

BARON:
Branch and Reduce Optimization
Navigator

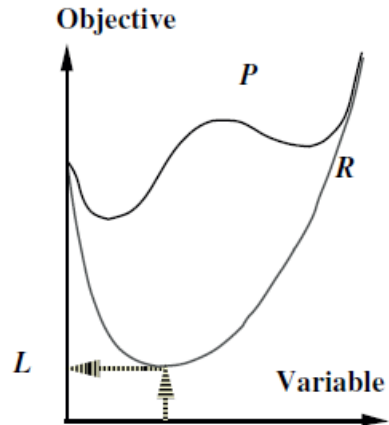
A High Level Overview for CME334

Assembled by Thomas Lipp

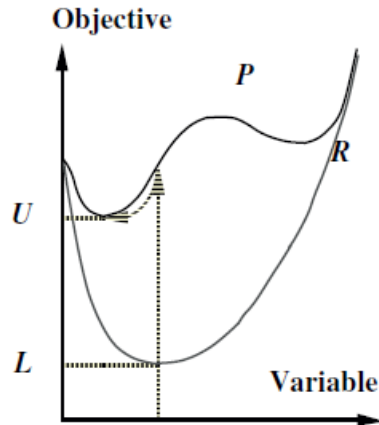
What is BARON?

- BARON is a branch and bound non-linear, mixed integer global optimization solver dating back to 1991, although last updated in 2010
- Developed at University of Illinois at Urbana-Champaign by Nikolaos Sahinidis.
- Problems can be stated in either the AIMMS or GAMS modeling languages, although computing time is available on the NEOS server for GAMS
- Objective and constraint functions must be factorable
- All non-linear variables and expressions must be bounded from above and below

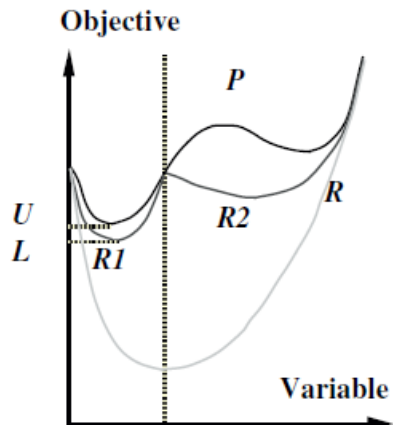
Branch and Bound Algorithms



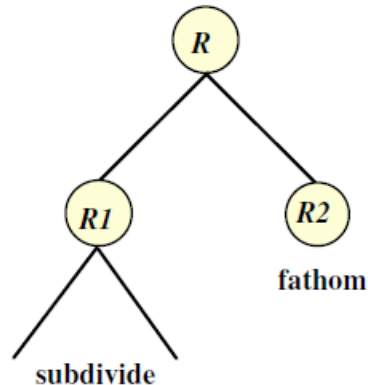
a. Lower bounding



b. Upper bounding



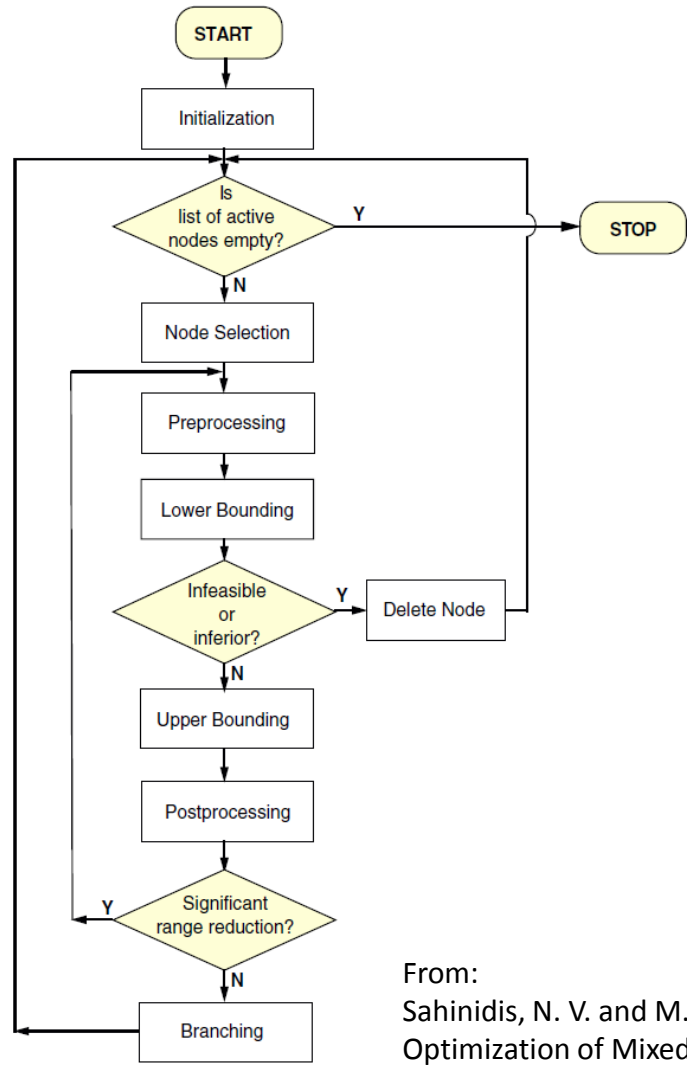
c. Domain subdivision



d. Search tree

A quick reminder of Branch and Bound algorithms.

What does the Baron Algorithm Look like?



From:
Sahinidis, N. V. and M. Tawarmalani, BARON 9.0.4: Global Optimization of Mixed-Integer Nonlinear Programs, User's manual, 2010

What does each step do?

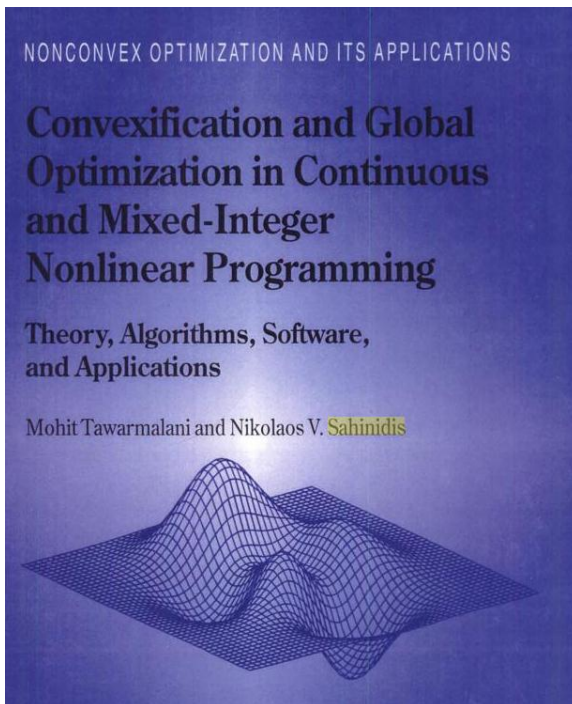
- Unfortunately, the Baron documentation does not address each step fully. BARON is written to be modular so that user subroutines can be used to replace certain steps as needed.
- The creators are also trying to market BARON:
<http://www.theoptimizationfirm.com/>



- The subproblems are generally fed to existing solvers: SNOPT, MINOS, CPLEX, etc.

Things I won't be able to answer

- How does BARON handle mixed integers
- What exactly occurs in any given step
- There may be more information available in:

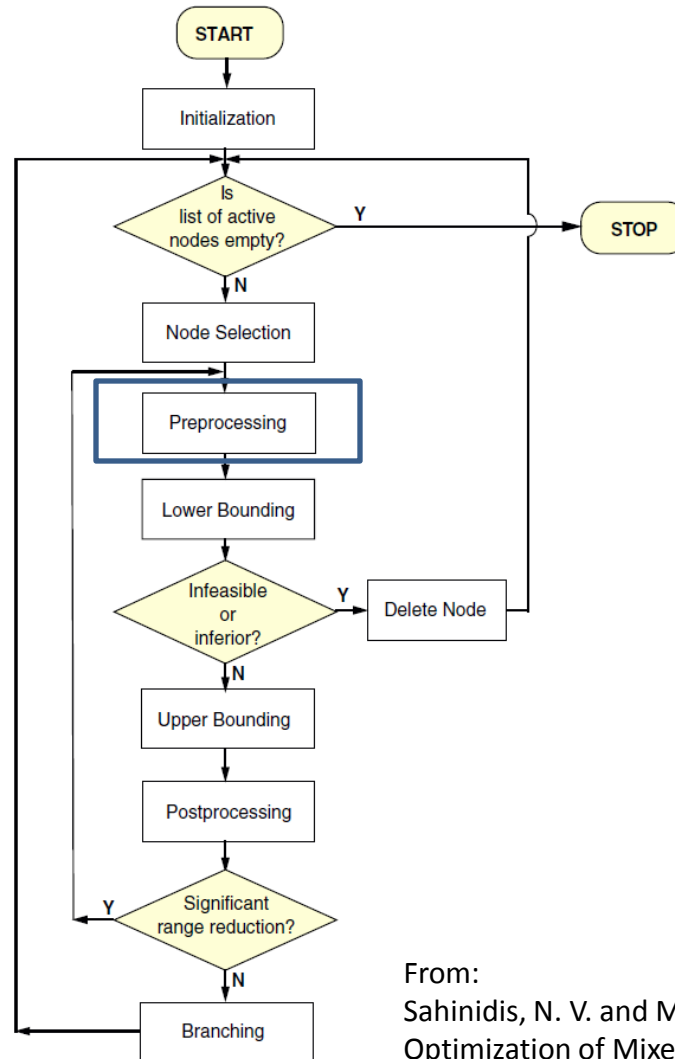


Available for \$221.09, and not currently owned by the Stanford Library.

What does BARON have to offer, then?

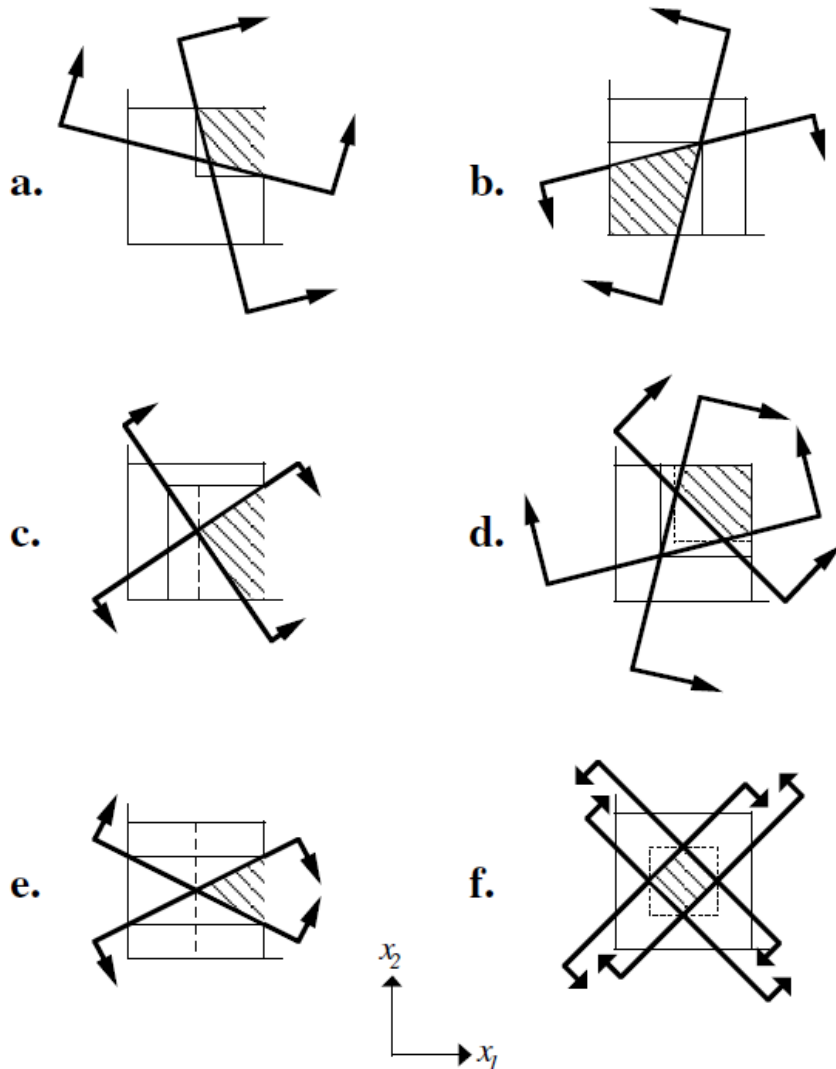
- BARON calls itself a branch and reduce algorithm rather than a branch and bound algorithm, and it is this “reducing” that will provide the most insight.
- The modular nature of BARON allows it to be tailored to your specific application.

Preprocessing on a Node



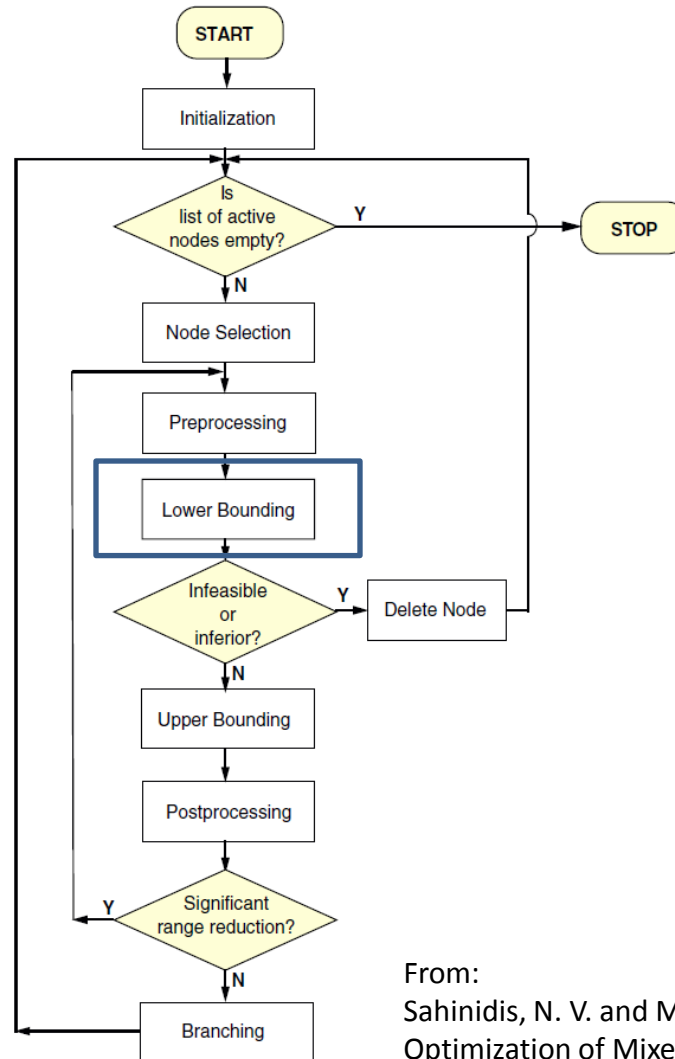
From:
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Feasibility Based Range Reduction: a “Poor Man’s LP”



- The solid outer box represents the initial domain of the node.
- The solid inner box represents the reduced domain by considering each of the constraints individually. “The Poor Man’s LP”
- The dotted boundary represents solving for the limits on each coordinate with all constraints active
- The shaded region is the true feasible region.

Lower Bounding



From:
Sahinidis, N. V. and M. Tawarmalani, BARON 9.0.4: Global Optimization of Mixed-Integer Nonlinear Programs, User's manual, 2010

Lower Bounding

This is an overview of the lower bounding scheme used by BARON, but is a little complex to go over. Remember that one requirement of all of our functions was that they be factorable:

Algorithm Relax $f(x)$

If $f(x)$ is a function of single variable $x \in [x^l, x^u]$, then

Construct under- and over-estimators for $f(x)$ over $[x^l, x^u]$,

else if $f(x) = g(x)/h(x)$, then

Fractional_Relax (f, g, h)

end of if

else if $f(x) = \prod_{i=1}^t f_i(x)$, then

for $i := 1$ to t do

Introduce variable y_{f_i} , such that $y_{f_i} = \text{Relax } f_i(x)$

end of for

Introduce variable y_f , such that $y_f = \text{Multilinear_Relax } \prod_{i=1}^t y_{f_i}$

else if $f(x) = \sum_{i=1}^t f_i(x)$, then

for $i := 1$ to t do

Introduce variable y_{f_i} , such that $y_{f_i} = \text{Relax } f_i(x)$

end of for

Introduce variable y_f , such that $y_f = \sum_{i=1}^t y_{f_i}$

else if $f(x) = g(h(x))$, then

Introduce variable $y_h = \text{Relax } h(x)$

Introduce variable $y_f = \text{Relax } g(y_h)$

end of if

From:

M. Tawarmalani and N. V. Sahinidis (2004), 'Global optimization of mixed-integer

nonlinear programs: A theoretical and computational study', Math. Program., DOI 10.1007/s10107-003-0467-6

Note: there are other bounding schemes discussed, but this appears to be the one that is used

Lower Bounding Subroutines

Algorithm Multilinear_Relax $\prod_{i=1}^t y_{r_i}$ (Recursive Arithmetic)

for $i := 2$ to t do

 Introduce variable $y_{r_1, \dots, r_i} = \text{Bilinear_Relax } y_{r_1, \dots, r_{i-1}} y_{r_i}$.

end of for

Algorithm Fractional_Relax $(\mathbf{f}, \mathbf{g}, \mathbf{h})$ (Product Disaggregation)

Introduce variable y_f

Introduce variable $y_g = \text{Relax } g(x)$

if $h(x) = \sum_{i=1}^t h_i(x)$, then

 for $i := 1$ to t do

 Introduce variable $y_{f, h_i} = \text{Relax } (y_f h_i(x))$

 end of for

 Introduce relation $y_g = \sum_{i=1, \dots, t} y_{f, h_i}$

else

 Introduce variable $y_h = \text{Relax } h(x)$

 Introduce relation $y_g = \text{Bilinear_Relax } y_f y_h$

end of if

Algorithm Bilinear_Relax $y_i y_j$ (Convex/Concave Envelope)

$$\text{Bilinear_Relax } y_i y_j \geq y_i^u y_j + y_j^u y_i - y_i^u y_j^u$$

$$\text{Bilinear_Relax } y_i y_j \geq y_i^l y_j + y_j^l y_i - y_i^l y_j^l$$

$$\text{Bilinear_Relax } y_i y_j \leq y_i^u y_j + y_j^l y_i - y_i^u y_j^l$$

$$\text{Bilinear_Relax } y_i y_j \leq y_i^l y_j + y_j^u y_i - y_i^l y_j^u$$

Algorithm Univariate_Relax $\mathbf{f}(x_j)$ (Recursive Sums and Products)

if $f(x_j) = cx_j^p$ then

 Introduce variable $y_f = \text{Monomial_Relax } cx_j^p$

else if $f(x_j) = cp^{x_j}$ then

 Introduce variable $y_f = \text{Power_Relax } cp^{x_j}$

else if $f(x_j) = c \log(x_j)$ then

 Introduce variable $y_f = \text{Logarithmic_Relax } c \log(x_j)$

else if $f(x_j) = \prod_{i=1}^t f_i(x_j)$, then

 for $i := 1$ to t do

 Introduce variable y_{f_i} , such that $y_{f_i} = \text{Relax } f_i(x)$

 end of for

 Introduce variable y_f , such that $y_f = \text{Multilinear_Relax } \prod_{i=1}^t y_{f_i}$

else if $f(x_j) = \sum_{i=1}^t f_i(x_j)$, then

 for $i := 1$ to t do

 Introduce variable y_{f_i} , such that $y_{f_i} = \text{Relax } f_i(x_j)$

 end of for

 Introduce variable y_f , such that $y_f = \sum_{i=1}^t y_{f_i}$

end of if

Lower Bounding Bilinearities

$$x_i x_j < 0$$

$$(x_i^U - x_i)(x_j^U - x_j) \geq 0$$

$$x_i^U x_j^U - x_i^U x_j - x_i x_j + x_i x_j > 0$$

$$x_i x_j > x_i^U x_j + x_i x_j - x_i^U x_j^U$$

$$x_i x_j > x_i^L x_j + x_i x_j - x_i^L x_j^L$$

So create a new variable to replace the bilinearity

$$w_{ij} < 0$$

$$w_{ij} > x_i^U x_j + x_i x_j - x_i^U x_j^U$$

$$w_{ij} > x_i^L x_j + x_i x_j - x_i^L x_j^L$$

A similar procedure can be performed in the case $x_i x_j > 0$ and for upper bounding.

Lower Bounding Linear Fractional Terms

- The same inequalities used in the bilinear case can be rearranged, with some added technicalities on sign to produce the linear fractional inequalities:

$$\frac{x_i}{x_j} \geq \frac{x_i^U}{x_j} + \frac{x_i}{x_j^L} - \frac{x_i^U}{x_j^L}$$

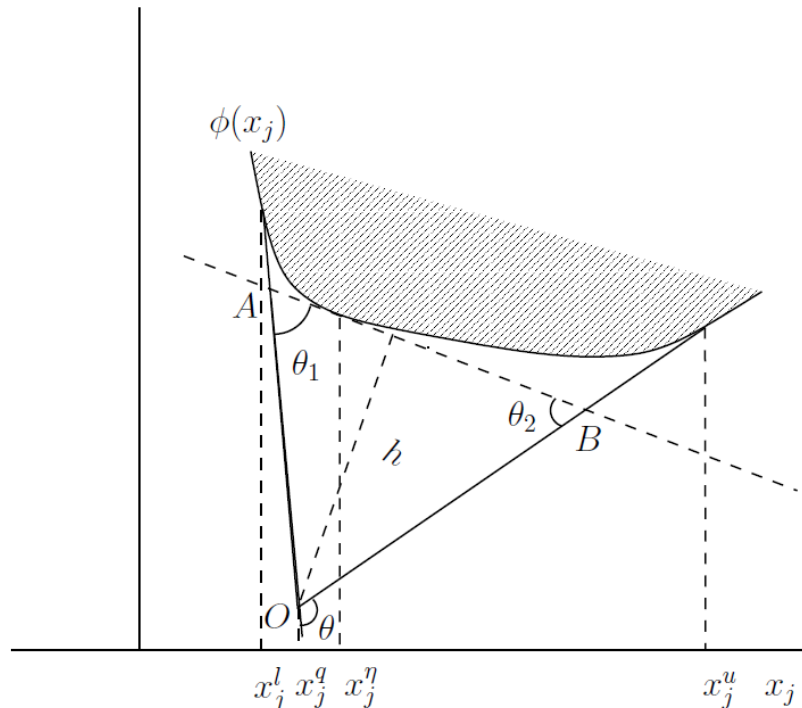
$$\frac{x_i}{x_j} \geq \frac{x_i^L}{x_j} + \frac{x_i}{x_j^U} - \frac{x_i^L}{x_j^U}$$

$$\frac{x_i}{x_j} \leq \frac{x_i^U}{x_j} + \frac{x_i}{x_j^U} - \frac{x_i^U}{x_j^U}$$

$$\frac{x_i}{x_j} \leq \frac{x_i^L}{x_j} + \frac{x_i}{x_j^L} - \frac{x_i^L}{x_j^L}$$

Further Lower Bounding

- Once a lower bound is created, a linear approximation of this bound will often be created to further speed computation.

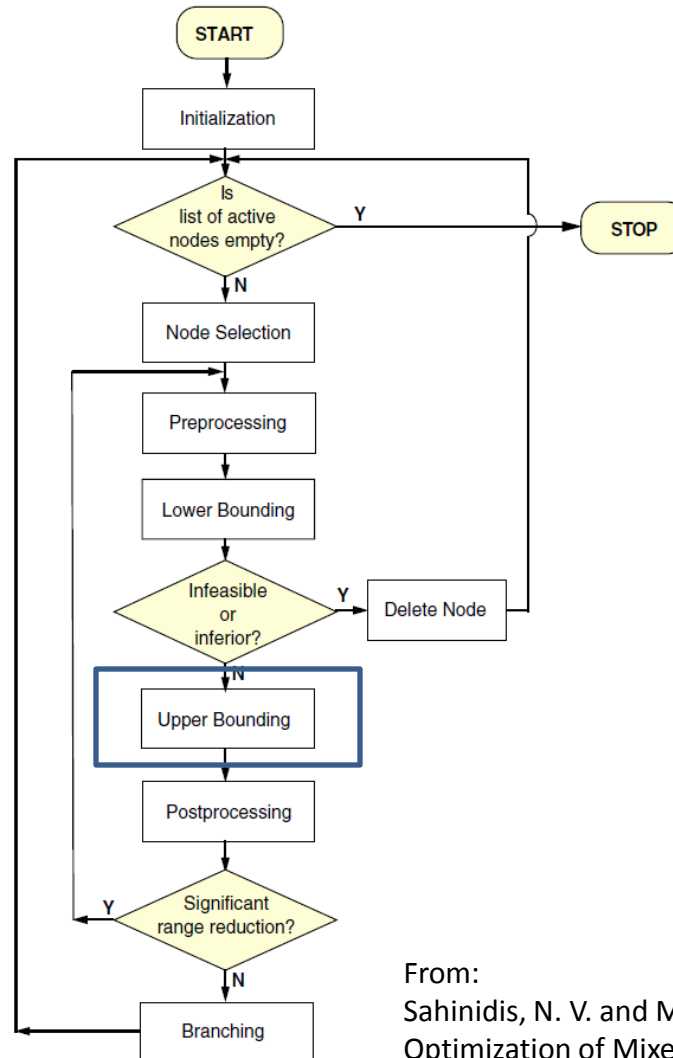


From:

M. Tawarmalani and N. V. Sahinidis
(2004), 'Global optimization of mixed-
integer

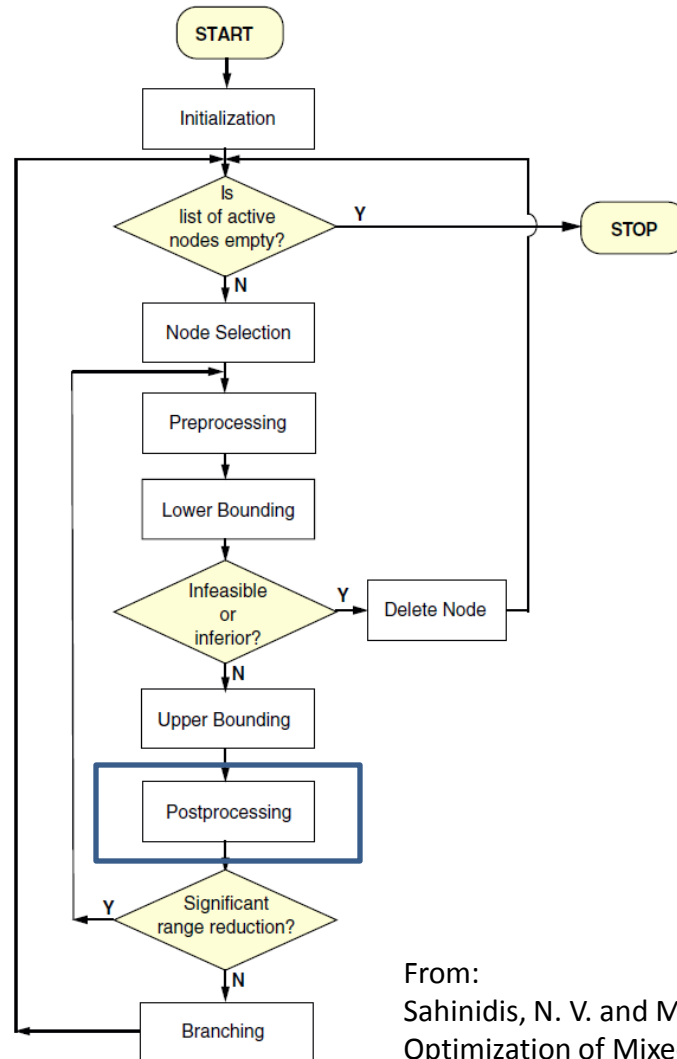
nonlinear programs: A theoretical and
computational study', Math. Program.,
DOI 10.1007/s10107-003-0467-6

Upper Bounding



From:
Sahinidis, N. V. and M. Tawarmalani, BARON 9.0.4: Global Optimization of Mixed-Integer Nonlinear Programs, User's manual, 2010

Post Processing on a Node



From:
Sahinidis, N. V. and M. Tawarmalani, BARON 9.0.4: Global Optimization of Mixed-Integer Nonlinear Programs, User's manual, 2010

Optimality Based Range Reduction

- We will first define three different problems:

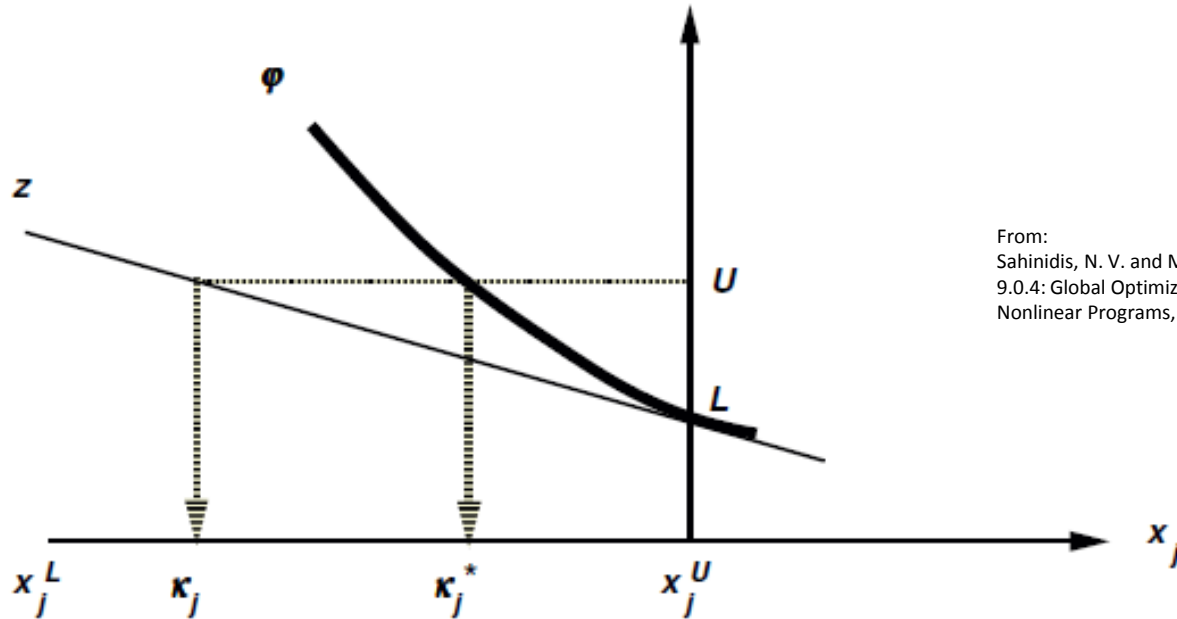
$$\begin{array}{ll} (P) : & \min f(x) \\ & \text{s.t. } g(x) \leq 0 \\ & x \in X \end{array} \quad \text{Original problem}$$

$$\begin{array}{ll} (R) : & \min \bar{f}(\bar{x}) \\ & \text{s.t. } \bar{g}(\bar{x}) \leq 0 \\ & \bar{x} \in \bar{X} \end{array} \quad \text{Relaxed problem}$$

$$\begin{array}{ll} (R_y) : & \varphi(y) = \min f(x) \\ & \text{s.t. } g(x) \leq y \\ & x \in X \end{array} \quad \text{Perturbed problem}$$

Optimality Based Range Reduction: on an active constraint

In this example the upper bound range constraint is active on x



From:
Sahinidis, N. V. and M. Tawarmalani, BARON
9.0.4: Global Optimization of Mixed-Integer
Nonlinear Programs, User's manual, 2010

L: the solution of the relaxed problem R

U: an upper bound on the original problem P

Φ : $\Phi(y)$ unknown solution of $R(y)$ but is L when $y = 0$, as $R(0) = R$.

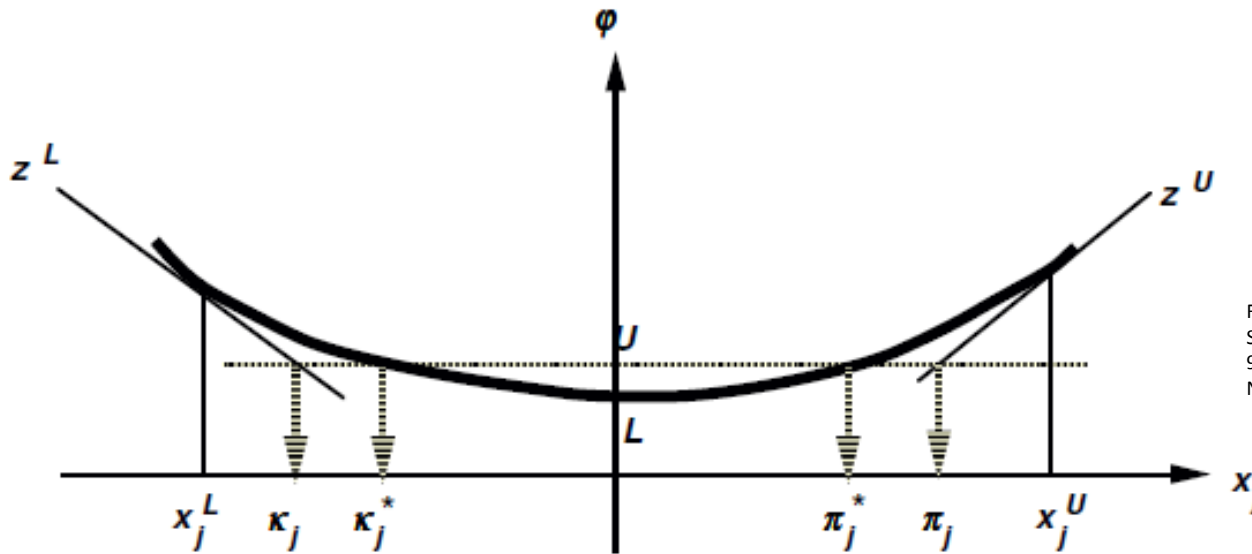
K_j^* : the range reduction if $\Phi(y)$

K_j : the range reduction based on the supporting hyperplane of $\Phi(y)$, z

Z: a supporting hyperplane based on the Lagrange multiplier in the solution of R.

Optimality Based Range Reduction: on an inactive constraint

In this example there is no range constraint active.



From:
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9.0.4: Global Optimization of Mixed-Integer
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L: the solution of the relaxed problem R

U: an upper bound on the original problem P

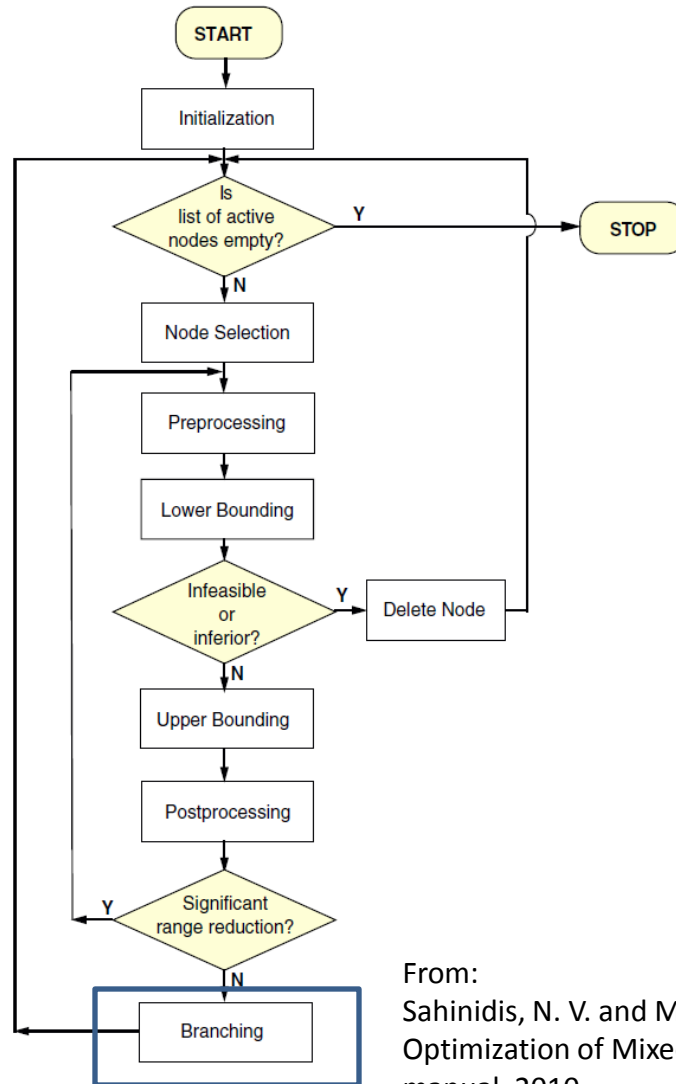
z^L : linear underestimator found by fixing the value of x to be x_j^L

z^U : linear underestimator found by fixing the value of x to x_j^U

κ_j^*/π_j^* : the range reduction if the entire value function were known.

κ_j/π_j : the range reduction based on the underestimators z .

Branching



From:
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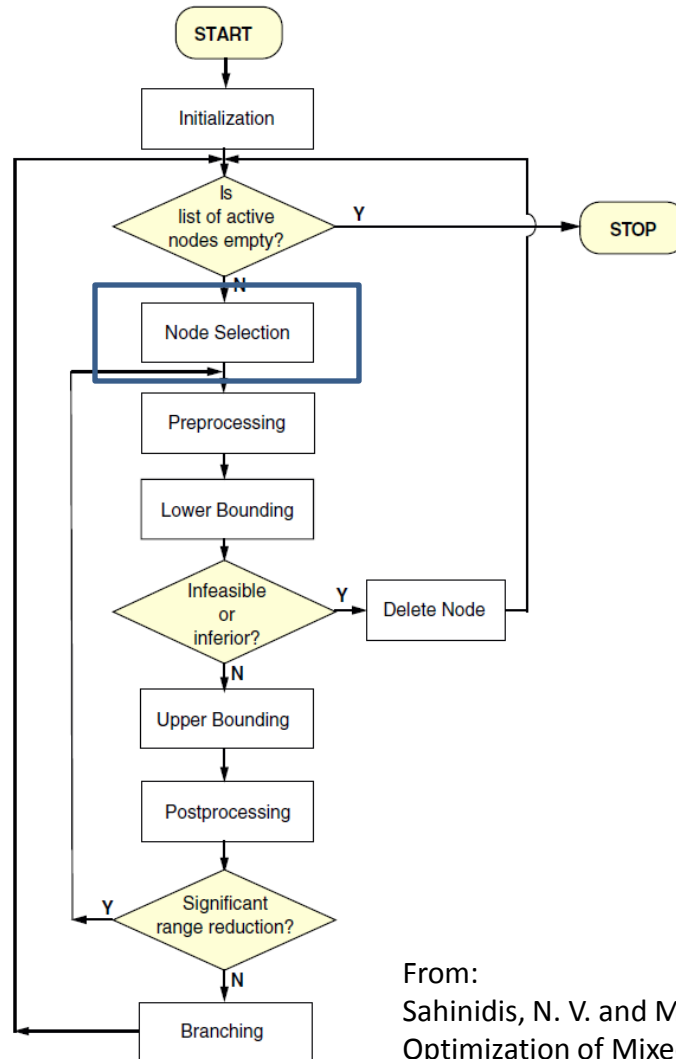
Branching: Variable Selection

- BARON uses a rectangular subdivision scheme, so only a single variable is chosen for each branching.
- The variable chosen to branch is the variable that contributes the most to the “relaxation gap”.
- This is a non trivial process as many additional variables were introduced in our underestimating step

Branching: Point Selection

- Periodically, branch at the midpoint
- Otherwise, branch at the solution of the lower bounding problem.
- This decision is stored, but not executed until the node is selected again.

Node Selection



From:
Sahinidis, N. V. and M. Tawarmalani, BARON 9.0.4: Global Optimization of Mixed-Integer Nonlinear Programs, User's manual, 2010

Node Selection

- Default BARON implementation is a composite value based on lower bound, violation (sum of violations of all variables) and order of creation.
- If memory limits are approached BARON switches to FIFO.
- As with all steps, customizable by the user.

For More Information

- Sahinidis, N. V. and M. Tawarmalani, BARON 9.0.4: Global Optimization of Mixed-Integer Nonlinear Programs, User's manual, 2010
- M. Tawarmalani and N. V. Sahinidis , 'Global optimization of mixed-integer nonlinear programs: A theoretical and computational study, Math Program, DOI 10.1007/s10107-003-0467-6, 2004
- M. Tawarmalani and N. V. Sahinidis, Convexification and global optimization in continuous and mixed-integer nonlinear programming: theory, algorithms, software, and applications, Springer, 2002