Then the slots of the word sequence are successively filled:

- We start with the minimal nodes in the dependence graph.
- If we fail to fill all slots of a word, we insert : :-)
- After every inserted instruction, we re-compute \( \Sigma \).

**Warning:**

→ The execution of two VLIWs can overlap !!!
→ Determining an optimal sequence, is NP-hard ...

The UD-edge \( (3, 4) \) has been inserted to exclude that \( z \) is over-written before use :-)

In the next step, each instruction is annotated with its (required reources, in particular, its) execution time.

Our goal is a maximally parallel correct sequence of words.

For that, we maintain the current system state:

\[
\Sigma : \text{Vars} \rightarrow \mathbb{N}
\]

\[
\Sigma(x) = \text{expected delay until } x \text{ is available}
\]

Initially:

\[
\Sigma(x) = 0
\]

As an invariant, we guarantee on entry of the basic block, that all operations are terminated :-)

Example: Word width \( k = 2 \)

<table>
<thead>
<tr>
<th>Word</th>
<th>State</th>
</tr>
</thead>
<tbody>
<tr>
<td>( x )</td>
<td>1 0 0 0</td>
</tr>
<tr>
<td>( y )</td>
<td>0 1 0 0</td>
</tr>
<tr>
<td>( z )</td>
<td>0 0 1 0</td>
</tr>
<tr>
<td>( t )</td>
<td>0 0 0 0</td>
</tr>
</tbody>
</table>

In each cycle, the execution of a new word is triggered.
The state just records the number of cycles still to be waited for the result :-)

663

664
Example: \( k = 2 \)

<table>
<thead>
<tr>
<th>Word</th>
<th>State</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>x</td>
</tr>
<tr>
<td>2</td>
<td>y</td>
</tr>
<tr>
<td></td>
<td>z</td>
</tr>
<tr>
<td></td>
<td>t</td>
</tr>
<tr>
<td>( x = x + 1 )</td>
<td>( y = M[A] )</td>
</tr>
<tr>
<td>( t = z )</td>
<td>( z = M[A + x] )</td>
</tr>
</tbody>
</table>

In each cycle, the execution of a new word is triggered.
The state just records the number of cycles still to be waited for the result. :-)

Note:

- If instructions put constraints on future selection, we also record these in \( \Sigma \).
- Overall, we still distinguish just finitely many system states. :-)
- The computation of the effect of a VLIW onto \( \Sigma \) can be compiled into a finite automaton. !!!
- This automaton, though, could be quite huge. :-(
- The challenge of making choices still remains. :-(
- Basic blocks usually are not very large
  \[ \implies \text{opportunities for parallelization are limited.} \] :-((

Extension 1: Acyclic Code

```c
if (x > 1) {
    y = M[A];
    z = x - 1;
} else {
    y = M[A + 1];
    z = x - 1;
}
y = y + 1;
```

The dependence graph must be enriched with extra control-dependencies ...

The statement \( z = x - 1; \) is executed with the same arguments in both branches and does not modify any of the remaining variables. :-)

We could have moved it before the \( \text{if} \) anyway. :-}
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We could have moved it before the if anyway :-))

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The following code could be generated:

| z = x - 1 | if \(!((x > 0))\) goto A |
| y = M[A]; | goto B |

\begin{align*}
A : & y = M[A + 1] \\
B : & y = y + 1 \\
\end{align*}

At every jump target, we guarantee the invariant :-(
If we allow several (known) states on entry of a sub-block, we can generate code which complies with all of these.

... in the Example:

\[
\begin{array}{ll}
z = x - 1 & \text{if } (! (x > 0)) \text{ goto } A \\
y = M[A] & \text{goto } B \\
A : & y = M[A + 1] \\
B : & y = y + 1
\end{array}
\]

If this parallelism is not yet sufficient, we could try to speculatively execute possibly useful tasks ...

For that, we require:

- an idea which alternative is executed more frequently;
- the wrong execution may not end in a catastrophe, i.e., run-time errors such as, e.g., division by 0;
- the wrong execution must allow roll-back (e.g., by delaying a commit) or may not have any observational effects ...

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... in the Example:

| $z = x - 1$ | $y = M[A]$ | if $(x > 0)$ goto $B$ |
| $y = M[A + 1]$ |

$B$:

| $y = y + 1$ |

In the case $x \leq 0$ we have $y = M[A]$ executed in advance. This value, however, is overwritten in the next step :-)

In general:

$x = e$; has no observable effect in a branch if $x$ is dead in this branch :-(

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<th>if (x &gt; 0) goto B</th>
</tr>
</thead>
<tbody>
<tr>
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<td></td>
<td></td>
</tr>
<tr>
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**Extension 2:** Unrolling of Loops

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Now it is clear which side of tests to prefer:
the side which stays within the unrolled body of the loop.

Warning:

- The different instances of the body are translated relative to possibly different initial states.
- The code behind the loop must be correct relative to the exit state corresponding to every jump out of the loop.

Example:

(for \( x = 0; x < n; x++ \))
\[ M[A + x] = z; \]

Duplication of the body yields:
for \( (x = 0; x < n; x++) \) {
    \( M[A + x] = z; \)
    \( x = x + 1; \)
    if \(!((x < n))\) break;
    \( M[A + x] = z; \)
}
Discussion:

- Elimination of the intermediate test together with the the fusion of all increments at the end reveals that the different loop iterations are in fact independent.
- Nonetheless, we do not gain much since we only allow one store per word.
- If right-hand sides, however, are more complex, we can interleave their evaluation with the stores.

In order to fuse two loops into one, we require that:

- the iteration schemes coincide;
- the two loops access different data.

In case of individual variables, this can easily be verified. This is more difficult in presence of arrays. Taking the source program into account, accesses to distinct statically allocated arrays can be identified.

An analysis of accesses to the same array is significantly more difficult...

---

Extension 3:

Sometimes, one loop alone does not provide enough opportunities for parallelization.

... but perhaps two successively in a row.

Example:

```
for (x = 0; x < n; x++) {
    R = B[x];
    S = C[x];
    T1 = R + S;
    A[x] = T1;
}
```

```java
for (x = 0; x < n; x++) {
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**Example:**

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for (x = 0; x < n; x++) {    for (x = 0; x < n; x++) {
    R = B[x];                    R = B[x];
    S = C[x];                    S = C[x];
    T1 = R + S;                  T2 = R - S;
    A[x] = T1;                   C[x] = T2;
}                                }
```

The first loop may in iteration $x$ not read data which the second loop writes to in iterations $< x$.

The second loop may in iteration $x$ not read data which the first loop writes to in iterations $> x$.

If the index expressions of jointly accessed arrays are linear, the given constraints can be verified through integer linear programming ...

```
\begin{align*}
  i & \geq 0 \\
  i & \leq x - 1 \\
  x_{\text{write}} &= i \\
  x_{\text{read}} &= x \\
  x_{\text{read}} &= x_{\text{write}} \\
\end{align*}
```

$: x_{\text{read}}$ read access to C by 1st loop
$: x_{\text{write}}$ write access to C by 2nd loop

... obviously has no solution ☺
The first loop may in iteration \( x \) not read data which the second loop writes to in iterations \( < x \).

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If the index expressions of jointly accessed arrays are linear, the given constraints can be verified through integer linear programming ...

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\begin{align*}
\lambda_{\text{write}} \cdot y & \geq 0 \\
\lambda_{\text{read}} \cdot y & \leq x - 1 \\
x_{\text{read}} & = x \\
x_{\text{write}} & = x
\end{align*}
\]

\# \( x_{\text{read}} \) read access to \( C \) by 1st loop
\# \( x_{\text{write}} \) write access to \( C \) by 2nd loop

... obviously has no solution \( \vdots \)

\( x \leq x - 1 \)

---

**Simple Case:**

The two inequations have no solution over \( \mathbb{Q} \).

Then they also have no solution over \( \mathbb{Z} \) \( \vdots \)

... in Our Example:

\[
\begin{align*}
x & = i \\
0 & \leq i = x \\
0 & \leq x - 1 - i = -1
\end{align*}
\]

The second inequation has no solution \( \vdots \)

---

2. \( x = S \)

**General Form:**

\[
\begin{align*}
s & \geq t_1 \\
t_2 & \geq s \\
y_t & = s_1 \\
y_i & = s_2 \\
y_l & = y_2
\end{align*}
\]

for linear expressions \( s, t_1, t_2, s_1, s_2 \) over \( i \) and the iteration variables.

This can be simplified to:

\[
0 \leq s - t_1 \quad 0 \leq t_2 - s \quad 0 = s_1 - s_2
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What should we do with it ???

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2 \( x = S \)

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The two inequations have no solution over $\mathbb{Q}$.
The then they also have no solution over $\mathbb{Z}$ :-)

... in Our Example:

$$x = i$$
$$0 \leq i$$
$$0 \leq x - 1 - i = -1$$

The second inequation has no solution :-)

One Variable:
The inequations where $x$ occurs positive, provide lower bounds.
The inequations where $x$ occurs negative, provide upper bounds.
If $G, L$ are the greatest lower and the least upper bound, respectively,
then all (integer) solution are in the interval $[G, L]$ :-)

Example:

$$0 \leq 13 - 7 \cdot x \iff x \leq \frac{13}{7}$$
$$0 \leq -1 + 5 \cdot x \iff x \geq \frac{1}{5}$$

The only integer solution of the system is $x = 1$ :-)

$$\frac{4}{5} \leq x \leq \frac{7}{3}$$

Discussion:

- Solutions only matter within the bounds to the iteration variables.
- Every integer solution there provides a conflict.
- Fusion of loops is possible if no conflicts occur :-)
- The given special case suffices to solve the case one variable over $\mathbb{Z}$ :-)
- The number of variables in the inequations corresponds to the nesting-depth of for-loops $\implies$ in general, is quite small :-)

Integer Linear Programming (ILP) can decide satisfiability of a
finite set of equations/inequations over $\mathbb{Z}$ of the form:

$$\sum_{i=1}^{n} a_i \cdot x_i = b \quad \text{bzw.} \quad \sum_{i=1}^{n} a_i \cdot x_i \geq b, \quad a_i \in \mathbb{Z}$$

Moreover, a (linear) cost function can be optimized :-)

Warning: The decision problem is in general, already NP-hard !!!
Notwithstanding that, surprisingly efficient implementations exist.
Not just loop fusion, but also other re-organizations of loops yield
ILP problems ...