```
read ../check/instances/MIP/bell5.mps
optimize
write solution mysolution.sol
quit
```

Displaying information

Use the `display ...` command to enter the menu and

- ▶ obtain `solution` information
- ▶ print the current `transproblem` to the console
- ▶ display plugin information, e.g., list all available branching rules

Changing Settings

Use the `set ...` command to list the settings menu.

Important Parameters

Numerical parameters

These must be set **before** reading a problem.

- ▶ `numerics/feastol`, default 10^{-6}
- ▶ `numerics/epsilon`, default 10^{-9}
- ▶ `numerics/infinity`, default 10^{20}

Limits

- ▶ `limits/time`
- ▶ `limits/nodes`
- ▶ `limits/gap`
- ▶ ...

Randomization

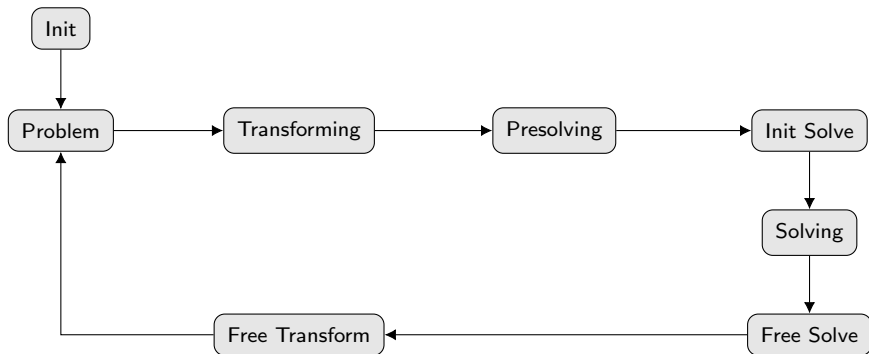
- ▶ `randomization/randomseedshift`
- ▶ `randomization/lpseed`
- ▶ `randomization/permutationseed`

- ▶ interactive shell supports 11 different input formats
→ cip, cnf, flatzinc, rlp, lp, mps, opb, pip, wbo, **zimpl**, smps
- ▶ C API/callable library
- ▶ C++ wrapper classes
- ▶ Python interface
 - ▶ New Dockerized SCIP
 - ▶ conda integration coming soon
- ▶ Java JNI interface
- ▶ Julia interface
- ▶ AMPL
- ▶ GAMS
- ▶ Matlab (see also OPTI toolbox, <http://www.i2c2.aut.ac.nz/Wiki/OPTI/>)

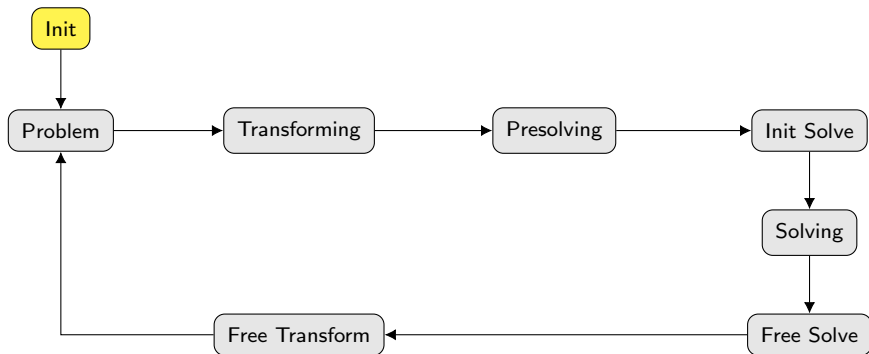
If you should ever get stuck, you can ...

1. type `help` in the interactive shell
2. read the documentation <http://scipopt.org/doc/html>
→ [FAQ](#), [HowTos for each plugin type](#), debugging, automatic testing, ...
3. search or post on [Stack Overflow](#) using the tag `scip` (more than 100 questions already answered)
4. active mailing list scip@zib.de (350+ members)
 - ▶ search the mailing list archive (append site:listserv/pipermail/scip)
 - ▶ register <http://listserv.zib.de/mailman/listinfo/scip/> and post

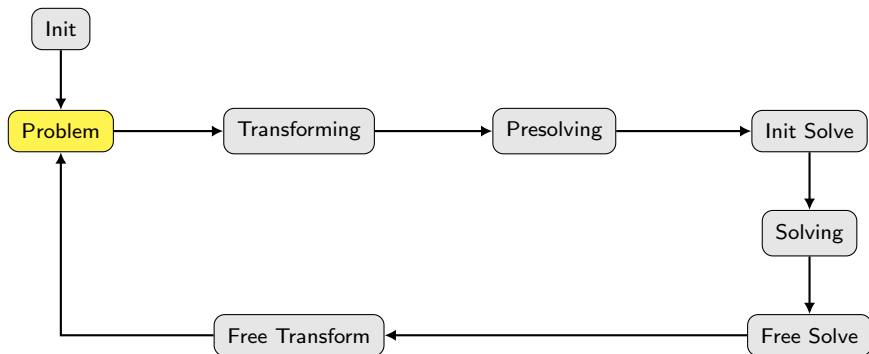
Operational Stages



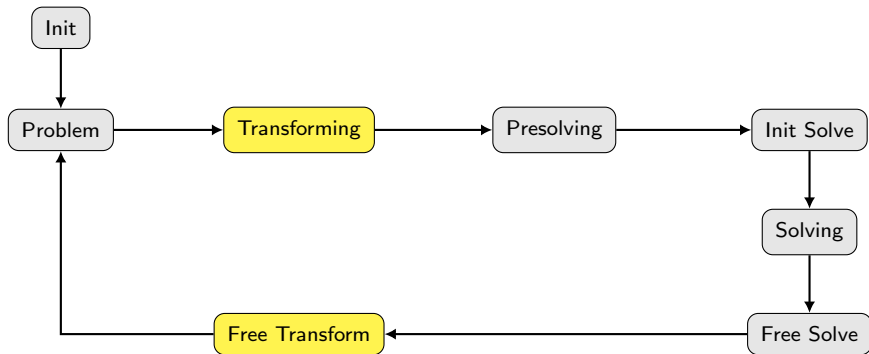
Operational Stages



- ▶ Basic data structures are allocated and initialized.
- ▶ User includes required plugins (or just takes default plugins).

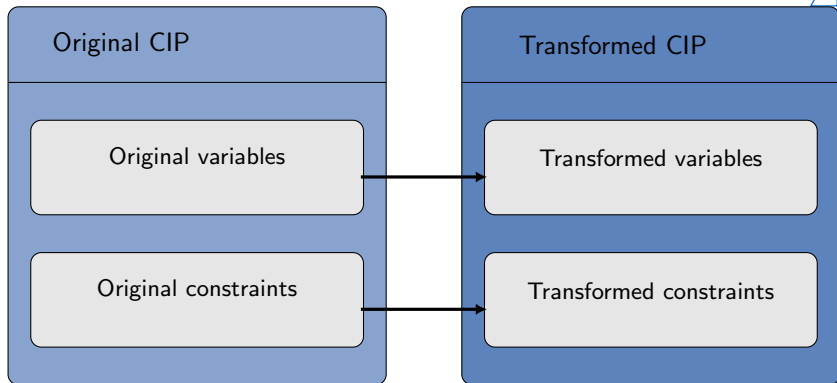


- ▶ User creates and modifies the original problem instance.
- ▶ Problem creation is usually done in file readers.

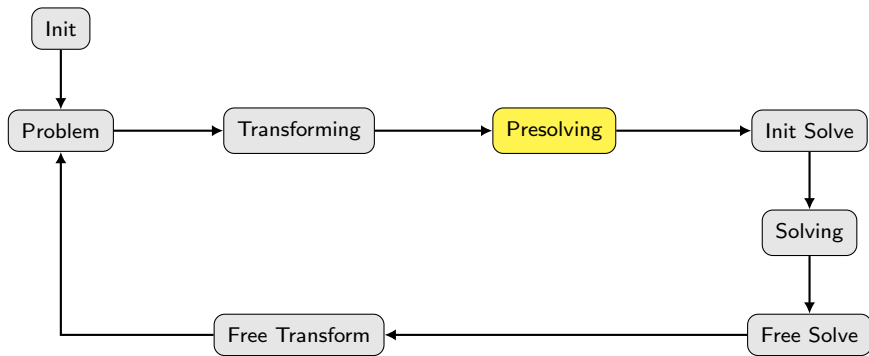


- Creates a working copy of the original problem.

Original and Transformed Problem



- ▶ data is copied into separate memory area
- ▶ presolving and solving operate on transformed problem
- ▶ original data can only be modified in problem modification stage



Presolving Tips and Parameters

Use `display presolvers` to **list** all presolvers of SCIP.

Disable Presolving

Disable **all** presolving for a model

```
set presolving emphasis off
```

Deactivate **single** techniques

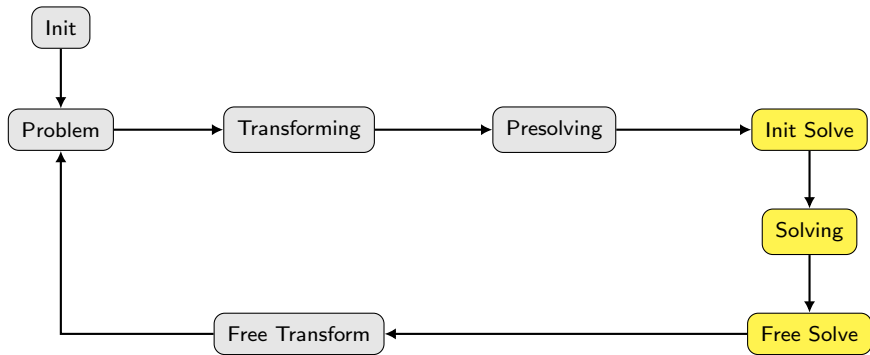
```
set presolving tworowbnd maxrounds 0
set propagating probing maxprerounds 0
set constraints components advanced maxprerounds 0
```

Aggressive Presolving

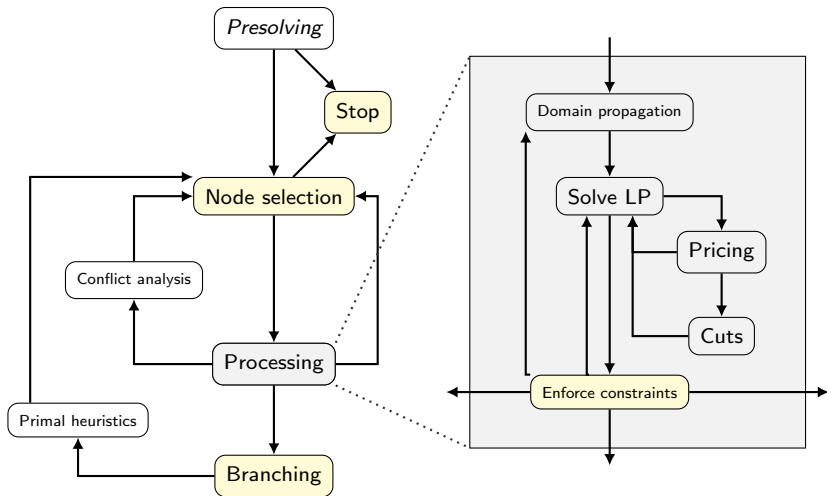
```
set presolving emphasis aggressive
```

General Rule of Thumb

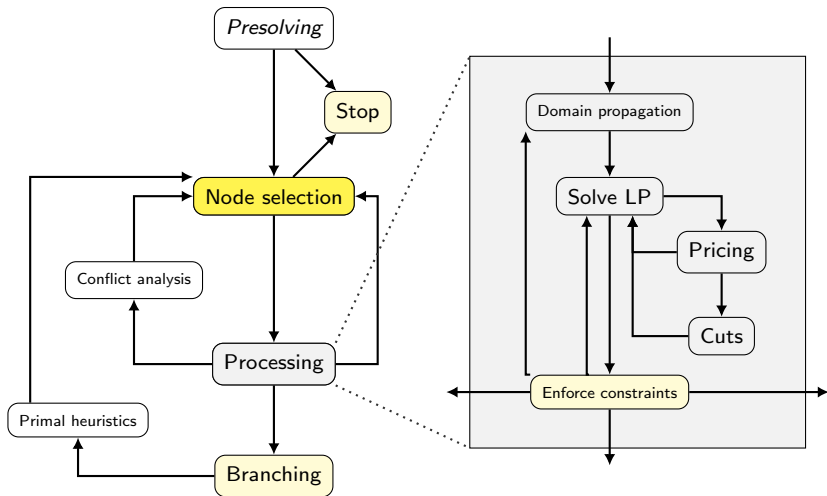
Only deactivate single presolving techniques if you encounter performance problems.



Flow Chart SCIP



Flow Chart SCIP



Node Selection Tips and Parameters

Available Node Selectors

```
display nodeselectors
```

<u>node selector</u>	<u>std priority</u>	<u>memsave</u>	<u>prio</u>	<u>description</u>
estimate	200000		100	best estimate search
bfs	100000		0	best first search
...				
dfs	0		100000	depth first search

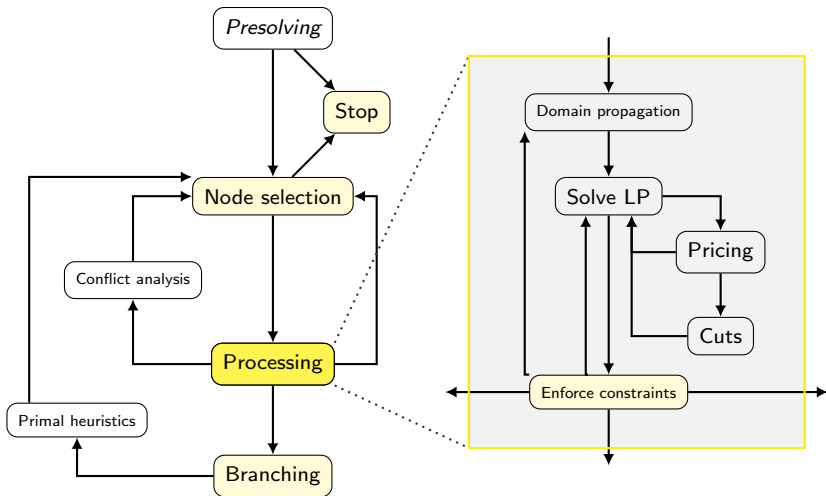
Switching Node Selectors

Only the node selector with **highest** standard priority is active. Use

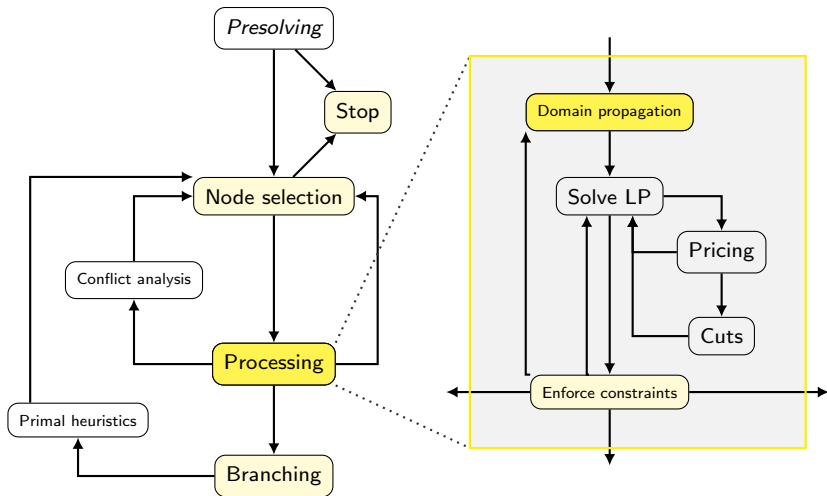
```
set nodeselection dfs stdpriority 1000000
```

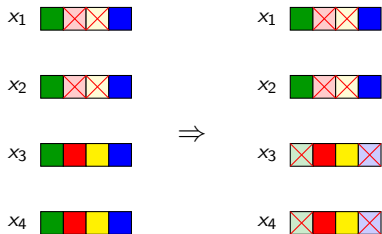
to activate depth first search also in non-memsave mode.

Flow Chart SCIP



Flow Chart SCIP





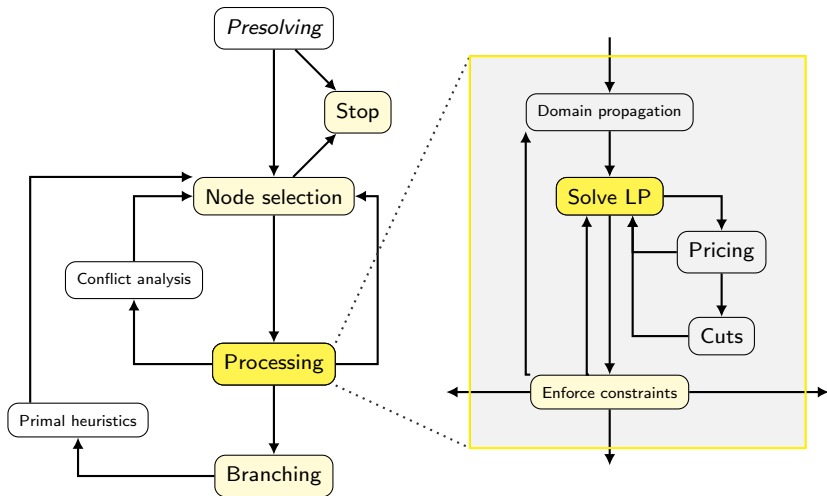
Task

- ▶ simplify model locally
- ▶ improve local dual bound
- ▶ detect infeasibility

Techniques

- ▶ **constraint specific**
 - ▶ each cons handler may provide a propagation routine
 - ▶ reduced presolving (usually)
- ▶ **dual propagation**
 - ▶ root reduced cost strengthening
 - ▶ objective function
- ▶ **special structures**
 - ▶ variable bounds

Flow Chart SCIP

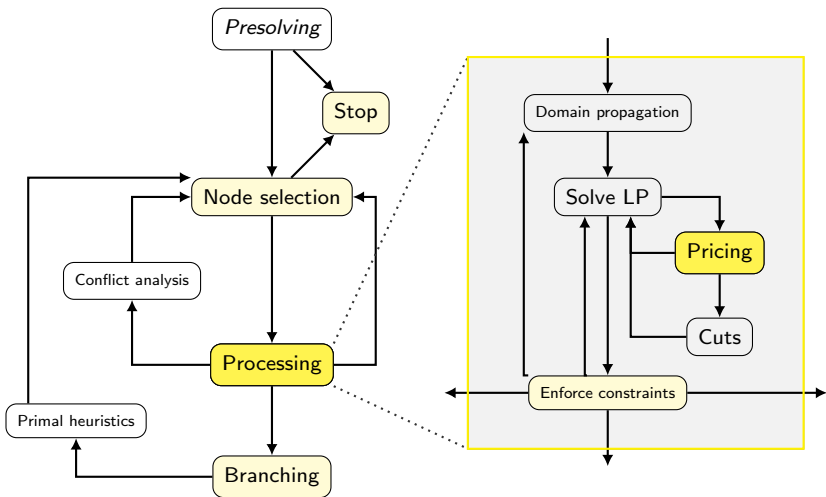


Most Important LP Parameters

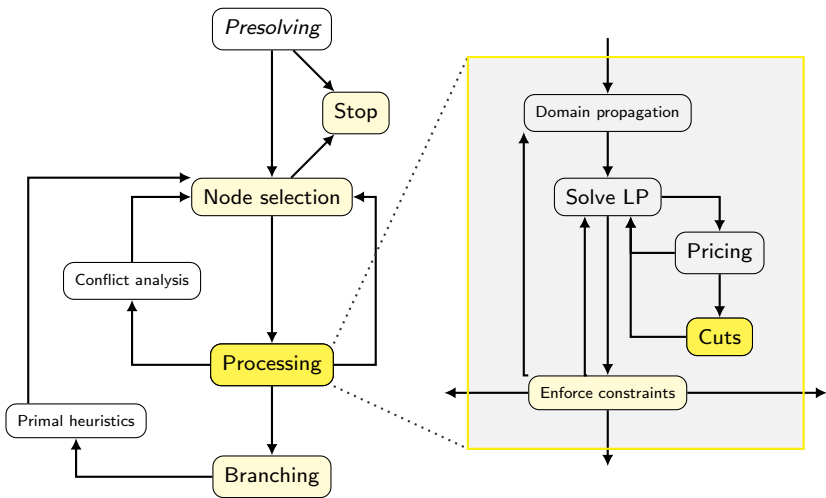
- ▶ `lp/initialalgorithm`, `lp/resolvealgorithm`
 - ▶ Primal/Dual Simplex Algorithm
 - ▶ Barrier w and w/o crossover
- ▶ `lp/pricing`
 - ▶ normally LP solver specific default
 - ▶ Devex
 - ▶ Steepest edge
 - ▶ Quick start steepest edge
- ▶ `lp/threads`

Slow LP performance is a blocker for the solving process and can sometimes be manually tuned significantly.

Flow Chart SCIP



Flow Chart SCIP



Separation Tips and Parameters

Disable/Speed up/Emphasize All Separation

```
set separating emphasis off/fast/aggressive
```

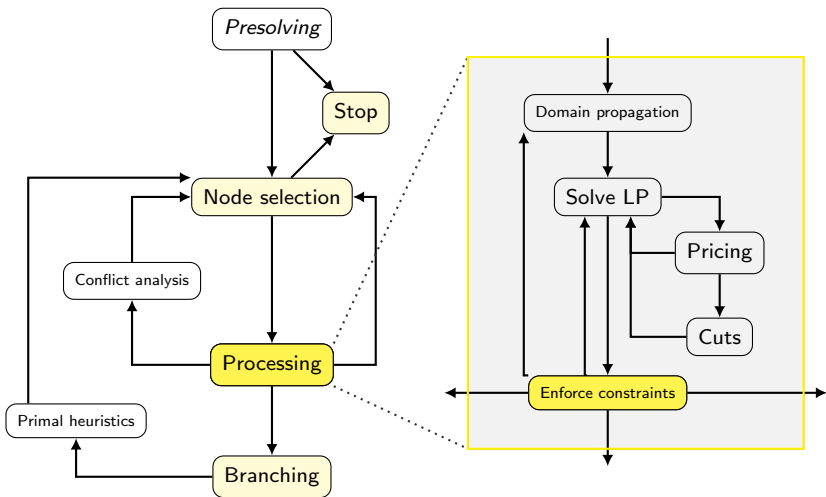
Disable Single Separation Techniques

```
set separating clique freq -1  
set constraints cardinality sepafreq -1
```

Some Important Parameters

- ▶ `separating/maxcuts`, `separating/maxcutsroot`
- ▶ `separating/maxrounds`, `separating/maxroundsroot`
- ▶ `separating/maxstallrounds`, `separating/maxstallroundsroot`

Flow Chart SCIP



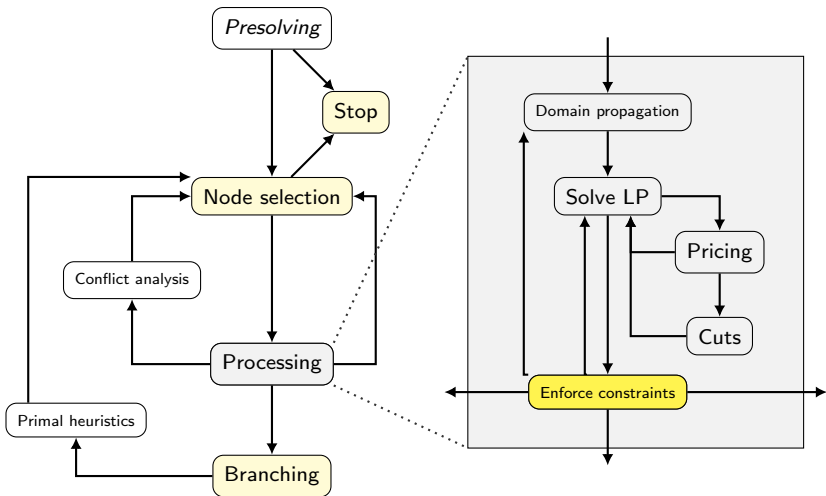
LP solution may violate a constraint not contained in the relaxation.

Enforcing is necessary for a correct implementation!

Constraint handler resolves the infeasibility by ...

- ▶ Reducing a variable's domain,
- ▶ Separating a cutting plane (may use integrality),
- ▶ Adding a (local) constraint,
- ▶ Creating a branching,
- ▶ Concluding that the subproblem is infeasible and can be cut off, or
- ▶ Just saying "solution infeasible".

Constraint Enforcement



▶ Reduced domain

▶ Added constraint

▶ Added cut

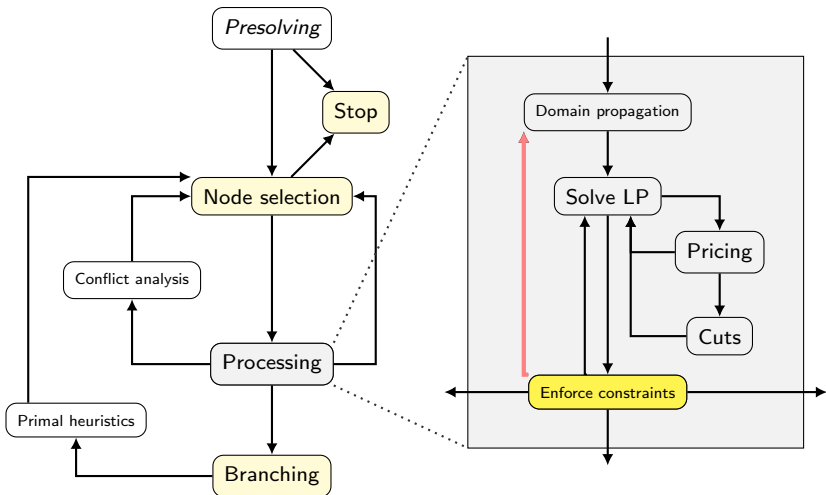
▶ Branched

▶ Cutoff

▶ Infeasible

▶ Feasible

Constraint Enforcement



▶ **Reduced domain**

▶ **Added constraint**

▶ Added cut

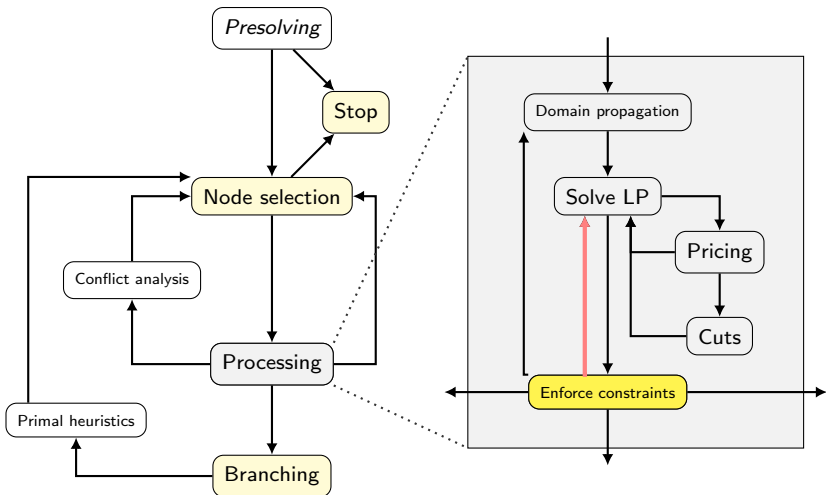
▶ Branched

▶ Cutoff

▶ Infeasible

▶ Feasible

Constraint Enforcement



▶ Reduced domain

▶ **Added cut**

▶ Cutoff

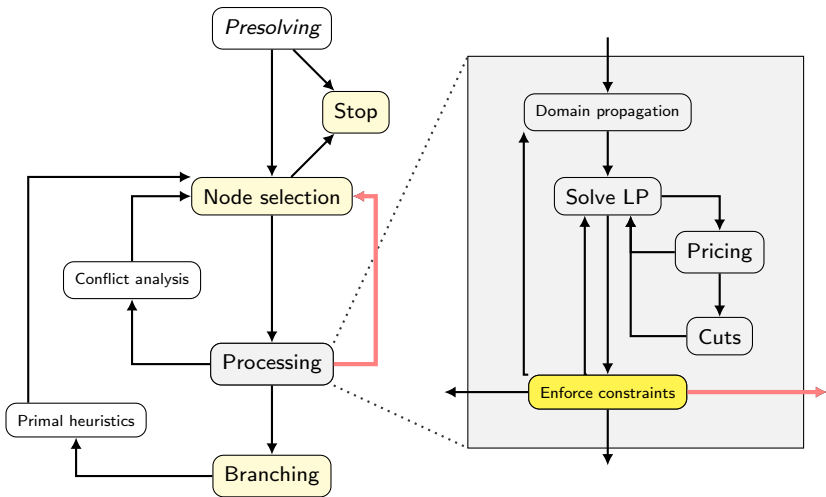
▶ Infeasible

▶ Added constraint

▶ Branched

▶ Feasible

Constraint Enforcement



▶ Reduced domain

▶ Added constraint

▶ Added cut

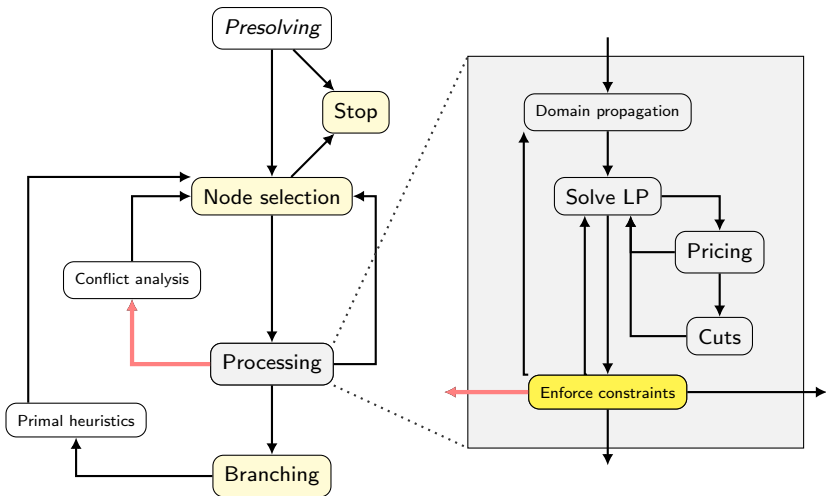
▶ **Branched**

▶ Cutoff

▶ Infeasible

▶ Feasible

Constraint Enforcement



▶ Reduced domain

▶ Added cut

▶ **Cutoff**

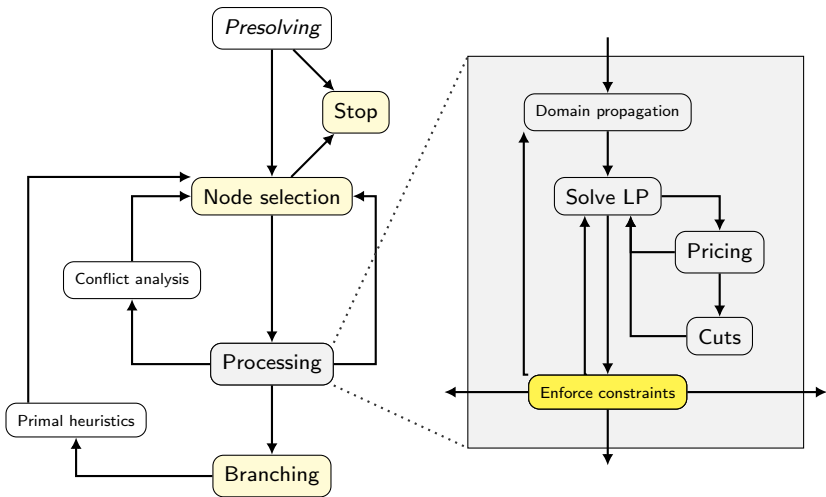
▶ Infeasible

▶ Added constraint

▶ Branched

▶ Feasible

Constraint Enforcement



▶ Reduced domain

▶ Added cut

▶ Cutoff

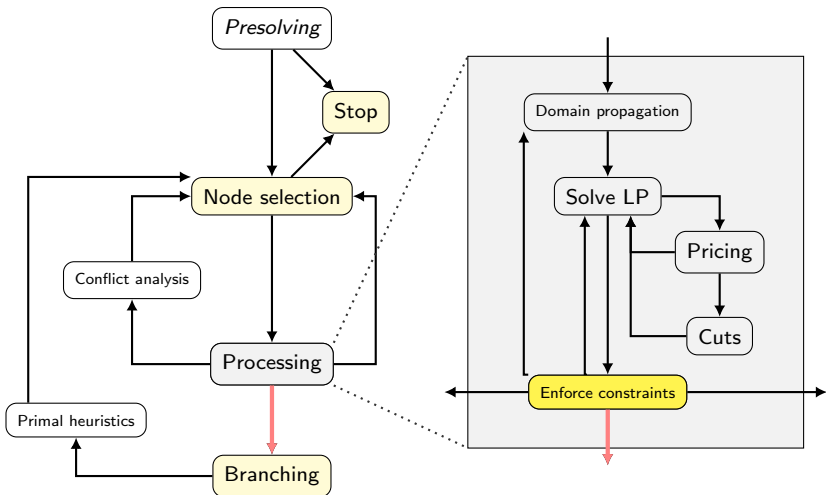
▶ Infeasible

▶ Added constraint

▶ Branched

▶ Feasible

Constraint Enforcement



▶ Reduced domain

▶ Added cut

▶ Cutoff

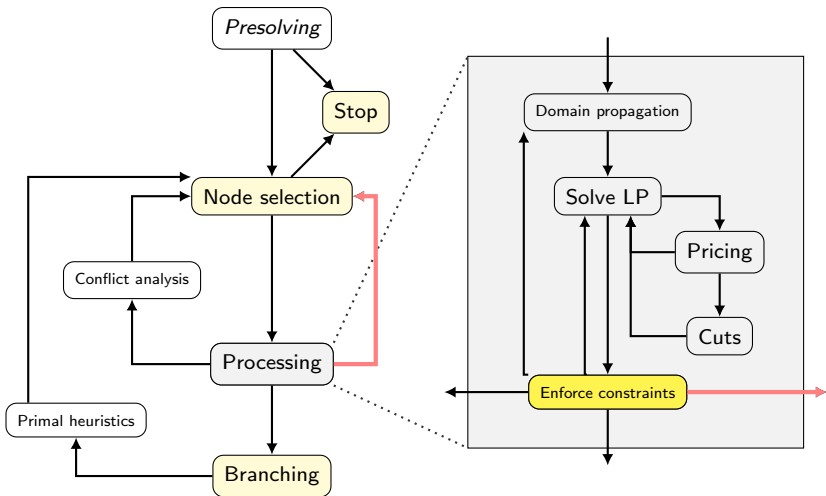
▶ **Infeasible**

▶ Added constraint

▶ Branched

▶ Feasible

Constraint Enforcement



▶ Reduced domain

▶ Added constraint

▶ Added cut

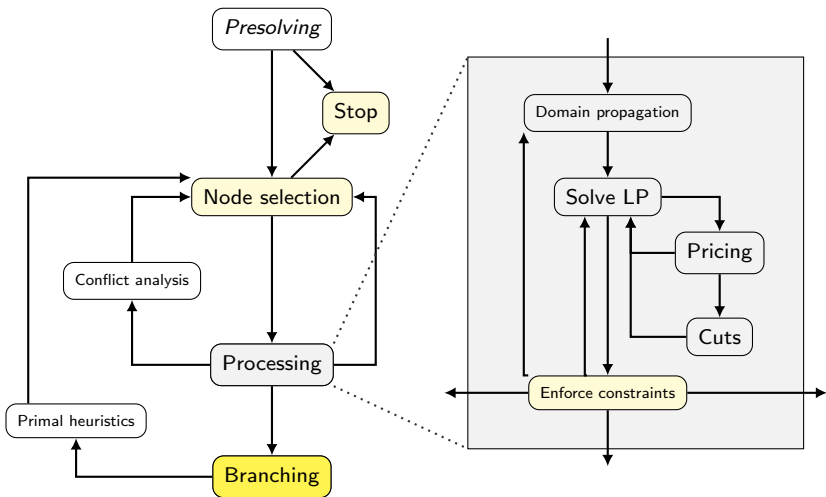
▶ Branched

▶ Cutoff

▶ Infeasible

▶ **Feasible**

Flow Chart SCIP



Branching Rule Tips and Parameters

Branching Rule Selection

Branching rules are applied in decreasing order of priority.

```
SCIP> display branching
```

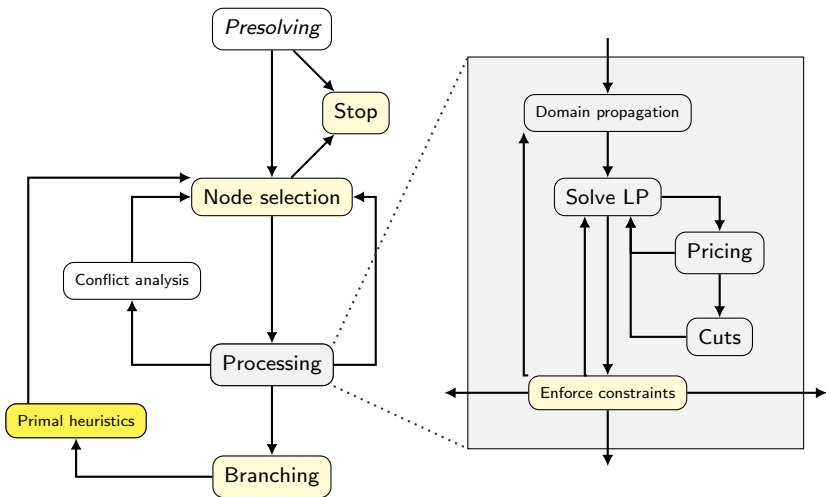
<u>branching rule</u>	<u>priority</u>	<u>maxdepth</u>	<u>maxbddist</u>
relpscost	10000	-1	100.0%
pscost	2000	-1	100.0%
inference	1000	-1	100.0%
mostinf	100	-1	100.0%

Reliability Branching Parameters

All parameters prefixed with `branching/relpscost/`

- ▶ `sbiterquot`, `sbiterofs` to increase the budget for strong branching
- ▶ `minreliable` (= 1), `maxreliable` (= 5) to increase threshold to consider pseudo costs as reliable

Flow Chart SCIP



Primal Heuristics Tips and Parameters

Disable/Speed Up/Emphasize Heuristics

```
set heuristics emphasis off/fast/aggressive
```

Disable an individual heuristic via

```
set heuristics feaspump freq -1
```

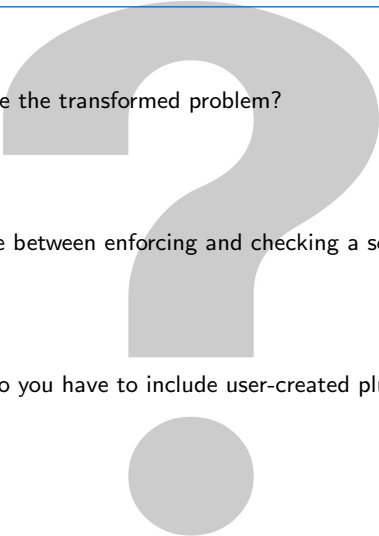
Important Parameters

- ▶ `heuristics/alns/nodesofs`, `heuristics/alns/nodesquot` to increase the computational budget of this LNS technique
- ▶ `heuristics/guideddiving/... lpsolvefreq`, `maxlpiterofs` `maxlpiterquot` to control the LP solving during this diving technique

Advice

Use emphasis settings. **Do not** attempt to individually tune heuristics by hand.

Quiz - Part III

- 
- ▶ Why does SCIP create the transformed problem?
 - ▶ What is the difference between enforcing and checking a solution?
 - ▶ During which stage do you have to include user-created plugins?

Quiz - Part III

- ▶ Why does SCIP create the transformed problem?

Preserve original problem, ensure feasibility

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Enforcing resolves infeasibility

- ▶ During which stage do you have to include user-created plugins?

- ▶ Why does SCIP create the transformed problem?

Preserve original problem, ensure feasibility

- ▶ What is the difference between enforcing and checking a solution?

Enforcing resolves infeasibility

- ▶ During which stage do you have to include user-created plugins?

The init stage

Important SCIP topics not covered in this talk:

- ▶ more Details in [Documentation](#)
- ▶ branch-and-price: SCIP can be extended to a problem-specific branch-cut-and-price solver
 - ▶ Q&A on youtube channel
 - ▶ see [The Bin Packing Example in C](#)
 - ▶ see also [GCG](#)
- ▶ allows for Benders decomposition since version 6.0
 - ▶ Q&A on youtube channel
- ▶ browse [technical reports](#) for details on recently added cutting plane selection, primal heuristics, symmetry breaking, and much more

Thank you for your attention

