

A Unified Framework for Nonlinear Dependence Testing and Symbolic Analysis

Kyle A. Gallivan

Collaboration with: R. van Engelen, J. Birch, B. Walsh, X. Shou

School of Computational Science and Information Technology
Florida State University



Research Interests

- high-performance algorithm technology for multiple applications
- algorithm/architecture interaction
- software support for design/implementation/tuning of high-performance algorithms

Algorithms

- large-scale numerical linear algebra – sparse linear systems, eigenproblems
- real-time numerical linear algebra — adaptive filtering
- model reduction of dynamical systems – novel algorithms and unification theory
- differential equations – highly oscillatory ODEs, hierarchical methods for circuits, PDEs
- inverse problems – autofocus for synthetic aperture radar, blind deconvolution

Algorithms

- encoding and retrieval – text retrieval, image retrieval
- large-scale manifold computations – geodesics, means, optimization
- Current funding
 - NSF ITR ACI0324944 with Purdue and Rice
 - NSF CCR9912415
- Previous – NSF, DARPA, ARO
- Pending – NSF on SAR autofocus with D. Munson U. Michigan

Algorithm/Architecture

- Systolic array design for adaptive filtering
- Cedar – algorithms, applications, performance analysis
- block-based algorithms – BLAS3
- Load/Store performance prediction
- cache models and iterative compilation with Knijnenburg and O'Boyle
- timing estimation and voltage scaling with van Engelen
- Current funding
 - NSF EIA 0072043 with D. Whalley FSU

Software support

- Cache and memory restructuring for uniformly-generated accesses with Gannon and Jalby
- FALCON – MATLAB restructuring compiler with Padua and Gallopoulos
- Chains of recurrences-based restructuring infrastructure with Van Engelen
- Current funding – NSF CCR0105422 with Van Engelen
- Pending funding – NSF Unified Compiler Framework with Van Engelen FSU

Introduction

- Efficient code → restructuring compilers
- efficacy of restructuring depends on power of analysis system
- efficiency of restructuring depends on efficiency of analysis implementation
- current systems are powerful but
 - tend to be collection of independent subsystems with heuristics
 - powerful symbolic processing tends to be slow
 - heuristics are faster but limited
- limitations are still significant

Limitations

- nonlinear symbolic processing
- pointer arithmetic analysis
- conditional control flow
- efficient parameterization of decisions/questions
- Recent commentary
 - Psarris (PC 03, TPDS 04)
 - Franke and O'Boyle (ETAPS CC 01)
 - Wu, Cohen, Hoeflinger Padua (ICS 01)

Unified Framework

Goal: Unified and efficient symbolic analysis system that subsumes most of current technology and extends restructuring capabilities significantly.

- an efficient technique for the detection and substitution of generalized induction variables
- improved array recovery methods for both linear and nonlinear pointer updates
- enhanced value range analysis and array region analysis for conditionally updated and coupled induction variables, and pointers

Unified Framework

- increased accuracy for data dependence analysis for affine and nonlinear array indices, and index expressions involving conditionally updated variables and pointers
- implementation of an efficient nonlinear dependence test that does not require closed forms or code adjustments, e.g. induction variable substitution and array recovery.
- application of technology to compiler support for related embedded systems problems of timing estimation and voltage scaling

Outline

- CR algebra
- Induction variables
- Pointer arithmetic and Array Recovery
- Monotonicity
- Value range analysis
- Array region analysis
- Nonlinear dependence analysis
- Conditional analysis
- Parametric time and power estimation

CR Algebra

- Chains of Recurrences (CR) algebra developed by E. Zima and O. Bachmann
- Extended with new algebraic rules by Van Engelen [ETAPS CC'01]
- Composition of CRs is closed under $+$ and \times
- CRs subsume functions of form $\chi(n) = p(n) + \alpha r^n$, i.e., generalized induction variables (GIVs)
- Normal form for CR expression comparison Van Engelen [FSU CS TR '00]

CR Algebra

- Set of complete rewriting rules for CR \leftrightarrow closed form
- tools to create CRs and their closed forms for high level code and RTL available and efficient
Birch [MS'03] and Walsh '04
- unifies the production of compile-time symbolic decisions that drive restructuring and run-time parameterized decisions for multiversion code.

Basic Function Form

A function f can be rewritten as an system of recurrence relations in terms of $f_0, f_1, f_2, \dots, f_k$. Evaluating the function represented by f_j for $j = 0, 1, \dots, k - 1$ within a loop $i \geq 0$,

$$f_j(i) = \begin{cases} \iota_j, & \text{if } i = 0 \\ f_j(i - 1) \odot_{j+1} f_{j+1}(i - 1), & \text{if } i > 0 \end{cases}$$

with $\odot_{j+1} \in \{+, *\}$, and ι_j representing a loop invariant expression. f_k is either loop invariant or another CR function.

Linear Function

Loop counter $i \geq 0$ and induction variable j with addition of constant h

```
for ( $i = 0; i < N; i++$ ) {  
     $j = j + h;$   
}
```

Variable	CR	function
i	$\{0, +, 1\}_i$	i
j	$\{j_0, +, h\}_i$	$j_0 + i * h$

Geometric Series

Loop counter $i \geq 0$ and induction variable j with multiplication by constant h

```
for ( $i = 0; i < N; i++$ ) {  
     $j = j * h;$   
}
```

Variable	CR	function
i	$\{0, +, 1\}_i$	i
j	$\{j_0, *, h\}_i$	$j_0 * h^i$

Multivariate Affine

Loop counters i_1 and i_2 and induction variable j with addition of constant h .

```
for ( $i_1 = 0; i_1 \leq N; i_1++$ ) {  
    for ( $i_2 = 0; i_2 < M; i_2++$ ) {  
         $j = j + h;$   
    }  
}
```

Variable	CR	function
i_1	$\{0, +, 1\}_{i_1}$	i_1
i_2	$\{0, +, 1\}_{i_2}$	i_2
j	$\{\{j_0, +, M * h\}_{i_1}, +, h\}_{i_2}$	$j + (M * h * i_1) + h * i_2$

Multivariate Exponential

Loop counters i_1 and i_2 and induction variable j with multiplication of constant h .

```
for ( $i_1 = 0$ ;  $i_1 \leq N$ ;  $i_1++$ ) {  
    for ( $i_2 = 0$ ;  $i_2 < M$ ;  $i_2++$ ) {  
         $j = j * h$ ;  
    }  
}
```

Variable	CR	function
i_1	$\{0, +, 1\}_{i_1}$	i_1
i_2	$\{0, +, 1\}_{i_2}$	i_2
j	$\{\{j_0, *, h^M\}_{i_1}, *, h\}_{i_2}$	$j * h^{i_2} * h^{i_1 * M}$

CR Algebra Rules

- Basic CR is a linear function over $i \in \mathbb{N}$:

$$f(i) = x_0 + i h \equiv \{x_0, +, h\}_i$$

- ... or a geometric series over $i \in \mathbb{N}$:

$$g(i) = x_0 * h^i \equiv \{x_0, *, h\}_i$$

- Some example algebraic identities:

$$c + \{x_0, +, h\}_i \Leftrightarrow \{c + x_0, +, h\}_i$$

$$c * \{x_0, +, h\}_i \Leftrightarrow \{c x_0, *, c h\}_i$$

$$\{x_0, +, h\}_i + \{y_0, +, g\}_i \Leftrightarrow \{x_0 + y_0, +, g + h\}_i$$

$$\{x_0, +, h\}_i * \{y_0, +, g\}_i \Leftrightarrow \{x_0 y_0, +, g h + g x + h y, +, 2 g h\}_i$$

Stride/Start Nesting

$$\Phi_i = \{\phi_0, \odot_1, \{\phi_1, \odot_2, \dots, \{\phi_{k-1}, \odot_k, f_k\}_i \dots\}_i\}_i$$

$$\Phi_i = \{\{\dots \{\phi_0, \odot_1, f_1\}_i, \odot_2, f_2\}_i, \dots\}_i, \odot_k, f_k\}_i$$

Multiply nested loops generate typically one of these two nested CRs

- recursively specify initial condition of loop by CR for next outer loop
- recursively specify stride of loop by CR for next inner loop

Flattened CR and Loop Forms

All forms can be flattened into a single flattened tuple

$$\Phi_i = \{\phi_0, \odot_1, \phi_1, \odot_2, \dots, \odot_k, f_k\}_i$$

with $k = L(\Phi_i)$ the length of the CR.

$$F[0] := \phi_0; cr_1 := \phi_1; \dots; cr_{k-1} := \phi_{k-1}; cr_k := f_k$$

for $i = 1$ **to** n **do**

$$F[i] := F[i - 1] \odot_1 cr_1$$

$$cr_1 := cr_1 \odot_2 cr_2$$

:

$$cr_{k-1} := cr_{k-1} \odot_k cr_k$$

od

Normal Forms

$$\Phi_i = \{\phi_0, \odot_1, \phi_1, \odot_2, \dots, \odot_k, f_k\}_i$$

- Φ_i is *polynomial* or *pure-sum CR* if $\odot_j = +$, $j = 1, \dots, k$.
- Φ_i is *exponential* or *pure-product CR* if $\odot_j = *$, for all $j = 1, \dots, k$.
- $\Phi_i = \{\phi_0, *, f_1\}_i$ is *geometric* if f_1 is i -loop invariant.
- Φ_i is a *GIV* if $\odot_j = +$ for $j = 1, \dots, k - 1$ and $\odot_k = *$.
- $\Phi_i = \{\phi_0, *, \phi_1, +, f_2\}_i$ is a *factorial CR* if $\phi_1 \geq 1$ and $f_2 = 1$, or $\phi_1 \leq -1$ and $f_2 = -1$.

Example

Expression $n * j + i + 2 * k + 1$, unit-stride index variables $i \geq 0$ and $j \geq 0$ and GIV $k(i) = (i^2 - i) / 2$ and $\Phi(k) = \{0, +, 0, +, 1\}_i$.

Replace i , j and k then normalize.

$$\begin{aligned} & \mathcal{CR}(\mathcal{CR}(\mathcal{CR}(n * j + i + 2 * k + 1))) \\ &= \mathcal{CR}(\mathcal{CR}(n * j + \{0, +, 1\}_i + 2 * k + 1)) \\ &= \mathcal{CR}(n * \{0, +, 1\}_j + \{0, +, 1\}_i + 2 * k + 1) \\ &= n * \{0, +, 1\}_j + \{0, +, 1\}_i + 2 * \{0, +, 0, +, 1\}_i + 1 \\ &= \{\{1, +, n\}_j, +, 1, +, 2\}_i \end{aligned}$$

Closed Forms

- Polynomial, factorial, GIV, and exponential CRs can always be converted to a closed-form. However, some CRs do not have equivalent closed-forms.
- Rewriting rules terminate with a closed form of the CR expression or showing that one does not exist.
- Polynomial closed form is easily constructed and evaluated in $O(k^2)$ time

$$\chi(i) = \sum_{j=0}^k \phi_j \binom{i}{j}$$

Induction Variable Analysis

- Detect (coupled) GIV recurrences RVE [ETAPS CC'01]:

```
for  $i = 1$  to  $n$  do  
   $j = j+k$   
   $k = k+1$ 
```

⇓ CR

```
for  $i = 1$  to  $n$  do  
   $j \equiv \{j_0, +, k_0, +, 1\}_i$   
   $k \equiv \{k_0, +, 1\}_i$ 
```

⇓ Closed form

```
for  $i = 1$  to  $n$  do  
   $j \equiv j_0 + i k_0 + \frac{i^2-i}{2}$   
   $k \equiv k_0 + i$ 
```

Induction Variable Substitution

```
int i, j = 0, k = 0, m = 1;
for (i = a; i <= b; i++) {
    ...
    A[f(i,j,k,m)] = A[g(i,j,k,m)];
    ...
    j = j + k;
    k = k + 1;
    m = m << 1;
    ...
}
```

(a) *Original*

```
int i;
for (i = 0; i <= b-a; i++) {
    ...
    A[f(i+a, (i2-i)/2, i, 1<<i)]
    = A[g(i+a, (i2-i)/2, i, 1<<i)];
    ...
}
```

(b) *After IVS*

Note nonlinear indices result.

TRED Loop

```
ijkl=0
ij=0
DO i=1,m
  DO j=1,i
    ij=ij+1
    ikl=ijkl+i-j+1
    DO k=i+1,m
      DO l=1,k
        ikl=ijkl+1
        xijkl[ijkl]=xkl[l]
      ENDDO
    ENDDO
    ikl=ijkl+ij+left
  ENDDO
ENDDO
```

(a) *Original*

```
DO i=0,m-1
  DO j=0,i-1
    DO k=0,m-i-2
      DO l=0,k
        xijkl[Φ(ijkl)]=xkl[Φ(l)]
      ENDDO
    ENDDO
  ENDDO
ENDDO
```

(b) *GIV Analyzed*

TRFD CRs and Closed Forms

$$\begin{aligned}\Phi(ijkl) = & \{ \{ \{ \{ 2, +, \text{left} + m(m+1)/2 + 2, +, \text{left} + m(m+1)/2 + 1 \}_i \\ & , +, \text{left} + m(m+1)/2 \}_j \\ & , +, \{ 2, +, 1 \}_i, +, 1 \}_k \\ & , +, 1 \}_l\end{aligned}$$

and

$$\Phi(l) = \{ 1, +, 1 \}_l$$

$$\begin{aligned}\mathcal{CR}^{-1}(\Phi(ijkl)) = & l + i * (k + (m + m^2 + 2 * \text{left} + 6)/4) \\ & + j * (\text{left} + (m + m^2)/2) \\ & + ((i * m)^2 + m * i^2)/4 \\ & + (k^2 + 3 * k + i^2 * (\text{left} + 1))/2 + 2\end{aligned}$$

Pointers and Array Recovery

- CR-GIV for pointers RVE/KAG [IWIA'01]
- Array recovery for dependence analysis or data layout
- Facilitates dependence analysis without array recovery

	<i>Access</i>	<i>PAD</i>
1	$a[i]$	$\{a, +, 1\}_i$
2	$a[2*i+1]$	$\{a+1, +, 2\}_i$
3	$a[(i*i-i)/2]$	$\{a, +, 0, +, 1\}_i$
4	$a[1<<i]$	$\{a+1, +, 1, *, 2\}_i$
5	$p++$	$\{p_0, +, 1\}_i$
6	$q-=2$	$\{q_0, +, -2\}_i$
7	$r+=i$	$\{r_0, +, 0, +, 1\}_i$
8	$s+=2*i$	$\{s_0, +, 0, +, 2\}_i$

GSM Speech CODEC

```
int *f = ..., *lsp = ...;
...
f += 2; lsp += 2;
for (i = 2; i <= 5; i++) {
    *f = f[-2];
    for (j = 1; j < i; j++, f--)
        *f += f[-2]-2*(*lsp)*f[-1];
    *f -= 2*(*lsp);
    f += i; lsp += 2;
}
```

(a) *Original*

```
int *f = ..., *lsp = ...;
...
for (i = 0; i <= 3; i++) {
    *({f+2, +, 1}_i) = *({f, +, 1}_i);
    for (j = 0; j <= {0, +, 1}_i; j++)
        *({{f+2, +, 1}_i, +, -1}_j) += *({{f, +, 1}_i, +, -1}_j)
        - 2 * (*{lsp+2, +, 2}_i) * (*{{f, +, 1}_i, +, -1}_j);
    *(f+1) -= 2*{lsp+2, +, 2}_i;
}
```

(b) *Analyzed*

ETSI Lsp_Az Code Segment

GSM Speech CODEC

```
int *f = ..., *lsp = ...;
...
for (i = 0; i <= 3; i++) {
    f[i+2] = f[i];
    for (j = 0; j <= i; j++)
        f[i-j+2] += f[i-j] - 2*lsp[2*i+2]*f[i-j+1];
    f[1] -= 2*lsp[2*i+2];
}
```

(c) *Transformed*

Step Functions

Definition: The *initial value* $V\Phi_i$ of a CR form Φ_i is

$$V\Phi_i = V\{\phi_0, \odot_1, \dots, \odot_k, \phi_k\}_i = \phi_0$$

Definition: The *step function* $\Delta\Phi_i$ of a CR form Φ_i is

$$\Delta\Phi_i = \Delta\{\phi_0, \odot_1, \phi_1, \odot_2, \dots, \odot_k, \phi_k\}_i = \{\phi_1, \odot_2, \dots, \odot_k, \phi_k\}_i$$

Definition: The *direction-wise step function* $\Delta_j\Phi_i$ of a multi-variate CR form Φ_i is the step function with respect to an index variable j

$$\Delta_j\Phi_i = \begin{cases} \Delta\Phi_i & \text{if } i = j \\ \Delta_j V\Phi_i & \text{otherwise} \end{cases}$$

Monotonicity

$$\Delta_i \Phi(ijkl) = \{\text{left} + m(m+1)/2 + 2, +, \text{left} + m(m+1)/2 + 1\}_i$$

$$\Delta_j \Phi(ijkl) = \text{left} + m(m+1)/2$$

$$\Delta_k \Phi(ijkl) = \{\{2, +, 1\}_i, +, 1\}_k$$

$$\Delta_l \Phi(ijkl) = 1$$

- $i \geq 0$ and $k \geq 0 \rightarrow \Delta_k > 0$ and $\Delta_l > 0$
- In the k, l direction the addressing of the $xijkl[ijkl]$ is monotonically increasing
- The inner k, l loop nest can be parallelized.
- $ijkl$ over i, j, k, l index space is monotonically increasing if $\text{left} > m(m+1)/2$.
- Closed-forms steps give runtime tests for multiversion code

Value Range Analysis

The accuracy of a range analyzer depends on how the analyzed expression is formed and if monotonicity can be exploited

Blume/Eigenman [IPPS'95], Haghghat [Kluwer'95]

Example without Monotonicity:

$i(i-1)$ and $2^i - i$ are monotonic and non-negative for $0 \leq i \leq n$.

Without using monotonic properties, bounds can be very loose,

$$[0, n]([0, n] - 1) = [0, n][-1, n-1]$$

$$= [-n, n^2 - n]$$

$$2^{[0, n]} - [0, n] = [1 - n, 2^n]$$

Value Range Analysis

- Techniques such as symbolic forward differencing can be used to detect monotonicity but very costly and not always safe.
- CR unified approach generates normal forms and is straightforward to integrate in a compiler framework.
- calculate CR normal forms on index space for GIVs; use the step functions to test monotonicity of expressions for value range analysis.

CR Bounds

Lower and upper CR bounds $L\Phi_i$ and $U\Phi_i$

$$L\Phi_i = \begin{cases} LV\Phi_i & \text{if } LM\Phi_i \geq 0 \\ LCR_i^{-1}(\Phi_i)[i \leftarrow n] & \text{if } UM\Phi_i \leq 0 \\ LCR_i^{-1}(\Phi_i) & \text{otherwise} \end{cases}$$

$$U\Phi_i = \begin{cases} UV\Phi_i & \text{if } UM\Phi_i \leq 0 \\ UCR_i^{-1}(\Phi_i)[i \leftarrow n] & \text{if } LM\Phi_i \geq 0 \\ UCR_i^{-1}(\Phi_i) & \text{otherwise} \end{cases}$$

$$M\Phi_i = \begin{cases} \Delta\Phi_i & \text{if } \odot_1 = + \\ \Delta\Phi_i - 1 & \text{if } \odot_1 = * \wedge LV\Phi_1 \geq 0 \wedge L\Delta\Phi_i > 0 \\ 1 - \Delta\Phi_i & \text{if } \odot_1 = * \wedge UV\Phi_1 < 0 \wedge L\Delta\Phi_i > 0 \\ \text{undefined} & \text{otherwise} \end{cases}$$

Handles GIVs, multivariate polynomial expressions, and shifts

Value Range Analysis Example

Assume $0 \leq i \leq n, 0 \leq j \leq m, k = (i^2 - i)/2$.

$$f(i, j, k) = n * j + i + 2 * k + 1$$

$$\mathcal{CR}(f) = \{\{1, +, n\}_j, +, 1, +, 2\}_i$$

The bounds can be derived by straightforward application of the rules:

$$\begin{aligned} & [L\{\{1, +, n\}_j, +, 1, +, 2\}_i , U\{\{1, +, n\}_j, +, 1, +, 2\}_i] \\ & = [L\{1, +, n\}_j , U(\{1, +, n\}_j + n^2)] \\ & = [1 , 1 + m n + n^2] \end{aligned}$$

Array Region Analysis

Bounds of accesses to an array region facilitate many code optimizations

- eliminating dynamic array bounds checking
- locality optimizations
- prefetching
- blocking

Array region analysis is also used by methods for dependence testing.

Array Region for Lsp_Az

$$\begin{aligned} & L\{\{f+2, +, 1\}_i, +, -1\}_j \\ &= LCR_j^{-1}(\{\{f+2, +, 1\}_i, +, -1\}_j)[j \leftarrow \{0, +, 1\}_i] \\ &= L(\{f+2, +, 1\}_i - \{0, +, 1\}_i) \\ &= f+2 \end{aligned}$$

$$\begin{aligned} & U\{\{f+2, +, 1\}_i, +, -1\}_j \\ &= U\{f+2, +, 1\}_i \\ &= UCR_i^{-1}\{f+2, +, 1\}_i[i \leftarrow 3] \\ &= f+5 \end{aligned}$$

- bounds $f[2..5]$
- iteration space is triangular and the bound on the inner loop variable $j \leq i$ is given by the CR form $\{0, +, 1\}_i$

Nonlinear dependence analysis

```
p = q = A;  
for (i = 0; i < 10; i++) {  
    for (j = 0; j <= i; j++)  
        *q += *++p;  
    q++;  
}
```

(a) *Loop Nest*

$$\{0, +, 1\}_{i^d} = \{\{1, +, 1, +, 1\}_{i^u}, +, 1\}_{j^u}$$

$$0 \leq i^d \leq 9$$

$$0 \leq i^u \leq 9$$

$$0 \leq j^d \leq i^d$$

$$0 \leq j^u \leq i^u$$

(b) *Dependence Equations*

CR-based Banerjee Bounds Test

- CR-based extreme value test
- direction vector hierarchy by performing subscript-by-subscript testing in multidimensional loops

$$\{\{\{-1, +, 1\}_{i^d}, +, -1, +, -1\}_{i^u}, +, -1\}_{j^u} = 0$$

fbw dependence $\rightarrow i^d < i^u$ and $j^d < j^u$ and

$$\left. \begin{array}{l} \left. \begin{array}{l} 1 \\ \{1, +, 1\}_{i^d} \end{array} \right\} \leq i^u \leq 9 \\ \left. \begin{array}{l} 1 \\ \{1, +, 1\}_{j^d} \end{array} \right\} \leq j^u \leq \{0, +, 1\}_{i^u} \end{array} \right| \begin{array}{l} 0 \leq i^d \leq \begin{cases} 8 \\ \{-1, +, 1\}_{i^u} \end{cases} \\ 0 \leq j^d \leq \begin{cases} \{-1, +, 1\}_{i^d} \\ \{-1, +, 1\}_{j^u} \end{cases} \end{array}$$

Lower Bound

$$\begin{aligned} & L\{\{\{-1, +, 1\}_{id}, +, -1, +, -1\}_{iu}, +, -1\}_{ju} \\ &= L(\{\{\{-1, +, 1\}_{id}, +, -1, +, -1\}_{iu} - j^u\}[j^u \leftarrow \{0, +, 1\}_{iu}]) \\ &= L(\{\{\{-1, +, 1\}_{id}, +, -1, +, -1\}_{iu} - \{0, +, 1\}_{iu}\} \quad (\text{subst.}) \\ &= L(\{\{\{-1, +, 1\}_{id}, +, -2, +, -1\}_{iu}\} \quad (\text{simplify}) \\ &= L(\{\{-1, +, 1\}_{id} - (3i^u - (i^u)^2)/2\}[i^u \leftarrow 9]) \\ &= L\{-55, +, 1\}_{id} \quad (\text{subst.}) \\ &= -55 \end{aligned}$$

Upper Bound

$$\begin{aligned} & U\{\{\{-1, +, 1\}_{i^d}, +, -1, +, -1\}_{i^u}, +, -1\}_{j^u} \\ &= U\{\{-1, +, 1\}_{i^d}, +, -1, +, -1\}_{i^u} \\ &= U\{\{-1, +, -1, +, -1\}_{i^u}, +, 1\}_{i^d} && \text{(swap)} \\ &= U((\{-1, +, -1, +, -1\}_{i^u} + i^d)[i^d \leftarrow \{-1, +, 1\}_{i^u}]) \\ &= U(\{-1, +, -1, +, -1\}_{i^u} + \{-1, +, 1\}_{i^u}) && \text{(subst.)} \\ &= U\{-2, +, 0, +, -1\}_{i^u} && \text{(simplify)} \\ &= -2 \end{aligned}$$

no flow dependence ($i^d < i^u$ and $j^d < j^u$) since

$$0 \notin [-55 \dots -2]$$

Conditional Analysis

- extended algorithms to handle conditionally updated variables that may or may not have closed forms
- a *set of* CR forms or PADs for a variable is determined (one for each path)
- combine the set of CR forms to bound the possible range of values of the conditionally updated variables
- use *min* and *max* bounding functions for a set of CR forms over an index space.

Min/Max

Definition: Let $\{\Phi_i^1, \dots, \Phi_i^n\}$ be a set of n multi-variate polynomial CR forms over i .

The *minimum* CR form is defined by

$$\begin{aligned} \min(\Phi_i^1, \dots, \Phi_i^n) = \\ \{ \min(V\Phi_i^1, \dots, V\Phi_i^n), +, \min(\Delta\Phi_i^1, \dots, \Delta\Phi_i^n) \}_i \end{aligned}$$

The *maximum* CR form is defined by

$$\begin{aligned} \max(\Phi_i^1, \dots, \Phi_i^n) = \\ \{ \max(V\Phi_i^1, \dots, V\Phi_i^n), +, \max(\Delta\Phi_i^1, \dots, \Delta\Phi_i^n) \}_i \end{aligned}$$

Conditional Analysis Example

```

p = A; q = B; k = 0; m = 0;
for (i = 0; i <= n; i++)
    if (C[k+2]) { // Path 1
        for (j = 0; j < i; j++)
            *p++ = *q++;
        k += 2;
    } else { // Path 2
        *p++ = 0;
        q += i;
        k += m;
        m += 2;
    }

```

A_{p_1} (Path 1) CRs

$p = \{A, +, 0, +, 1\}_i$

$q = \{B, +, 0, +, 1\}_i$

$k = \{0, +, 2\}_i$

$m = 0$

A_{p_2} (Path 2) CRs

$p = \{A, +, 1\}_i$

$q = \{B, +, 0, +, 1\}_i$

$k = 0$

$k = \{0, +, 0, +, 2\}_i$

$m = \{0, +, 2\}_i$

Minimum CRs

$p \geq A$

$q \geq \{B, +, 0, +, 1\}_i$

$k \geq 0$

$m \geq 0$

Maximum CRs

$p \leq \{A, +, 1, +, 1\}_i$

$q \leq \{B, +, 0, +, 1\}_i$

$k \leq \{0, +, 2, +, 2\}_i$

$m \leq \{0, +, 2\}_i$

Conditional Analysis Example

<u>A_{p_1} (Path 1) CRs</u>	<u>Minimum CRs</u>	<u>Minimum Closed</u>
$p = \{A, +, 0, +, 1\}_i$	$p \geq A$	$p \geq A$
$q = \{B, +, 0, +, 1\}_i$	$q \geq \{B, +, 0, +, 1\}_i$	$q \geq \&B[(i^2 - i)/2]$
$k = \{0, +, 2\}_i$	$k \geq 0$	$k \geq 0$
$m = 0$	$m \geq 0$	$m \geq 0$
<u>A_{p_2} (Path 2) CRs</u>	<u>Maximum CRs</u>	<u>Maximum Closed</u>
$p = \{A, +, 1\}_i$	$p \leq \{A, +, 1, +, 1\}_i$	$p \leq \&A[(i^2 + i)/2]$
$q = \{B, +, 0, +, 1\}_i$	$q \leq \{B, +, 0, +, 1\}_i$	$q \leq \&B[(i^2 - i)/2]$
$k = 0$	$k \leq \{0, +, 2, +, 2\}_i$	$k \leq i^2 + i$
$k = \{0, +, 0, +, 2\}_i$	$m \leq \{0, +, 2\}_i$	$m \leq 2i$
$m = \{0, +, 2\}_i$		

Conditional Analysis Example

```
p = A; q = B; k = 0; m = 0;
```

```
for (i = 0; i <= n; i++)
```

```
    if (C[k+2]) { // Path 1
```

```
        for (j = 0; j < i; j++)
```

```
            *p++ = *q++;
```

```
            k += 2;
```

```
        } else { // Path 2
```

```
            *p++ = 0;
```

```
            q += i;
```

```
            k += m;
```

```
            m += 2;
```

```
    }
```

$\Delta_i \Phi_i$ Strides

$A[] = 0, \{0, +, 1\}_i$

$B[] = \{0, +, 1\}_i$

$C[] = 0, \{0, +, 2\}_i$

$L\Phi_i..U\Phi_i$ Bounds

$A[0..(n^2+n/2)]$

$B[0..(n^2-n/2)]$

$C[2..n^2+n+2]$

Dynamic Voltage Scaling

$$T_{dead} = \frac{1}{f} * RWEC$$

$$E_N \approx V^2(N c_L + c_R)$$

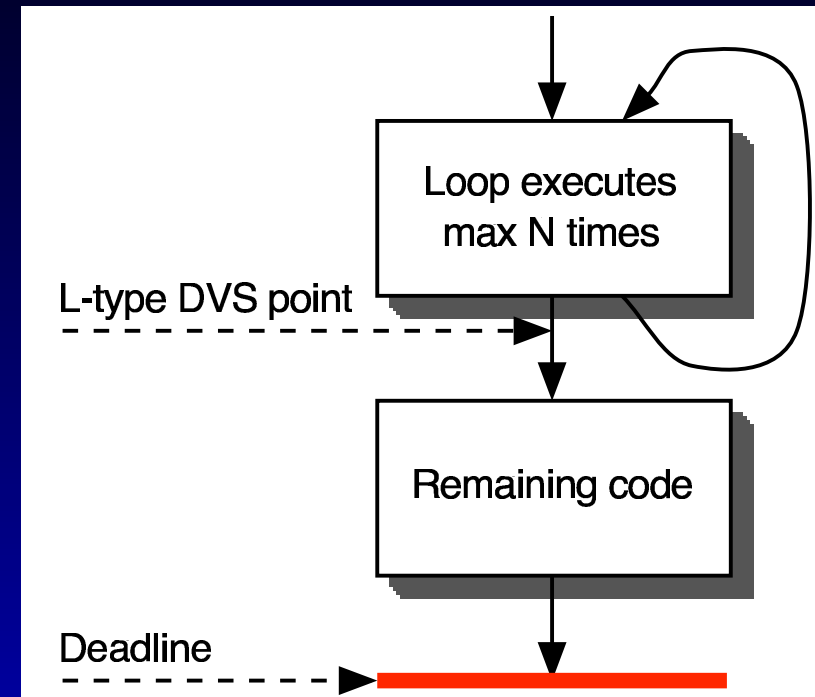
$$E_{act} \approx V^2(n c_L + c_R)$$

$$E_{Ltype} \approx V^2(n c_L + \gamma^2 c_R)$$

$$\gamma = \left(\frac{c_R}{(N-n)c_L + c_R} \right)$$

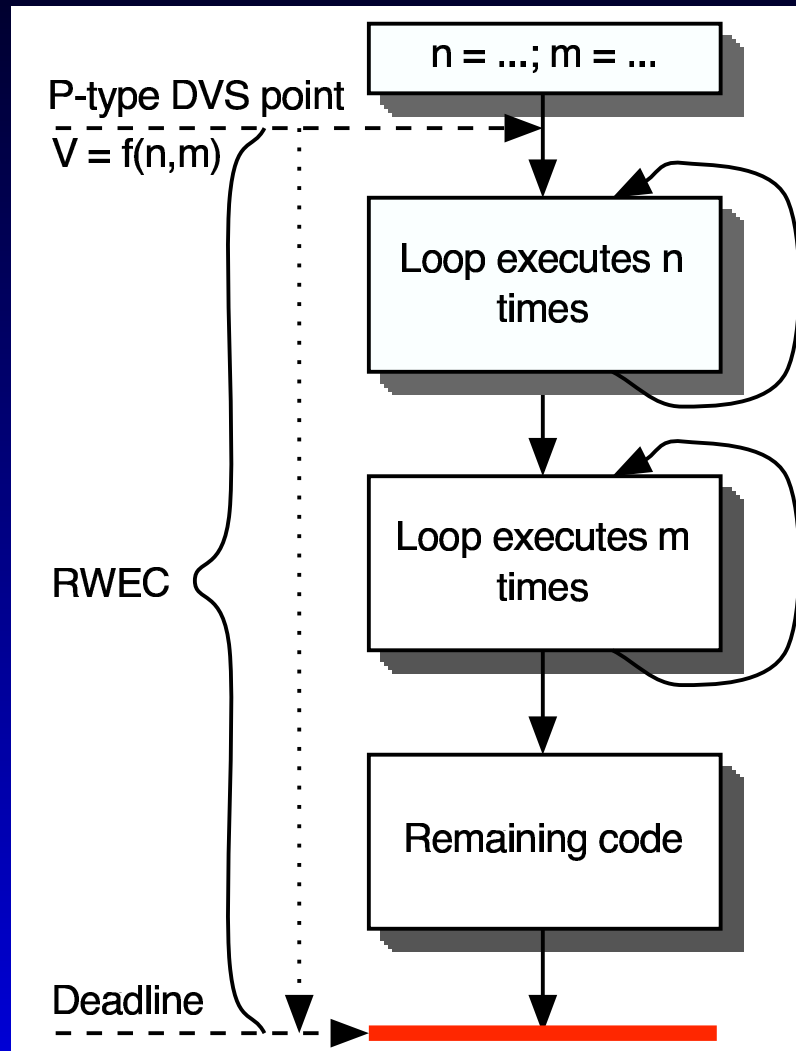
$$E_{Ptype} \approx \left(V \frac{n c_L + c_R}{N c_L + c_R} \right)^2 (n c_L + c_R)$$

$$E_{Ptype} \leq E_{Ltype} \leq E_{act}$$



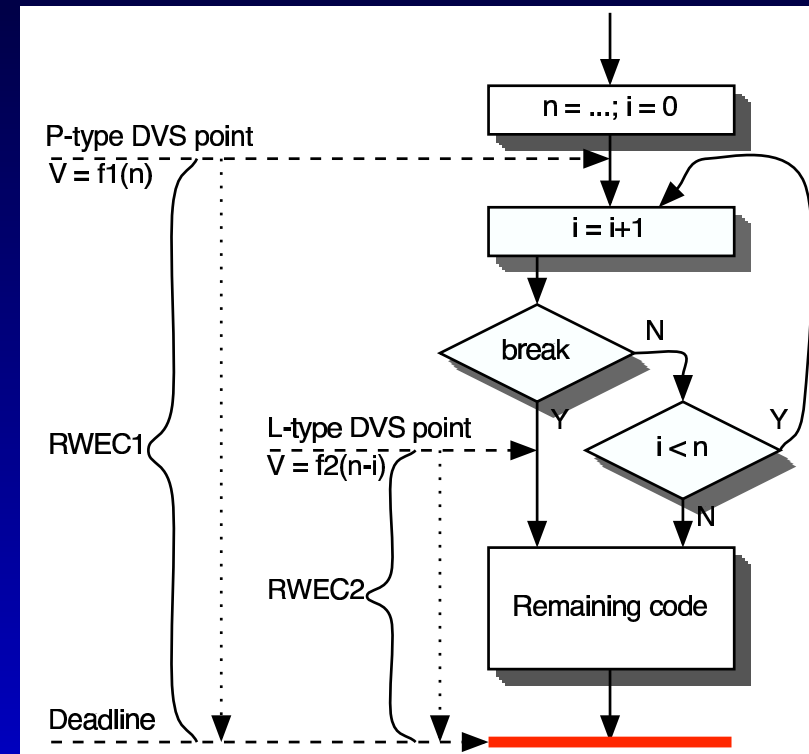
L-Type DVS
Shin/Kim [DAC'01]

P-Type DVS



P-Type DVS

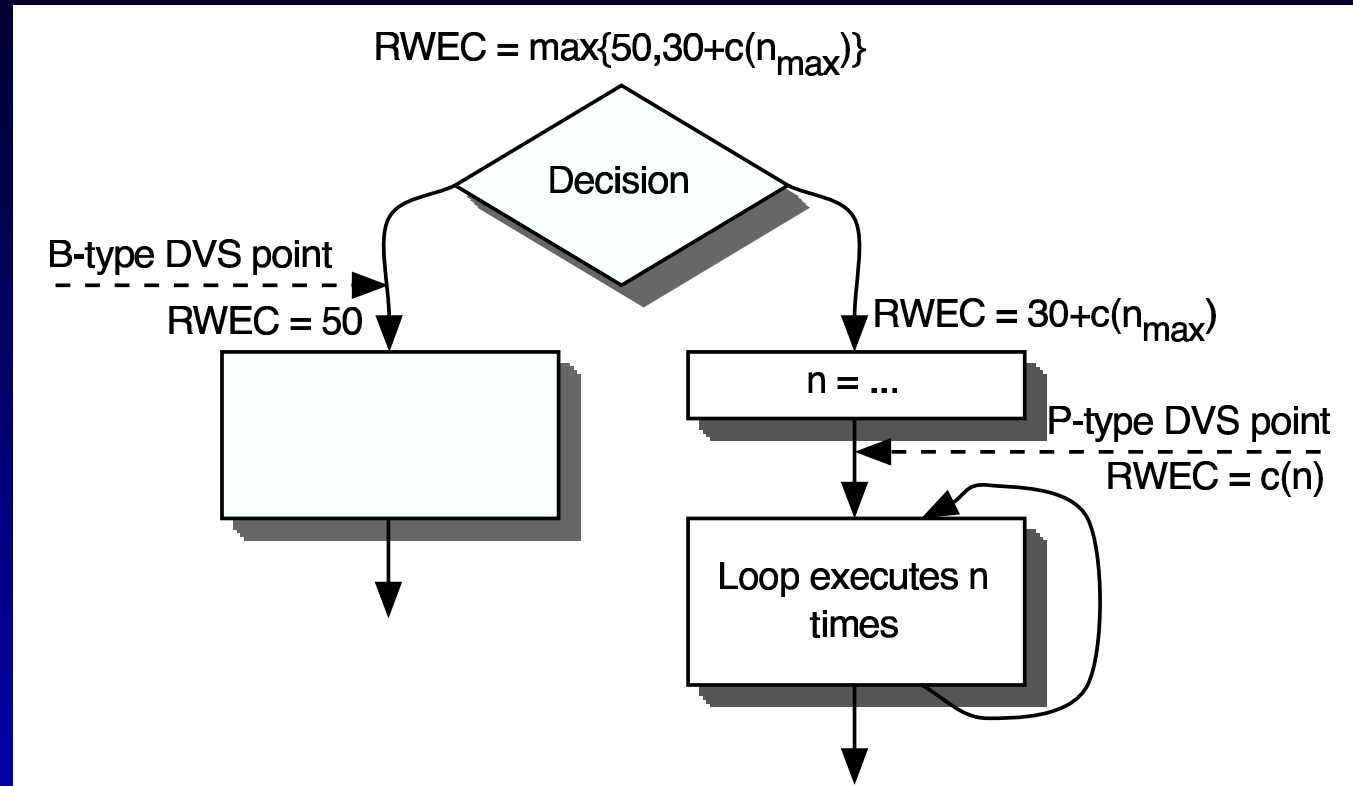
RVE/KAG/BW [COLP'03]



L/P-Type DVS

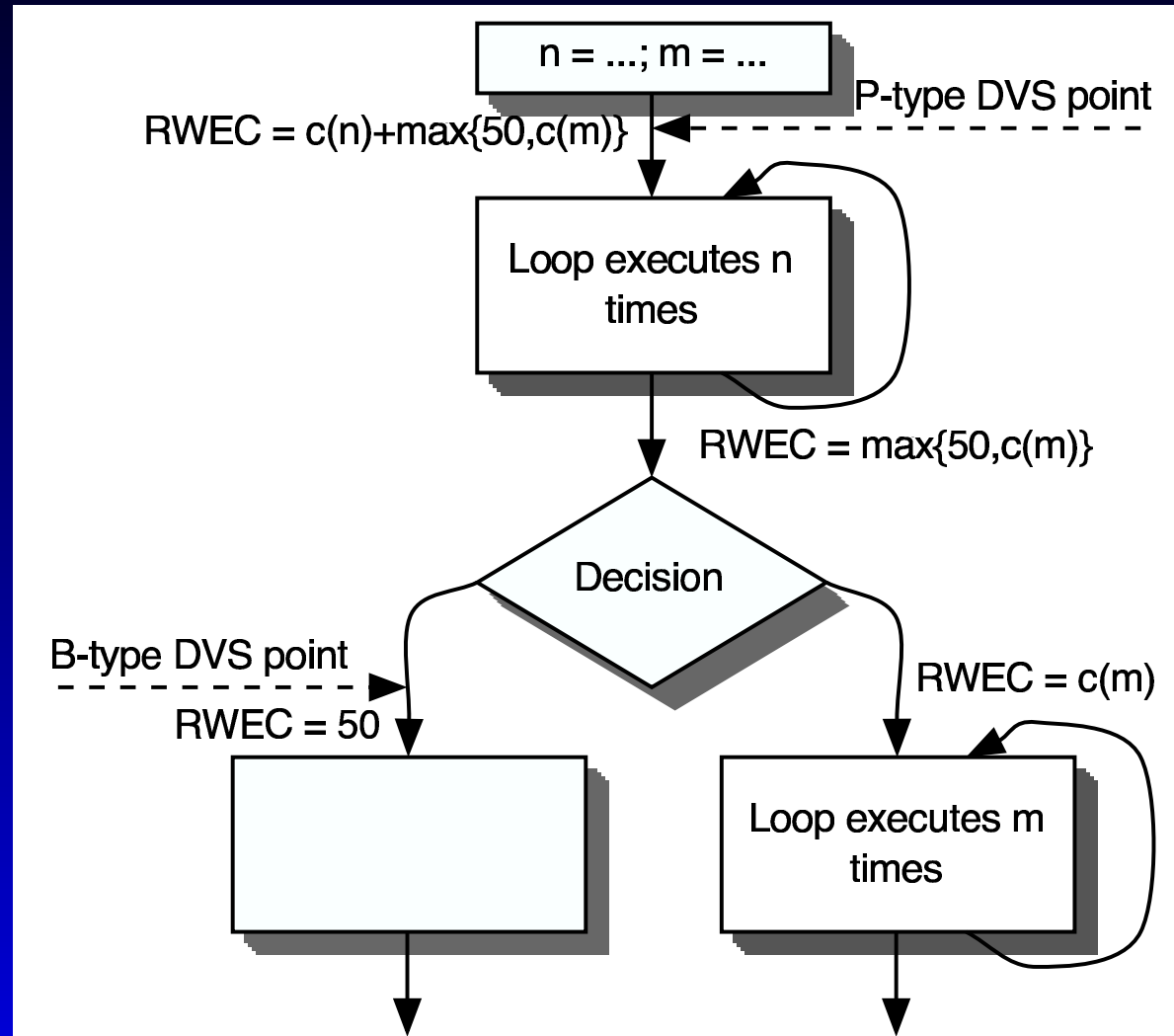
RVE/KAG/BW [COLP'03]

B and P-Type DVS



B/P-Type DVS
RVE/KAG/BW [COLP'03]

B and P-Type DVS



B/P-Type DVS
RVE/KAG/BW[COLP'03]

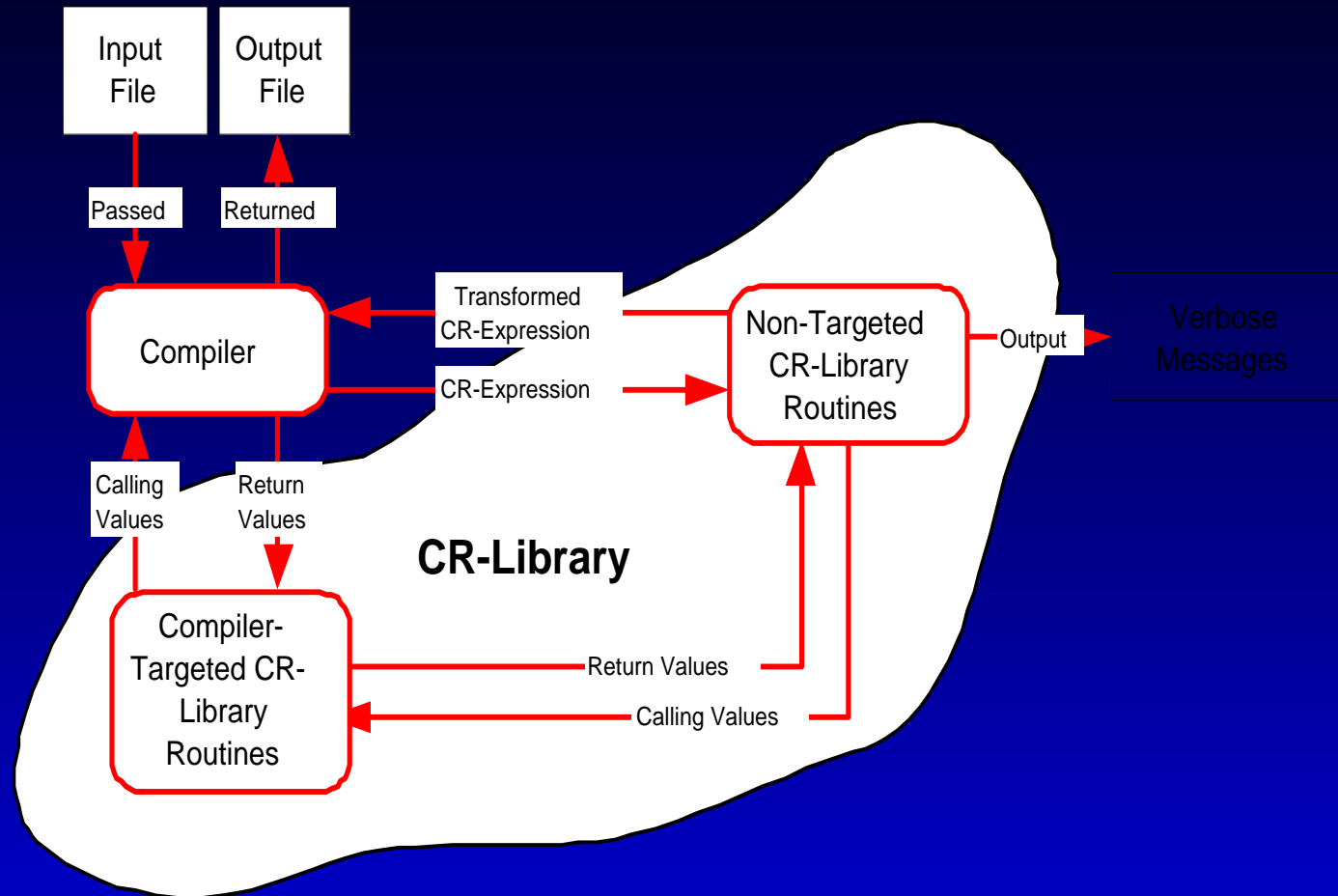
Parametric Time Estimation

- Initial work has been done to apply CR-based weighted summation to timing estimation
- RVE/KAG [IWIA'02] and RVE/KAG/Walsh [CPC'03]
- Based on Newton-Gregory polynomial form and CRs
- currently being evaluated for DSP architectures

Weighted Summation

NEED ONE OF THESE

Unified Framework



Summary

Unified framework for:

- Generalized induction variable analysis
- Pointer arithmetic and array conversion
- Value range and array range analysis
- Conditional GIVs and value range analysis
- Affine and non-linear dependence analysis
- <http://www.csit.fsu.edu/~birch/research/crDemo3.php>
- Parametric execution time estimation of loops
- Energy and power analysis of loop nests

Future Work

- nonlinear GCD for polynomial CRs
- Fourier-Motzkin-like Elimination via CR analysis
- detailed capability and efficiency comparison with extant compilers
- high-level language and RTL infrastructure for CR-based framework
- improved accuracy for time and power estimation
- locality analysis and transformation – data layout, blocking
- CR-based scheduling

Separator

THE END

GCD

Goal: Determines the existence of an integer solution to the dependence equation. If the GCD of the left hand side equations evenly divides the right side, dependence is possible.

Positives:

- Efficient, simple to implement
- Helps form general solutions

Negatives:

- The GCD is often 1
- Does not handle coupled subscripts
- Does not consider bounding constraints
- Restricted to affine equations

GCD

Consider the example:

$S_1 \rightarrow$ for ($i = 0; i < 10; i = i + 2$) {

$S_2 \rightarrow$ $A[i] = \dots$

$S_3 \rightarrow$ $\dots = A[i + 1]$

$S_4 \rightarrow$ }

Here the dependence equation for S_2 and S_3 is $2i^d - 2i^u = 1$. The GCD of the coefficients of i^d and i^u is 2, which does not evenly divide 1, hence no dependence is possible.

Generalized GCD

Goal: Looks for simultaneous solution to a system of dependence equation, to determine if dependence is possible. The test drives A , in $Ax = b$, to an echelon matrix that can be solved. If an integer solutions exist, dependence is possible.

Positives:

- Finds simultaneous solution to subscripts
- Often produces under-determined parameters which may be used to consider other constraints

Negatives:

- Does not consider bounding constraints
- Restricted to affine equations

Generalized GCD

Consider the example:

```
S1 → for (i = 0; i < 10; i = i + 1) {  
S2 →     for (j = 0; j < 10; j = j + 1) {  
S3 →         A[i + j + 1][j + 1] = ...  
S4 →         ... = A[i + j][j]  
S5 →     }  
S6 → }
```

Where the dependence equations:

$$i_d - i_u + j_d - j_u = -1$$

$$j_d - j_u = -1$$

is written in matrix form $\mathbf{Ax} = \mathbf{b}$.

Generalized GCD

Continuing:

Applying only elementary row operations we drive the augmented matrix

$$[\mathbf{U}|\mathbf{A}^T] = \left[\begin{array}{cccc|cc} 1 & 0 & 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 & -1 & 0 \\ 0 & 0 & 1 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 & -1 & -1 \end{array} \right]$$

to upper echelon form:

$$[\mathbf{U}|\mathbf{D}^T] = \left[\begin{array}{cccc|cc} 0 & 0 & 0 & -1 & -1 & -1 \\ 0 & 1 & 0 & -1 & 0 & -1 \\ 1 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 1 & 0 & 0 \end{array} \right] .$$

Generalized GCD

Continuing:

Now that we have our echelon \mathbf{D}^T the algorithm tries to solve for integer vector \mathbf{t} in $\mathbf{D}\mathbf{t} = \mathbf{b}$, written out:

$$\begin{bmatrix} -1 & 0 & 0 & 0 \\ -1 & -1 & 0 & 0 \end{bmatrix} \begin{bmatrix} t_1 \\ t_2 \\ t_3 \\ t_4 \end{bmatrix} = \begin{bmatrix} -1 \\ -1 \end{bmatrix}.$$

, to produce $t_1 = 1$ and $t_2 = 0$, with t_3 and t_4 free variables. Since t is composed of only integers the generalized GCD test demonstrates that there is a simultaneous integer solution to the system, the test returns there is dependence possible.

Banerjee Bounds

Goal: Determines the existence of a real solution to the dependence equation. If the right side constant of the equation lies between the upper bound and lower bound of the left side, dependence is possible.

Positives:

- Efficient, simple to implement
- Determines direction vector hierarchy
- Considers bounding constraints

Negatives:

- Determines real solutions exist, not integer solutions
- Restricted to affine equations

Banerjee Bounds

Consider the example:

$S_1 \rightarrow$ for ($i = 0; i < 10; i = i + 1$) {

$S_2 \rightarrow$ $A[2i] = B[i + 6]$

$S_3 \rightarrow$ $D[i] = A[3i - 1]$

$S_4 \rightarrow$ }

With dependence equation $2i^d - 3i^u = -1$ we compute the upper and lower bounds

Lower Bound	=	$2i^d - 3i^u$	Upper Bound	=	$2i^d - 3i^u$
	=	$2(0) - 3(10)$		=	$2(10) - 3(0)$
	=	-30		=	20

Since -1 lies in the range of [-30,20], dependence is possible.

I-Test

Goal: Extends the Banerjee test. Interval equations are produced, and terms are eliminated, updating the interval. Each step, a modified GCD test is performed. If all terms are eliminated or the Banerjee test on remaining terms does not fail, dependence is possible.

Positives:

- Efficient, simple to implement
- Can provide benefits of both GCD and Banerjee

Negatives:

- Determines real solutions exist, not integer
- Does not handle coupled subscripts
- Restricted to affine equations

I-Test

Consider the example:

```
S1 →   for (i = 0; i < 10; i = i + 1) {  
S2 →           for (j = 0; j < 10; j = j + 1) {  
S3 →                   A[i] = ...  
S4 →                   ... = A[4j + 8]  
S5 →           }  
S6 →   }
```

Dependence equation $i^d - 4j^u - 8 = 0$ is converted to interval form $i^d - 4j^u - 8 = [0, 0]$. At each step of elimination, determine if $\lfloor L/d \rfloor \leq \lfloor U/d \rfloor$ where d is the GCD of the coefficients a_1, a_2, \dots, a_n , and $[L, U]$ is the interval. Terms are incorporated into the interval if $|a_n| \leq U - L + 1$.

I-Test

Continuing:

$$\text{Interval} = i^d - 4j^u - 8 = [0, 0]$$

$$= i^d - 4j^u = [8, 8] \quad \text{rid constant}$$

$$= -4j^u = [-1, 8] \quad \text{rid } i^d \text{ since } |1| \leq 0 - 0 + 1$$

$$= 0 = [-1, 44] \quad \text{rid } j^u \text{ since } |-4| \leq 8 - (-1) + 1$$

Since 0 lies in the range of [-1,44], an integer solution is possible, and hence, dependence is possible.

Fourier Motzkin

Goal: A linear programming method applied to dependence analysis. Test systematically eliminates variables from a system of inequalities until all but a single variable has been eliminated. If no contradictions occur, dependence is possible.

Positives:

- Determines direction vector hierarchy
- Considers bounding constraints

Negatives:

- Determines real solutions exist, not integer
- Can be expensive, not practical.
- Restricted to affine equations

Omega Test

Goal: An integer programming method based on Fourier Motzkin. Variables projected produce shadows based on integer solutions.

Positives:

- Produces fewest inexact returns.
- Determines integer solution.

Negatives:

- Can be expensive, not practical
- Restricted to affine equations

Fourier Motzkin vs. Omega Test

Step 1: Rewrite constraints

$$c_1 \rightarrow 3x + 4y \geq 16 \quad x \geq 16/3 - 4y/3$$

$$c_2 \rightarrow 4x + 7y \leq 56 \quad x \leq 14 - 7y/4$$

$$c_3 \rightarrow 4x + 7y \leq 20 \quad x \leq 5 + 7y/4$$

$$c_4 \rightarrow 2x = 3y \geq -9 \quad x \geq -9/2 + 3y/2$$

Step 2: Eliminate x

Fourier Motzkin

$$c_{1,3} \rightarrow 5 + 7y/4 \geq 16/3 - 4y/3$$

$$c_{3,4} \rightarrow 5 + 7y/4 \geq 16/3 - 9/2 + 3y/2$$

$$c_{1,2} \rightarrow 14 - 7y/4 \geq 16/3 - 4y/3$$

$$c_{2,4} \rightarrow 14 - 7y/4 \geq -9/2 + 3y/2$$

Omega

$$5 + 7y/4 \geq 16/3 - 4y/3 + 1$$

$$5 + 7y/4 \geq -9/2 + 3y/2 + 1$$

$$14 - 7y/4 \geq 16/3 - 4y/3 + 1$$

$$14 - 7y/4 \geq -9/2 + 3y/2 + 1$$

Step 3: Test for contradictions

$$4/37 \leq y \leq 74/13 \quad 16/37 \leq y \leq 66/13$$

Both test show dependence is possible.