

# IBM ILOG CPLEX Optimization Studio OPL Language Reference Manual

Version 12 Release 8



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## **Contents**

Figures v	Limitations on constraints
_	Formal parameters
Tables vii	Basic formal parameters
	Tuples of parameters 61
Languago Poforonos Manuel iv	Filtering in tuples of parameters 62
Language Reference Manual ix	Scheduling
	Introduction
Chapter 1. Why an Optimization	Piecewise linear and stepwise functions 64
Programming Language? 1	Interval variables 66
	Unary constraints on interval variables 69
Chapter 2. OPL, the modeling language 3	Precedence constraints between interval variables 70
Building a model	Constraints on groups of interval variables 70
Data types	A logical constraint between interval variables:
Basic data types	presenceOf
Data structures	Expressions on interval variables
Data sources	Sequencing of interval variables
Data initialization	Cumulative functions
Spreadsheet Input/Output	State functions
Data consistency 27	Notation
Data consistency	
Pitfall of lazy initialization of data	Chapter 3. IBM ILOG Script for OPL 87
Decision types	Language structure
Decision variables	Syntax
Expressions of decision variables	Expressions in IBM ILOG Script
Dynamic collection of elements into arrays 34	Statements
Expressions	Built-in values and functions
Usage of expressions	Numbers
Data and decision variable identifiers	IBM ILOG Script strings 106
Integer and float expressions	IBM ILOG Script Booleans
Aggregate operators	IBM ILOG Script arrays
Piecewise-linear functions	Objects
Discontinuous piecewise-linear functions 40	Dates
Set expressions	The null value
Boolean expressions 45	The undefined value
Constraints	IBM ILOG Script functions
Introduction	Miscellaneous functions
Using constraints 46	
Constraint labels 48	Index
Types of constraints	

# **Figures**

1.	A Piecewise-linear function	3.	Discontinuous costs.					. 43
2.	The discontinuous piecewise-linear function	4.	Labeled assert					. 51
	sign()							

## **Tables**

1.	Escape sequences inside literal strings 5	23.	Escape sequences in strings
2.	Operations allowed on sets	24.	Examples of string literals using escape
3.	Initializing sets in the model file		sequences
4.	A discontinuous cost function 42	25.	String property
5.	Functions over sets 44	26.	String methods
6.	Reference syntax of script variables 89	27.	String functions
7.	Property access syntax 90	28.	String operators
8.	Assignment operator syntax 90	29.	Boolean method
9.	Syntactic shorthand 91	30.	Logical operators
10.	Function call syntax 91	31.	Array constructor
11.	Special keyword syntax 92	32.	Array properties
12.	Special operator syntax	33.	Array methods
13.	Other operator syntax	34.	Object constructor
14.	Conditional statement syntax 95	35.	User-defined constructors
15.	Loop syntax	36.	Built-in method
16.	Declaration syntax for script variables 97	37.	Date constructor
<i>17</i> .	Function definition syntax	38.	Date methods
18.	Default value syntax	39.	Date functions
19.	Number method	40.	Methods of null
20.	Numeric functions	41.	Methods of undefined
21.	Numeric constants	42.	Function methods
22	Numeric operators 104	43	Miscellaneous functions

# **Language Reference Manual**

This manual provides reference information about IBM® ILOG® Optimization Programming Language (OPL), the modeling language used in CPLEX® Studio. For details of prerequisites, naming conventions, and other general information, see *How to use the documentation* in the Starting Kit.

## **Chapter 1. Why an Optimization Programming Language?**

OPL is a modeling language for combinatorial optimization that aims at simplifying the solving of optimization problems.

Linear programming, integer programming, and combinatorial optimization problems arise in a variety of application areas, which include planning, scheduling, sequencing, resource allocation, design, and configuration.

In this context, OPL is a modeling language for combinatorial optimization that aims at simplifying the solving of these optimization problems. As such, it provides support in the form of computer equivalents for modeling linear, quadratic, and integer programs, and provides access to state-of-the-art algorithms for linear programming, mathematical integer programming, and quadratic programming.

Within the IBM ILOG OPL product, OPL as a modeling language has been redesigned to better accommodate IBM ILOG Script, its associated script language.

## Chapter 2. OPL, the modeling language

Presents the modeling language of IBM ILOG OPL, namely: the overall structure of OPL models; the basic modeling concepts; how data can be initialized internally as it is declared or externally in a .dat file; how to connect to, read from, and write to databases and spreadsheets; expressions and relations; constraints; and formal parameters.

## **Building a model**

Describes the primary elements of the OPL language and how they are used to build an optimization model.

The basic building blocks of OPL are integers, floating-point numbers, identifiers, strings, and the keywords of the language. Identifiers in OPL start with a letter or the underscore character ( \_ ) and can contain only letters, digits, and the underscore character. Note that letters in OPL are case-sensitive. Integers are sequences of digits, possibly prefixed by a minus sign. Floats can be described in decimal notation (3.4 or -2.5) or in scientific notation (3.5e-3 or -3.4e10).

The OPL reserved words are listed in OPL keywords.

Comments in OPL are written in between /\* and \*/ as in:

```
/*
This is a
multiline comment
*/
```

The characters // also start a comment that terminates at the end of the line on which they occur as in:

```
dvar int cost in 0..maxCost; // decision variable
```

An OPL model consists of:

- a sequence of declarations
- optional preprocessing instructions
- the model/problem definition
- optional postprocessing instructions
- optional flow control (main block)

Details about these elements are given in the chapters that follow.

A simple model (volsay.mod)

A typical model is shown here.

```
dvar float+ Gas;
dvar float+ Chloride;

maximize
   40 * Gas + 50 * Chloride;
subject to {
   ctMaxTotal:
     Gas + Chloride <= 50;</pre>
```

```
ctMaxTotal2:
   3 * Gas + 4 * Chloride <= 180;
ctMaxChloride:
   Chloride <= 40;</pre>
```

In this example, there is only the declarative initial part and the model definition. There is no preprocessing, postprocessing, or flow control.

### **Data types**

Describes basic data types and data structures available for modeling data in OPL.

### Basic data types

Describes integers, floats, strings, piecewise linear functions, and stepwise functions in OPL.

#### Integers

Describes integers (int) in OPL.

Shows how to declare integers in the OPL language.

OPL contains the integer constant maxint, which represents the largest positive integer available. OPL provides the subset of the integers ranging from -maxint to maxint as a basic data type.

A declaration of the form

```
int i = 25;
```

declares an integer, i, whose value is 25.

The initialization of an integer can be specified by an expression. For instance, the declaration

```
int n = 3;
int size = n*n;
```

initializes size as the square of n. Expressions are covered in detail in "Expressions" on page 36.

#### **Floats**

Describes floats (float) in OPL.

Shows how to declare floats in the OPL language.

OPL also provides a subset of the real numbers as the basic data type float. The implementation of floats in OPL obeys the IEEE 754 standard for floating-point arithmetic and the data type float in OPL uses double-precision floating-point numbers. OPL also has a predefined float constant, <u>infinity</u>, to represent the IEEE infinity symbol. Declarations of floats are similar to declarations of integers.

```
The declaration
```

```
float f = 3.2;
```

declares a float f whose value is 3.2.

The value of the float can be specified by an arbitrary expression.

#### **Strings**

Describes strings (string) in OPL.

Shows how to declare strings in the OPL language.

defines and initializes <u>a set of strings</u>. Strings can appear in tuples and index arrays.

Strings are manipulated in the <u>preprocessing phase</u> via ILOG Script. See Chapter 3, "IBM ILOG Script for OPL," on page 87 for details on the scripting language.

The OPL parser supports the following escape sequences inside literal strings:

Table 1. Escape sequences inside literal strings

Escape sequence	Meaning
<b>\</b> b	backspace
\t	tab
\n	newline
\f	form feed
\r	carriage return
\"	double quote
\\	backslash
\000	octal character ooo
\xXX	hexadecimal character XX

To continue a literal string <u>over several lines</u>, you need to <u>escape the new line</u> <u>character</u>:

"first line \

second line"

#### **Piecewise linear functions**

Describes piecewise linear functions (pwlFunction) in OPL.

Shows how to declare piecewise linear functions in the OPL language.

Piecewise linear functions are typically used to <u>model a known function of time</u>, for instance the cost incurred for completing an activity after a known date.

Note that you must ensure that the array of values T[i] is sorted.

The meanings of the S, T, and V vectors are described in "Piecewise linear and stepwise functions" on page 64, in the *Language Reference Manual*.

Syntax

```
pwlFunction F = piecewise(i in 1..n){ S[i]->T[i]; S[n+1] } (t0, v0);

pwlFunction F = piecewise{ V[1]->T[1], ..., V[n]->T[n], V[n+1] };

pwlFunction F[i in ...] = piecewise (...)[ ... ];

Example
int n=2;
float objectiveforxequals0=300;
float breakpoint[1..n]=[100,200];
float slope[1..n+1]=[1,2,-3];

int points[1..n+1] = [5, 150, 400];
float values[x in 1..n+1] = piecewise(i in 1..n) {slope[i] -> breakpoint[i]; slope[n+1]}(0,0) points

execute {
    writeln( values );
}

dvar int x;

maximize piecewise(i in 1..n) {slope[i] -> breakpoint[i]; slope[n+1]}(0,objectiveforxequals0) x;
subject to {true;}
```

Piecewise linear functions are covered in detail in "Piecewise linear and stepwise functions" on page 64.

#### Stepwise functions

Describes stepwise functions (stepFunction) in OPL.

Shows how to declare stepwise functions in the OPL language.

Stepwise linear functions are typically used to model the efficiency of a resource over time. A stepwise function is a special case of piecewise linear function where all slopes are equal to 0 and the domain and image of F are integer.

Note that you must ensure that the array of values <code>T[i]</code> is sorted.

```
Syntax
```

```
stepFunction F = stepwise(i in 1..n){ V[i] \rightarrow T[i]; V[n+1] }; stepFunction F = stepwise{ V[1] \rightarrow T[1], ..., V[n] \rightarrow T[n], V[n+1] }; stepFunction F[i in ...] = stepwise (...)[ ... ];
```

#### Example

A declaration of the form stepFunction f=stepwise {0->3; 2}; assert f(-1)==0; assert f(3)==2; assert f(3.1)==2;

declares a stepwise function, f.

#### Example

Another example, declaring the stepwise function F2:

```
stepFunction F2 = stepwise{ 0->0; 100->20; 60->30; 100 };
int ii= F2( 10 );
execute {
          writeln( ii );
          writeln( F2( 25 ) );
}
```

Stepwise functions are covered in detail in "Piecewise linear and stepwise functions" on page 64.

#### **Data structures**

Describes how the basic data types can be combined using arrays, tuples, and sets to obtain complex data structures.

#### Ranges

Integer ranges are fundamental in OPL, since they are often used in arrays and variable declarations, as well as in aggregate operators, queries, and quantifiers.

#### **Declaring ranges**

To specify an integer range, you give its lower and upper bounds, as in range Rows = 1..10;

which declares the range value 1..10. The lower and upper bounds can also be given by expressions as in

```
int n = 8;
range Rows = n+1..2*n+1;
```

Once a range has been defined, you can use it as an array indexer.

Whenever a range is empty, i.e. its upper bound is less than its lower bound, it is automatically normalized to 0..-1 (in other words, all empty ranges are equal).

#### The range declaration

An integer range is typically used:

 as an array index in an array initialization expression range R = 1..100;

```
int A[R]; // A is an array of 100 integers
```

as an iteration range

```
range R = 1..100;
forall(i in R) {
   //element of a loop
   ...
}
```

 as the domain of an integer decision variable dvar int i in R;

#### The range float declaration

A range float data type consists of a couple of float values specifying an interval. It is typically used as the domain of a floating-point decision variable.

```
range float X = 1.0..100.0;
dvar float x in X;
```

#### **Arrays**

Describes one-dimensional arrays and multidimensional arrays.

Arrays are fundamental in many applications.

#### One-dimensional arrays

One-dimensional arrays are the simplest arrays in OPL and vary according to the type of their elements and index sets. A declaration of the form

```
int a[1..4] = [10, 20, 30, 40];
```

declares an array of four integers a[1],...,a[4] whose values are 10, 20, 30, and 40. It is of course possible to define arrays of other basic types. For instance, the instructions

```
int a[1..4] = [10, 20, 30, 40]; float f[1..4] = [1.2, 2.3, 3.4, 4.5]; string d[1..2] = ["Monday", "Wednesday"];
```

declare arrays of natural numbers, floats, and strings, respectively.

The index sets of arrays in OPL are very general and can be integer ranges and arbitrary finite sets. In the examples so far, index sets were given explicitly, but it is possible to use a previously defined range, as in

```
range R = 1..4;
int a[R] = [10, 20, 30, 40];
```

The declaration:

```
int a[Days] = [10, 20, 30, 40, 50, 60, 70];
```

describes an array indexed by a set of strings; its elements are a ["Monday"],...,a ["Sunday"].

Arrays can also be indexed by finite sets of arbitrary types. This feature is fundamental in OPL to exploit sparsity in large linear programming applications, as discussed in detail in Exploiting sparsity in the *Language User's Manual*.

For example, the declaration:

```
tuple Edges {
    int orig;
    int dest;
}
{Edge} Edges = {<1,2>, <1,4>, <1,5>};
int a[Edges] = [10,20,30];
```

defines an integer array, a, indexed by a finite set of tuples. Its elements are a[<1,2>], a[<1,4>], and a[<1,5>]. Tuples are described in detail in "Tuples" on page 9.

#### Multidimensional arrays

OPL supports the declaration of multidimensional arrays (see "Data initialization" on page 15 about the ellipsis syntax). For example, the declaration:

```
int a[1..2][1..3] = ...;
```

declares a two-dimensional array, a, indexed by two integer ranges. Indexed sets of different types can of course be combined, as in

```
int a[Days][1...3] = ...;
```

which is a two-dimensional array whose elements are of the form a [Monday] [1]. It is interesting to contrast multidimensional and one-dimensional arrays of tuples.

Consider the declaration:

```
{string} Warehouses = ...;
{string} Customers = ...;
int transp[Warehouses, Customers] = ...;
```

that declares a two-dimensional array transp. This array may represent the units shipped from a warehouse to a customer. In large-scale applications, it is likely that a given warehouse delivers only to a subset of the customers. The array transp is thus likely to be sparse, i.e. it will contain many zero values.

The sparsity can be exploited by declarations of the form:

This declaration specifies a set, routes, that contains only the relevant pairs (warehouse, customer). The array transp can then be indexed by this set, exploiting the sparsity present in the application. It should be clear that, for large-scale applications, this approach leads to substantial reductions in memory consumption.

You can initialize arrays by listing its values, as in most of the examples presented so far. See "Initializing arrays" on page 17, "As generic arrays" on page 18, and "As generic indexed arrays" on page 18 for details.

#### **Tuples**

Data structures in OPL can be constructed using tuples that cluster together closely related data. This topic describes how to declare tuples, use keys on tuples, and initialize tuples.

#### **Declaring tuples**

For example, the declaration:

```
tuple Point {
        int x;
        int y;
    };
Point point[i in 1..3] = <i, i+1>;
```

declares a tuple Point consisting of two fields x and y of type integer. Once a tuple type T has been declared, tuples, arrays of tuples, sets of tuples of type T, tuples of tuples can be declared, as in:

```
Point p = <2,3>;
Point point[i in 1..3] = <i, i+1>;
{Point} points = {<1,2>, <2,3>};
```

```
tuple Rectangle {
    Point 11;
    Point ur;
}
```

These declarations respectively declare a point, an array of three points, a set of two points, and a tuple type where the fields are points. The various fields of a tuple can be accessed in the traditional way by suffixing the tuple name with a dot and the field name, as in

```
Point p = <2,3>;
int x = p.x;
```

which initializes x to the field x of tuple p. Note that the field names are local to the scope of the tuples.

#### Note:

Multidimensional arrays are not supported in tuples.

#### Keys in tuple declarations

As in database systems, tuple structures can be associated with *keys*. Tuple keys enable you to access data organized in tuples using a set of unique identifiers. In the following example, the nurse tuple is declared with the key name of type string.

Declaring a tuple using a single key (nurses.mod)

```
tuple nurse {
   key string name;
   int seniority;
   int qualification;
   int payRate;
}
```

Then, supposing Isabelle must not work more than 20 hours a week, just write: NurseWorkTime[<"Isabelle">]<=20;

leaving out the fields with no keys. This is equivalent to:

```
NurseWorkTime[<"Isabelle",3,1,16>]<=20;</pre>
```

Using keys in tuple declarations has practical consequences, in particular:

- The key field can be used as a unique identifier for the tuple, for example the field name in Declaring a tuple using a single key (nurses.mod). In this example, it means that there will be no two tuples with the same name in a set of tuples of the type nurse. If a user inadvertently attempts to add two different tuples with the same name, OPL raises an error.
- Defining keys enables you to access elements of the tuple set by using only the value of the key field (name in the nurses example). Slicing is one of the features that benefit from it: you can slice on the tuple set using only key fields.

You can also declare a tuple using a non singleton set of keys, such as the shift tuple of the nurses example below.

```
Declaring a tuple using a set of keys (nurses.mod)
tuple shift {
   key string departmentName;
   key string day;
```

```
key int startTime;
key int endTime;
int minRequirement;
int maxRequirement;
}
```

In this example, a shift is uniquely identified by the department name, the date, and start and end times, all defined as key fields.

#### Initializing tuples

You initialize tuples by giving the list of the values of the various fields, as in: Point p = <2,3>;

which initializes p.x to 2 and p.y to 3. See "Initializing tuples" on page 19 for details.

#### Limitations on tuples

When using tuples in your models, you should be aware of various limitations.

#### Data types in tuples

Not all data types are allowed inside tuples. The limitations are given here.

Data types allowed in tuples

- Primitives (int, float, string)
- Tuples (also known as subtuples)
- Arrays with primitive items (not string), that is: integer or float arrays
- Sets with primitive items, that is: integer, float or string sets

#### Note:

Arrays and sets in tuples are not compared by content. If collections inside a tuple are modified, duplicates are not detected.

Data types not allowed in tuples

- Sets of tuples (instances of IloTupleSet)
- Arrays of strings, tuples, and tuple sets
- Multidimensional arrays

#### Tuple indices and tuple patterns

You cannot mix tuple indexes and patterns within the declaration and the use of decision expressions. For example, these code lines raise the following error message Data not consistent for "xxx": can not mix pattern and index between declaration of dexpr and instantiation.

Do not mix tuple indices and tuple patterns in dexpr

```
dexpr float y[i in t] = ...;
subject to {
   forall(<a,b,c> in t) y[<a,b,c>]==...; };
dexpr float y[<a,b,c> in t] = ...;
subject to {
   forall(i in t) y[i]==...;
};
```

#### Performance and memory consumption

If you choose to label constraints in large models, use tuple indices instead of tuple patterns to avoid increasing the performance and memory cost. See "Constraint labels" on page 48.

#### Sets

Gives a definition of sets, a list of the operations allowed on sets, and a few words on their initialization.

#### **Definition**

Sets are non-indexed collections of elements without duplicates.

OPL supports sets of arbitrary types to model data in applications. If T is a type, then {T}, or alternatively setof(T), denotes the type "set of T". For example, the declaration:

```
{int} setInt = ...;
setof(Precedence) precedences = ...;
```

declares a set of integers and a set of precedences.

Sets may be ordered, sorted, or reversed. By default, sets are ordered, which means that:

- Their elements are considered in the order in which they have been created.
- Functions and operations applied to ordered sets preserve the order.

See "Sorted and ordered sets" on page 13 for details.

#### Operations on sets

The following operations are allowed on sets. See OPL functions in *Language Quick Reference* for more information about functions. For the functions on sets, the index starts at 0.

Table 2. Operations allowed on sets

Operations	Syntax				
union, inter, diff, symdiff	set = set1 function set2				
first, last	elt = function(set)				
next, prev	elt = function(set,elt,int)				
nextc, prevc	elt = function(set,elt,int)				
item	elt = function(set,int)				
ord	int = function(set,elt)				

#### Initializing sets

A set can be initialized in various ways. The simplest way is by listing its values explicitly. For example:

```
tuple Precedence {
    int before;
    int after;
}
{Precedence} precedences = {<1,2>, <1,3>, <3,4>};
```

various functions in Cplex See "Initializing sets" on page 21 for details.

#### Methods for sets

For advanced users:

The methods for sets can be found in the class IloDiscreteDataCollection, in the *ILOG Script Reference Manual*.

#### Sorted and ordered sets

Shows the effect of the sorted property on simple sets, tuple sets, and sets used in piecewise linear functions.

#### **Definitions**

An ordered set is a set whose elements are arranged in the order in which they
were added to the set. Note that this is how sets are created by default. For
example:

```
{int} S1 = {3,2,5};
and
ordered {int} S1 = {3,2,5};
are equivalent.
```

• A sorted set is a set in which elements are arranged in their natural, ascending (or descending) order. For strings, the natural order is the lexicographic order. The natural order also depends on the system locale. For example:

```
sorted {int} sortedS = {3,2,5};
and
ordered {int} orderedS = {2,3,5};
```

are equivalent, and iterating over sortedS or orderedS will have the same behavior.

To specify the descending order, you add the keyword reversed.

#### Simple sets

The following code sample enables you to see the difference between the union of ordered sets and the union of sorted sets.

```
{int} s1 = {3,5,1};
{int} s2 = {4,2};
{int} orderedS = s1 union s2;
sorted {int} sortedS = s1 union s2;
execute{
    writeln("ordered union = ", orderedS);
    writeln("sorted union = ", sortedS);
}

The statement
{int} orderedS = s1 union s2;
returns
ordered union = {3 5 1 4 2}

while the statement
sorted {int} sortedS = s1 union s2;
returns
```

```
sorted union = \{1 \ 2 \ 3 \ 4 \ 5\}
```

#### Sorted tuple sets

When a tuple set does not use keys, the entire tuple, except set and array fields, is taken into account for sorting. For tuple sets with keys, sorting takes place on all keys in their order of declaration. In other words, it is not possible to sort a tuple set on one (or more) given column(s) only.

The following code extract declares a team of people who are defined by their first name, last name, and nickname, then prints the list of team members first in the creation order, then in alphabetical order.

```
tuple person {
      string firstname;
      string lastname;
      string nickname;
tuple personKeys {
      key string firstname;
      key string lastname;
      string nickname;
{person} devTeam = {
<"David", "Atkinson", "Dave">, <"David", "Doe", "Skinner">,
<"Gregory", "Simons", "Greg">,
<"David", "Smith", "Lewis">,
<"Kevin", "Morgan", "Kev">,
<"Gregory", "McNamara ", "Mac">
sorted {personKeys} sortedDevTeam = {<i,j,k> | <i,j,k> in devTeam};
execute{
         writeln(devTeam);
         writeln(sortedDevTeam);
The person tuple uses no keys.
tuple person {
      string firstname;
      string lastname;
      string nickname;
}
The personKeys tuple uses keys for the first and last names, not for the nickname.
tuple personKeys {
      key string firstname;
      key string lastname;
      string nickname;
}
```

The data shows that the team includes three people whose first name is David, two people whose first name is Gregory, and one person whose first name is Kevin.

```
As a consequence, the statement sorted {personKeys} sortedDevTeam = {<i,j,k> | <i,j,k> in devTeam};
```

lists the David tuples before the Gregory tuples, which themselves appear before the Kevin tuple. Within the David tuples, "David" "Doe" "Skinner" comes before "David" "Smith" "Lewis" because a second sorting also takes place on the second

field with the key lastname. In contrast, since there is no person with the same first name and last name, no sort is ever done on the last field nickname.

The output of sortedDevTeam is displayed in the CPLEX Studio IDE as:

#### Sorted sets in piecewise linear functions

In piecewise linear functions, breakpoints must be strictly increasing. However, in most cases, the data supplied by a database or a .dat file is not sorted in an increasing numeric or lexicographic order. As a consequence, you have to add complex and verbose scripting statements to sort the data.

To avoid these extra code lines, the sorted property of sets enables you to sort data by specifying a single keyword, as shown in the following code extract. Writing piecewise linear functions becomes easier, as one code line is sufficient instead of several dozens.

```
tuple Cost{
    key int BreakPoint;
    float Slope;
}
sorted {Cost} sS = { <1, 1.5>, <0, 2.5>, <3, 4.5>, <2, 4.5>};
float lastSlope = 3.5;

dvar float+ x;
minimize piecewise(t in sS)
    {t.Slope -> t.BreakPoint; lastSlope} x;
```

See also "Piecewise-linear functions" on page 39.

Other related topics:

"Data sources"

Describes data initialization, spreadsheet input/output, data consistency, and preprocessing.

Describes data initialization.

Introduction to scripting

Defines IBM ILOG Script as a scripting language and describes the situations in which it is used: preprocessing, postprocessing, and flow control. Also provides programming tips and warns you of pitfalls to avoid.

Describes how to set declarations.

#### **Data sources**

Describes data initialization, spreadsheet input/output, data consistency, and preprocessing.

#### Data initialization

Defines internal versus external initialization, describes how to initialize arrays, tuples, and sets, and discusses memory allocation aspects of data initialization.

#### Internal and external initialization

Defines these two kinds of data initialization.

In OPL, you can initialize data internally or externally. Your choice affects memory allocation. See "Initialization and memory allocation" on page 23 for details.

#### Internally

This initialization mode consists in initializing the data in the model file at the same time as it is declared. Inline initializations can contain expressions to initialize data items, such as

```
int a[1..5] = [b+1, b+2, b+3, b+4, b+5];
```

#### Note:

If you choose to initialize data within a model file, you will get an error message if you try to access it later by means of a scripting statement such as:

```
myData.myArray inMod[1] = 2;
```

#### **Externally**

This initialization mode consists in specifying initialization subsequently as an OPL statement in a separate .dat file (see OPL Syntax in the *Language Quick Reference*). This includes reading from a spreadsheet as explained in "Spreadsheet Input/Output" on page 24.

You declare external data using the ellipsis syntax. However, data initialization instructions cannot contain expressions, since they are intended to specify data. Data initialization instructions make it possible to specify sets of tuples in very compact ways. Consider these types in a .mod file:

```
{string} Product = {"flour", "wheat", "sugar"};
{string} City = {"Providence", "Boston", "Mansfield"};
tuple Ship {
string orig;
string dest;
string p;
}
{Ship} shipData = ...;
```

and assume that the set of shipments is initialized externally in a separate .dat file like this:

#### Note:

In .dat files, the separating comma is optional. For strings without any special characters, even the enclosing quotes are optional.

#### Initializing arrays

Describes the various ways in which you can initialize arrays.

#### You can initialize arrays:

- · "Externally"
- "Internally" on page 18
- "In preprocessing instructions" on page 18
- "As generic arrays" on page 18
- "As generic indexed arrays" on page 18

#### **Externally**

Arrays can be initialized by external data, in which case the declaration has the form:

```
int a[1..2] [1..3] = ...;
```

and the actual initialization is given in a data source, which is a separate .dat file in IBM ILOG OPL.

#### • By listing values

This is how arrays are initialized in most of the examples presented so far. Multidimensional arrays in OPL are, in fact, arrays of arrays and must be initialized accordingly. For example, the declaration:

```
/* .mod file */
int a[1..2][1..3] = ...;
/* .dat file */
a = [
[10, 20, 30],
[40, 50, 60]
]:
```

initializes a two-dimensional array by giving initializations for the one-dimensional arrays of its first dimension. It is easy to see how to generalize this initialization to any number of dimensions.

#### • By specifying pairs

An array can also be initialized by specifying pairs (index, value), as in the declaration:

```
/* .mod file */
int a[Days] = ...;
/* .dat file */
a = #[
"Monday": 1,
"Tuesday": 2,
"Wednesday": 3,
"Thursday": 4,
"Friday": 5,
"Saturday": 6,
"Sunday": 7
]#;
```

#### Note:

- 1. When the initialization is specified by (index, value) pairs, the delimiters #[ and ]# must be used instead of [ and ].
- 2. The ordering of the pairs can be arbitrary.

These two forms of initialization can be combined arbitrarily, as in:

```
/* .mod file */
int a[1..2][1..3] = ...;
/* .dat file */
a = #[
2: [40, 50, 60],
1: [10, 20, 30]
]#;
```

#### Internally

You can initialize arrays internally (that is, in the .mod file) using the same syntax as in .dat files. Here, the array items may be expressions that are evaluated during initialization. The syntax for pairs #[, ]# is not available for internal initialization.

#### In preprocessing instructions

Arrays can also be initialized in the preprocessing instructions, as in:

```
range R = 1..8;
int a[R];
execute {
  for(var i in R) {
    a[i] = i + 1;
  }}
```

which initializes the array in such a way that a[1] = 2, a[2] = 3, and so on. See "Preprocessing of data" on page 30.

#### As generic arrays

OPL also supports generic arrays, that is, arrays whose items are initialized by an expression. These generic arrays may significantly simplify the modeling of an application. The declaration:

```
int a[i in 1..10] = i+1;
```

declares an array of 10 elements such that the value of a[i] is i+1. Generic arrays can of course be multidimensional, as in:

```
int m[i in 0..10][j in 0..10] = 10*i + j;
```

which initializes element m[i][j] to 10\*i + j. Generic arrays are useful in performing some simple transformations. For instance, generic arrays can be used to transpose matrices in a simple way, as in:

```
int m[Dim1][Dim2] = ...;
int t[j in Dim2][i in Dim1] = m[i][j];
```

More generally speaking, generic arrays can be used to permute the indices of arrays in simple ways.

#### As generic indexed arrays

To have more flexibility when initializing arrays in a generic way, OPL enables you to control the **index** value in addition to the **item** value, as described earlier in *As generic arrays*. To illustrate the syntax, the same examples can be expressed as follows:

```
int a[1..10] = [ i-1 : i | <mark>i in 2..11</mark> ];
int m[0..10][0..10] = [ i : [ j : <mark>10*i+j</mark> ] | <mark>i,j in 0..10</mark> ];
```

indexed pairs 列表解析

This syntax is close to the syntax used for initializing arrays in .dat files by means of indices, delimited by #[ and ] #, as explained above in Specifying pairs. Using this syntax is an efficient means of initializing arrays used to index data.

The oilDB.mod example contains an execute block that performs initialization. Instead of:

```
GasType gas[Gasolines];
execute {
  for(var g in gasData) {
   gas[g.name] = g
}
```

the same can be expressed with the syntax for generic indexed arrays as:

```
GasType gas[Gasolines] = [ g.name : g | g in gasData ];
```

```
Likewise, this syntax, from transp4.mod
float Cost[Routes];
execute INITIALIZE {
  for( var t in TableRoutes ) {
      Cost[Routes.get(t.p,Connections.get(t.o,t.d))] = t.cost;
}
```

is equivalent to

```
float Cost[Routes] = [ <t.p,<t.o,t.d>>:t.cost | t in TableRoutes ];
```

#### Note:

- 1. It is recommended to use generic arrays or generic indexed arrays whenever possible, since they make the model more explicit and readable.
- 2. If an index is met more than once, no warning is issued and the latest value set for this index is the one kept.

```
For example:
int n=5;
{int} s = \{1,3,4,2,5\};
sorted {int} s2=asSet(1..n);;
reversed {int} s3=asSet(1..n);;
int x[1..n] = [max](n-i,i): i | i in s];
int x2[1..n]=[maxl(n-i,i): i | i in s2];
int x3[1..n]=[maxl(n-i,i): i | i in s3];
execute
 writeln(x);
 writeln(x2);
 writeln(x3);
gives out
[0 0 2 4 5]
[0 0 3 4 5]
[0 0 2 1 5]
```

#### Initializing tuples

Describes the two ways of initializing tuples.

You initialize tuples either by giving the list of the values of the various fields (see "Tuples" on page 9) or by listing the fields and values. For example:

```
In the .mod file, you write:
tuple point
{
  int x;
  int y;
}
point p1=...;
point p2=...;
In the .dat file, you write:
p1=#<y:1,x:2>#;
p2=<2,1>;
```

As with arrays, the delimiters < and > are replaced by #< and ># and the ordering of the pairs is not important. OPL checks whether all fields are initialized exactly once.

The type of the fields can be arbitrary and the fields can contain arrays and sets.

#### **Example 1: tuple Rectangle**

For example, the following code lines declare a tuple with three fields: the first is an integer and the other two are arrays of two points.

```
tuple Rectangle {
        int id;
        int x[1..2];
        int y[1..2];
}

Rectangle r = ...;

execute
{
    writeln(r);
}
```

A specific "rectangle" can be declared in the data file as: r=<1, [0,10], [0,10];

#### **Example 2: tuple Precedence**

```
The declaration
tuple Precedence {
    string name;
    {string} after;
}
```

defines a tuple in which the first field is a set item and the second field is a set of values. A possible precedence can be declared as follows:

```
Precedence p = <a1, {a2, a3, a4, a5}>; assuming that a1,..,a5 are strings.
```

You can also initialize tuples internally within the .mod file. If you choose to do so, you cannot use the named tuple component syntax #<, >#, which is supported in .dat files but **not** in .mod files. Components may be expressions and will be evaluated during initialization.

#### Initializing sets

You can initialize sets in three different ways.

- "Externally"
- "Internally"
- "As generic sets" on page 22

#### Externally

As stated in "Initializing sets" on page 12, in the Sets topic, the simplest way to initialize a set is by listing its values explicitly in the .dat file.

For example, the declaration:

```
/* .mod file */
tuple Precedence {
 int before;
 int after;
{Precedence} precedences = ...;
/* .dat file */
precedences = {<1,2>, <1,3>, <3,4>};
```

initializes a set of tuples.

#### Internally

You can also initialize sets internally (in the .mod file), more precisely by using set expressions using previously defined sets and operations such as union, intersection, difference, and symmetric difference. The symmetric difference of two sets A and B is

```
(A \cup B) \setminus (A \cap B)
```

described in "Expressions" on page 36.

For example, the declarations:

```
{int} s1 = \{1,2,3\};
{int} s2 = \{1,4,5\};
{int} i = s1 inter s2;
\{int\} j = \{1,4,8,10\} inter s2;
\{int\}\ u = s1\ union\ \{5,7,9\};
{int} d = s1 diff s2;
\{int\}\ sd = s1\ symdiff \{1,4,5\};
```

initialize i to  $\{1\}$ , u to  $\{1,2,3,5,7,9\}$ , d to  $\{2,3\}$ , and sd to  $\{2,3,4,5\}$ .

It is also possible to initialize a set from a range expression. For example, the declaration:

```
\{int\}\ s = asSet(1..10)
initializes s to \{1,2,...,10\}
```

It is important to point out at this point that sets initialized by ranges are represented explicitly (unlike ranges). As a consequence, a declaration of the form {int} s = asSet(1..100000);

creates a set where all the values 1, 2, ..., 100000 are explicitly represented, while the range

```
range s = 1..100000;
```

represents only the bounds explicitly.

Note that when writing the assignment s2=s1, you add one element to s1, that element is also added to s2. If you do not want this, write

```
s1=\{i \mid i \text{ in } s2\}
```

For example, compare the statements in the following table.

Table 3. Initializing sets in the model file

If you write	the result is
{int} s1={1,2};	{1 2 3}
{int} s2=s1;	
execute	
{	
s2.add(3);	
writeln(s1);	
}	
{int} s1={1,2};	{1 2}
{int} s2={ i   i in s1};	
//{int} s2=s1;	
execute	
{	
s2.add(3);	
writeln(s1);	
}	

#### As generic sets

OPL supports generic sets which have an expressive power similar to relational database queries. For example, the declaration:

```
{int} s = \{i \mid i \text{ in } 1...10: i \text{ mod } 3 == 1\};
```

initializes s with the set {1,4,7,10}. A generic set is a conjunction of expressions of the form

```
p in S : condition
```

where p is a parameter (or a tuple of parameters), S is a range or a finite set, and condition is a Boolean expression. These expressions are also used in forall statements and aggregate operators and are discussed in detail in "Formal parameters" on page 59.

The declaration:

地址共享,是引用

```
首先知道数据结构,然后
学会在适当的时间地点应
用这些数据结构,进而对
程序进行优化
```

is a more interesting example, showing a conjunction of expressions, and is explained in detail in "Formal parameters" on page 59. Generic sets are often useful when you transform a data structure (e.g. the data stored in a file) into a data structure more appropriate for stating the model effectively. Consider, for example, the declarations:

```
{string} Nodes ...;
int edges[Nodes] [Nodes] = ...;
```

which describe the edges of a graph in terms of a Boolean adjacency matrix. It may be important for the model to use a sparse representation of the edges (because, for instance, edges are used to index an array). The declaration:

```
tuple Edge {
   Nodes o;
   Nodes d;
}
{Edge} setEdges = {<o,d> | o,d in Nodes : edges[o][d]==1};
```

computes this sparse representation using a simple generic set. It is of course possible to define generic arrays of sets. For example, the declaration:

```
{int} a[i in 3..4] = {e | e in 1..10: e mod i == 0};
```

initializes a[3] to  $\{3,6,9\}$  and a[4] to  $\{4,8\}$ .

#### Initialization and memory allocation

Describes how memory is allocated to data initialization.

In OPL, the initialization mode you choose affects memory allocation. Namely, external initialization from a .dat file, while enabling a more modular design, may have a significant impact on memory usage.

#### Internal initialization

Internal data (directly from the model file) is initialized when first used. This is also called "lazy initialization". Unused internal data elements are not allocated any memory. In other words, internal data is "pulled" from OPL as needed.

Example of lazy initialization

```
int a=2;
int b=2;
int a2=2*a;
int b2=2*b;
execute
{
    a2;
    a++;
    b++;
```

```
writeln(a2);
writeln(b2);
assert a2==4;
assert b2==6;
```

#### **External initialization**

In contrast, data from a data file is initialized while the .dat file is parsed and is allocated memory whether it is used by the model or not. In other words, external data is "pushed" to OPL.

See also "Pitfall of lazy initialization of data" on page 31.

## Spreadsheet Input/Output

Describes how to connect an MS Excel spreadsheet, read from it, and write the results to the connected spreadsheet.

#### The oilsheet example

Explains spreadsheet input and output in the context of an oil spreadsheet.

This section uses the oil Sheet example to demonstrate operations with an MS Excel spreadsheet. You can find this example in

```
<Install dir>/opl/examples/opl/oil
```

where <Install dir> is your installation directory.

#### Connection to a spreadsheet

Explains how to connect OPL to a spreadsheet.

The spreadsheet operations in OPL all refer to a spreadsheet connection. The instruction

```
/* .dat file */
SheetConnection sheet("transport.xls");
```

establishes a connection sheet to a spreadsheet named transport.xls. The connection sheet should be viewed as a handle on the spreadsheet. It is possible in OPL to connect to several spreadsheets within the same model.

The instruction SheetConnection takes only one parameter and you do not need to specify the full path to the spreadsheet name. Relative paths are resolved using the current directory of .dat files.

Note that, in this section, we often use the word "spreadsheet" for "spreadsheet connection".

#### Reading from a spreadsheet

Explains how to read from a spreadsheet from within OPL.

In OPL, spreadsheet ranges can be read into one- or two-dimensional arrays or sets. For instance, the instructions:

```
/* .mod file */
{string} Gasolines = ...;
tuple GasType {
```

```
float demand;
float price;
float octane;
float lead;
}

GasType gas[Gasolines] = ...;
/* .dat file */
SheetConnection sheet("oilSheet.xls");

Gasolines from SheetRead(sheet, "'gas data'!A2:A4");
Gas from SheetRead(sheet, "'gas data'!B2:E4");
```

### What data can be read from an Excel spreadsheet?

OPL opens a spreadsheet in read-only mode to read data from it.

The types of data elements supported are:

- · sets with integers, floats, strings, or tuples;
- · scalar integers, floats, or strings;
- arrays with integers, floats, one- or two-dimensional strings, or one-dimensional tuples;
- one- or two-dimensional arrays of simple types: for such arrays, the data must be formatted, that is, it must have the same width/length as the array to be filled. OPL automatically determines whether the data must be read line by line or column by column. When facing a square zone (a two-dimensional array with [x] [x] as dimensions), the engine reads the data line by line.

#### Note:

- Only tuples with integer, float, and string components are supported.
- If the sheet name contains a space, the name should be surrounded by single quotes in the SheetRead instruction. For example:

```
Oil from SheetRead(sheet, "'oil data'!B2:E4");
```

• OPL does not support the R1C1 reference style to specify the range when reading data from an Excel spreadsheet.

### Accessing named ranges in Excel

IBM ILOG OPL supports the convention of *names*, which are a word or string of characters used to represent a cell, range of cells, formula, or constant value, and that can be used in other formulas.

Thus you can use easy-to-understand names, such as Nutrients, to refer to hard-to-understand ranges, such as B4:J15 or IncreasedProtein to refer to a constraint. You can then substitute these names in formulas for the range of cells or constraint.

Excel *names*, or named ranges, can be accessed using the SheetRead command, using the following syntax:

```
SheetConnection sheetData("C:\\ILOG_Files\\myExcelFile.xls");
prods from SheetRead(sheetData, "Product");
```

Note the double separator  $\setminus \setminus$  in the SheetConnection command.

The SheetRead command is normal, and in this example the Excel name Product replaces the normal syntax of, say, C13:072.

To create named ranges in Excel 2007:

- 1. Highlight the range of cells you want to name, then click the **Name** box at the left end of the Formula Bar.
- 2. Type the name you want to assign to this range and press Enter.
- 3. Save the spreadsheet file.

**Note:** For a list of supported versions of Microsoft Excel, please see the detailed system requirements http://www-01.ibm.com/support/docview.wss?uid=swg27019100.

Additional information on named ranges

- Excel automatically updates (expands) a named range when a row is added somewhere within the range. However, one must careful adding rows at the end of a range as the range does not get automatically updated in that case. It would have to be updated manually.
- OPL allows blank rows in a named range. If you are reading a set of strings, it will consider the blank cells as having the value 0. If you are reading a set of strings, then it inserts an empty string "" into the set. For example:

```
s2 = {"Monday" "" "Wednesday" "Thursday" "Friday"}
```

This behavior is the same when you don't use named range but instead use explicit ranges like C1:C5, where C2 is empty.

- With the Excel VBA one can name the first (top left) cell of a named range and access the whole range. OPL does not support this.
- When using sheetWrite to write to named ranges, the size of the range does not have to match the size of the data you are writing to Excel. If the set is smaller, then only the top most cells will be filled.

If you try to write more data than the range can accommodate, you receive the error message: "Exception from IBM ILOG Concert: excel: range is not wide enough to write the set".

In this sense, named ranges behave in exactly the same way as "regular" ranges.

### Format of the Excel data

Here we must differentiate between simple types and tuples:

- Sets of simple types: The engine reads data from left to right and top to bottom. Data can therefore be read either horizontally, vertically, or from a rectangular zone.
- Sets of tuples: The data has to be formatted because the tuple schema has an arity. As in databases and manual tables, the data format is "fixed width, variable length". Therefore, tuple sets are read only line by line in Excel: this is the same representation as in pure data files.

## Writing to a spreadsheet

Explains how to write to a spreadsheet from within OPL.

This section uses extracts from the oilSheet.dat data file.

Publishing results to a spreadsheet can be performed using instructions such as: a to SheetWrite(sheet, "RESULT!A2:A4"); blend to SheetWrite(sheet, "RESULT!B2:D4");

OPL then opens spreadsheets in read-write mode. This action may fail if another process is already using the .xls file.

The types of data elements supported for writing are just the same as for reading. Cells in Microsoft Excel spreadsheets are filled from left to right and from top to bottom.

Excel names (or named ranges) can be accessed using the SheetWrite command, using the following syntax:

```
SheetConnection sheetData("C:\\ILOG_Files\\myExcelFile.xls");
prods to SheetWrite(sheetData, "Product");
```

Note the double separator \\ in the SheetConnection command. The SheetWrite command is normal, and, in this example, the Excel name Product replaces the normal syntax of, say, C13:072.

#### Note:

• OPL does not support the R1C1 reference style to specify the range when writing data to an Excel spreadsheet.

For more information on named ranges, see "Accessing named ranges in Excel" on page 25.

## Data consistency

Defines the purpose of data consistency and describes data membership and assertions as ways to ensure consistency.

## **Purpose**

Provides an overview of data consistency issues in OPL.

For an optimization problem to give relevant solutions, it is fundamental that you provide good quality data to your projects. In particular, it may be interesting to check that the data is consistent. If the project data is not consistent, the solving engine may find a wrong solution, or no solution, and you may think that the model is erroneous and therefore waste time trying to improve it.

OPL offers several ways to check the consistency of the data used by your projects.

### In particular:

• "Data membership consistency": use the keyword with to ensure that cells of a given tuple in a tuple set correctly belong to a given set of possible values.

#### Note:

You can also use the keyword key for data consistency when declaring tuples. See "Keys in tuple declarations" on page 10.

• "Assertions" on page 29: use the keyword assert to ensure that some assertions on the data are respected.

### Data membership consistency

Explains the use of the with keyword to ensure data consistency.

The keyword with enables you to indicate that a given element of a tuple must be contained in a given set. If you use it, OPL checks the consistency of the tuple set at run time when initializing the set. The syntax is:

```
{tupletype} tupleset with cell1 in set1, cell2 in set = ...;
```

Let's take an example. You have a set of arcs between nodes. Nodes are defined by a tuple set of tuples consisting of an origin node and a destination node. The with syntax enables you to ensure that the origin and destination nodes do belong to a given set of nodes. Compare "Data found inconsistent (keyword with)" and "Data found consistent (keyword with)":

### Data found inconsistent (keyword with)

```
{int} nodes = {1, 5, 7};

tuple arc {
  int origin;
  int destination;
}

{arc} arcs2 with origin in nodes, destination in nodes =
  {<1,4>, <5,7>};

execute {
  writeln(arcs2);
};
```

### Data found consistent (keyword with)

```
{int} nodes = {1, 5, 7};

tuple arc {
  int origin;
  int destination;
}

{arc} arcs1 with origin in nodes, destination in nodes =
  {<1,5>, <5,7>};

execute {
  writeln(arcs1);
};
```

If you write "Data found inconsistent (keyword with)," an error will be raised when the set arcs2 is initialized because the with syntax will detect that the statement

```
(int) nodes = (1, 5, 7);
is not consistent with the statement
```

```
with origin in nodes, destination in nodes = \{<1,4>, <5,7>\}
```

If you write "Data found consistent (keyword with)," the initialization of the set arcs1 will work properly because the with syntax will find that the statement (int) nodes = (1, 5, 7);

```
is consistent with the statement
with origin in nodes, destination in nodes =
  {<1,5>, <5,7>}
```

## Initializing tuple sets referring to other sets

To initialize tuple sets that refer to other sets with keys for data consistency, you must use initialization expressions that provide only those key values, as shown in

"Initializing tuple sets referring to other sets" on page 28. This is true if you initialize those tuple sets as internal data or as external data in .dat files, databases, or spreadsheets.

Initializing tuple sets referring to other sets

tuple node
{
 key int node\_id;
 string city;
 string country;
}

{node} nodes = {<1,"Paris","France">,<5,"Madrid","Spain">, <7,"New York","USA">};

tuple arc {
 node origin;
 node destination;
}

{arc} arcs1 with origin in nodes, destination in nodes=...;

execute {
 writeln(arcs1);

#### **Assertions**

Explains the use of assertions with regard to data consistency.

OPL provides assertions to verify the consistency of the model data. This functionality enables you to avoid wrong results due to incorrect input data. In their simplest form, assertions are simply Boolean expressions that must be true; otherwise, they raise an execution error. For instance, it is common in some transportation problems to require that the demand matches the supply. The declaration

```
int demand[Customers] = ...;
int supply[Suppliers] = ...;
assert sum(s in Suppliers) supply[s] == sum(c in Customers) demand[c];
```

makes sure that the total supply by the suppliers meets the total demand from the customers. This assertion can be generalized to the case of multiple products, as in

```
int demand[Customers] [Products] = ...;
int supply[Suppliers] [Products] = ...;
assert
  forall(p in Products)
      sum(s in Suppliers) supply[s][p] == sum(c in Customers) demand[c][p];
```

This assertion verifies that the total supply meets the total demand for each product. The use of assertions is highly recommended, since they make it possible to detect errors in the data input early, avoiding tedious inspection of the model data and results.

Assertions can be labeled. See "Labeled assert statements" on page 50.

### Note:

In OPL, elements are initialized when they are first used. This is called 'lazy initialization' (see "Data initialization" on page 15).

However, the execution of an assert statement triggers an evaluation of the elements involved in the assertion. So, if you use an OPL element for the first time in a skipped assert expression, the lazy initialization will be triggered later. Skipping or removing assert statements may impact the order in which OPL elements are initialized and thus change the behavior of a program.

## Preprocessing of data

Provides an overview of preprocessing operations in OPL.

You can preprocess data before the optimization model is created by using IBM ILOG Script/JavaScript syntax encapsulated in execute blocks.

OPL provides script integration with the modeling language. All declared model elements are available for scripting via their name.

The functionality available for an element depends on its type. All elements can be read, but modifications are possible only for primitive types (int, float, string) and primitive items of arrays and tuples. See the to the *Reference Manual of IBM ILOG Script Extensions for OPL* about these limitations.

You can change the domain boundaries for decision variables, as well as their priority, in the preprocessing phase.

You can also use preprocessing to change CPLEX or CP Optimizer parameter settings (see Changing option values in the *Language User's Manual*).

Elements of a range or constraint type are immutable.

```
Example:
int n = ...;
range R = 1..n;
int A[R] = ...

execute {
   for(r in R) {
       if ( A[r] < 0 ) {
            A[r] = 0;
       }
   }
}</pre>
```

## Initialization and processing order

Preprocessing items are processed according to their category, not in absolute declaration order.

Within categories, the order is:

- External data, in the order in which the data sources are added to the OPL model
- 2. All execute blocks and assert statements, in declaration order

Internal data is initialized as encountered during external data initialization, in execute blocks and in assert statements. The following examples use internal data.

If you write:

```
{int} s1={1,2};
{int} s2={ i | i in s1};
execute
writeln(s2);
s1.add(3);
writeln(s1,s2);
the result is:
{1 2}
{1 2 3} {1 2}
whereas if you write:
{int} s1={1,2};
{int} s2={ i | i in s1};
execute
//writeln(s2);
s1.add(3);
writeln(s1,s2);
the result is:
{1 2 3} {1 2 3}
```

The order in which you use the internal data is important. The order in which you use external data is not important because it is all created, or initiated, at the beginning of the run.

See also "Pitfall of lazy initialization of data."

Use the profiler feature to inspect the instantiation sequence of your model. See Profiling the execution of a model in *IDE Tutorials*.

See Chapter 3, "IBM ILOG Script for OPL," on page 87 for details on the scripting language and its extensions for OPL.

## Lazy initialization

It is important to be aware from OPL 5.2 onwards that during processing, declared data elements are initialized on demand when referenced for the first time. See *Data preprocessing* in OPL Migration Guide for migration aspects.

# Pitfall of lazy initialization of data

In OPL, external data is always initialized before internal data.

If you are not aware of the lazy initialization of data in OPL, you may be surprised by unexpected results.

OPL is a declarative language. The order of elements as they are written in the source file is less important than with imperative languages. We initialize OPL data elements when they are first used, which may result in an initialization order different from the declaration order in a .mod file. External data (with =... syntax) is always initialized before internal data. Scripting blocks are executed after the external data.

## **Example**

In the following example, the array arrayInitExternally is initialized with three elements in the .dat file, but, because set1 has four elements, the result shows that arrayInitExternally has one more element (initialized by default with zero).

```
string arrayInitByScript[0..3];
execute initializeArray {
  for(var i=0; i<=3; i++) {
   if(i >= 1) arrayInitByScript[i] = "A";
    else arrayInitByScript[i] = "B";
// set1 is later used to index an array initialized externally.
// This external initialization of arrayInitExternally alters the content of set1
// because external initialization is done prior to script execution.
// So instead of having set1=\{1\ 2\ 3\}, set1 has an additional element "0"
// Before the script is executed, arrayInitByScript contains four empty elements ["" "" ""]
{int} set1 = union(s in 0..3 : arrayInitByScript[s] != "B") {s};
{int} set2 = union(s in 0..3 : arrayInitByScript[s] != "B") {s};
int arrayInitExternally[set1] = ...; // [0, 0, 0]; //
execute {
writeln(arrayInitByScript);
writeln(set1);
writeln(set2);
                                               initializing ordering
writeln(arrayInitExternally);
This produces the output:
Γ"B" "A" "A" "A"]
{0 1 2 3}
{1 2 3}
[0 0 0 0]
```

# **Decision types**

Variables in an OPL application are decision variables (dvar). OPL also supports decision expressions, that is, expressions that enable you to reuse decision variables (dexpr). A specific syntax is available in OPL to dynamically collect elements into arrays.

## **Decision variables**

Shows how to declare and use decision variables in the OPL language.

A decision variable is an unknown in an optimization problem. It has a domain, which is a compact representation of the set of all possible values for the variable. Decision variable types are *references* to objects whose exact nature depends on the underlying optimizer of a model. A decision variable can be instantiated only in the context of a given model instance.

The purpose of an OPL model is to find values for the decision variables such that all constraints are satisfied or, in optimization problems, to find values for the variables that satisfy all constraints and optimize a specific objective function. Variables in OPL are thus essentially decision variables and differ fundamentally from variables in programming languages such as Java, and ILOG Script.

### Note:

OPL decision variables are noted with the dvar keyword while the keyword var denotes ILOG Script variables.

A decision variable declaration in OPL specifies the type and set of possible values for the variable. Once again, decision variables can be of different types (integer, float) and it is possible to define multidimensional arrays of decision variables. The declaration

```
dvar int transp[Orig][Dest] in 0..100;
```

declares a two-dimensional array of integer variables. The decision variables are constrained to take their values in the range 0..100; i.e., any solution to the model containing this declaration must assign values between 0 and 100 to these variables. Note that all integer variables need a finite range in OPL. Arrays of decision variables can be constructed using the same index sets as arrays of data. In particular, it is also possible, and desirable for larger problems, to index arrays of decision variables by finite sets. For example, the excerpt:

```
tuple Route {
    City orig;
    City dest
}
{Route} routes = ...;
dvar int transp[routes] in 0..100;
```

declares an array of decision variables transp that is indexed by the finite set of tuples routes. Genericity can be used to initialize the domain of the variables. For example, the excerpt:

```
tuple Route {
         City orig;
         City dest;
}
{Route} routes = ...;
int capacity[routes] = ...;
dvar int transp[r in routes] in 0..capacity[r];
```

declares an array of decision variables indexed by the finite set routes such that variable transp[r] ranges over 0..capacity[r]. The array capacity is also indexed by the finite set routes. Note that decision variables can be declared to range over a user-defined range. For example, the excerpt:

```
range Capacity = 0..limitCapacity;
dvar int transp[Orig][Dest] in Capacity;
```

declares an array of integer variables ranging over Capacity.

Decision variables can of course be declared individually, as in: dvar int averageDelay in 0..maxDelay;

For convenience, OPL proposes the types float+, int+ and boolean to define the domain of a decision variable. The declarations

```
dvar int+ x; // non negative integer decision variable
dvar float+ y; // non-negative decision variable
dvar boolean z; // boolean decision variable
```

```
are therefore equivalent to
dvar int x in 0..maxint;
dvar float y in 0..infinity;
dvar int z in 0..1;
```

Decision variables in an array can be assigned item-specific ranges, as in
dvar float transp[o in Orig][d in Dest] in O..cap[o][d];

which declares a two-dimensional array of float variables, where variable transp[0][d] ranges over the set 0..cap[0][d].

## **Expressions of decision variables**

Shows how to declare and use decision variable expressions in the OPL language.

The keyword dexpr allows you to create reusable decision expressions. Indeed, if an expression has a particular meaning in terms of the original problem, writing it as a decision expression (dexpr) makes the model more readable.

For example, the scalableWarehouse.mod example expresses the total fixed costs as a decision expression:

```
dexpr int TotalFixedCost = sum( w in Warehouses ) Fixed * Open[w];
dexpr float TotalSupplyCost = sum( w in Warehouses, s in Stores ) SupplyCost[s][w] * Supply[s][w];
```

This way, the two total cost expressions defined are shown in the OPL IDE along with their values.

You can also use arrays of decision expressions. For example: dexpr int slack[i in r] = x[i] - y[i];

This array is handled efficiently as only the "definition" is kept. Not all the expressions for each value of the indices are created. As a consequence, you cannot change the definition of the dexpr for a particular element of the array.

Using decision expressions is particularly useful and recommended if you plan to write objectives to be used with ODM Enterprise. Please refer to the ODM Enterprise documentation.

# Dynamic collection of elements into arrays

Discusses the "all" syntax, explicit arrays, appending arrays, and dynamic initialization of decision variable arrays.

#### Introduction

Provides an overview of how elements are collected into arrays in OPL.

Some expressions (such as count) and constraints (such as allDifferent) need arrays of variables or constants to be created. In some models, these expressions or constraints can be used in an aggregate statement (for example, in a forall statement) and the exact content of the arrays depends on the iteration.

### The all syntax

Shows how to use the all syntax in the OPL language.

Then, the all syntax allows you to dynamically collect some decision variables or constants into an array. The syntax is similar to sum and forall, it contains a series of possible generators (an index and a set or a range in which this index is to be contained), some possible filters (to filter out some of the enumerated combinations), and a body (here of the form x[i][j]...). The variables or values in the resulting array follow the logical order of enumerating the index combinations as defined by the generators.

By default, this dynamic array is indexed from 0 to numberOfElements-1. As some constraints make a particular usage of the index, it may be interesting to define another indexing schema. For this, it is possible to dynamically define the range of the resulting array of variables by using the syntax [minindex..maxindex]. Finally, it is possible to use "\*" as maxIndex to indicate that only the minIndex is defined; the maxIndex will be set accordingly depending on the number of elements.

Here is a complete of usage of the syntax:

```
using CP;
int n = 5;
range R = 1..n;
dvar int x[R] in R;
subject to {
   allDifferent(all(i in R:i%2==1) x[i]);
}
```

This is a way of defining an array of variables or values, and in all the constraints and expressions that take arrays. You can use either this syntax or pass directly a named array. When you pass a named array and indexes make sense in the constraint, its indexer will be used to index the elements if it has one dimension only. If it has two dimensions, the indexer cannot be used.

## **Explicit arrays**

Describes explicit arrays in OPL.

Another useful syntax to dynamically create arrays to be used in expressions or constraints is to explicitly define the array using the [] notation and including any variables or values into it.

Mixes are not allowed. For example, you can write:

```
forall(i in R)
  allDifferent([x[i], y[i], z[i]);
```

## Appending arrays

You can concatenate several arrays using the append function.

For example, if you want to express that all variables from array x and array y are different, you can use an allDifferent constraint applied to the appended arrays, as shown in this example.

### Appending arrays

```
using CP;
range R=1..10;
dvar int x[R] in 0..20;
dvar int y[R] in 0..20;
minimize sum(i in R)(x[i]+y[i]);
subject to
{
   allDifferent(append(all(i in 1..2) x[i],all(i in 4..6) y[i]));
}
```

### Initialization of decision variable arrays

Shows how to initialize your decision variable arrays in OPL.

The dynamic collection of decision variables allows you to dynamically initialize an array of decision variables. The variables are then shared between the two arrays of variables.

Here is an example of what is possible:

```
int n = 5;
range R = 1..n;

dvar int x[i in R] in R;

dvar int y1 = x[1];
dvar int y[R] = all[R](i in R) x[i];
dvar int y2[i in R] = x[i];
dvar int y3[R] = [ i:x[i] | i in R ];
dvar int y4[R] = [ x[1], x[2], x[3], x[4], x[5] ];

dvar int y5[0..n-1] =
append(all(i in R: i<2) x[i], all(i in R: i>=2) x[i]);
```

## **Expressions**

Describes data and decision variable identifiers, integer and float expressions, aggregate operators, piecewise-linear functions (continuous and discontinuous), set expressions, and Boolean expressions.

## **Usage of expressions**

Describes how to use expressions in OPL.

Expressions are used in fundamentally different ways in OPL:

- to specify items in generic arrays and sets (described in this chapter)
- to filter iterations (see "Formal parameters" on page 59)
- to state constraints over decision variables (see "Constraints" on page 46)

In the first two cases, the expressions do not contain decision variables, since decision variables have no value at this stage of the computation. These expressions are said to be *ground* and they are subject to almost no restrictions.

In the second case, of course, the expressions may contain decision variables. Boolean expressions containing decision variables are called *constraints* and are subject to a number of restrictions; for example, float constraints must be linear, piecewise-linear, or quadratic.

## Data and decision variable identifiers

Describes the use of identifiers within OPL expressions.

Since data and decision variable identifiers are the basic components of expressions, we will review briefly here how they are used to build expressions. If r is a tuple with a field capacity of type T, then r.capacity is an expression of type T. If a is an n-dimensional array of type T, a[e 1]...[e n] is an expression of type T, provided that e i is well-typed. For instance, the excerpt int limit[routes] = ...;

```
dvar int transp[r in routes] in 0..limit[r];
```

contains an expression limit[r] of type integer. Indices of arrays can be complex expressions. For instance, the excerpt

```
int nbFlights = ...;
range Flight = 1..nbFlights;
{string} Employee = ...;
dvar int crew[Flight][Employee] in 0..1;
constraints {
    forall(e in Employee)
        forall(i in 1..nbFlights - 2)
            crew[i][e] + crew[i+1][e] + crew[i+2][e] >= 1;
}
```

contains an integer expression crew[i+1][e] whose first index is itself an integer expression.

## Integer and float expressions

Describes the use of constants, data, decision variables, and operators within OPL expressions.

## Integer expressions

Integer expressions are constructed from integer constants, integer data, integer decision variables, and the traditional integer operators such as +, -, \*, div, mod (or %). The operator div represents the integer division (for example, 8 div 3 == 2) and the operator mod or % represents the integer remainder. OPL also supports the function abs, which returns the absolute value of its argument, and the built-in constant maxint, which represents the largest integer representable in OPL.

Note that expressions involving large integers may produce an overflow. In the following example for int, an overflow is detected.

```
int a=maxint+2;
float b=infinity+2;
execute
{
    writeln(a);
    writeln(b);
}
```

The Problems tab in the IDE displays the error messages.

```
lint x = maxint;

0 int a=x±2;

3 float b=infinity+2;

6 (

6 (

7 writeln(b);

9);

10

E Problems IS Sorpting log ( Solutions  Conflicts  Relaxations ) Fingree log Statistics  Problem ( History ) Console ( Herors, 0 warrings, 0 others

Description 

© Errors (4 kems)

© Computation error: an overflow occurred.

© Exception of "annormnous#1" failed: "C\1LOG|CFLEX_Studio123[opf]examples]opf[yoor]foo.mod", line 7: Element "a" does not exist in OPI, model...

© Invalid initialization expression for element "a".

© Scripting runtime error: Element "a" does not exist in OPI, model...
```

Most int expressions (such as % or div) are not available for constraints defined in CPLEX models, but are available for CP models. See also "Constraints available in constraint programming" on page 56.

## Float expressions

Float expressions are constructed from floats, float data and variables, as well as operators such as +, -, /, \*. In addition, OPL contains a float constant infinity to represent  $\infty$  and a variety of float functions, depicted in *OPL functions* in the *Language Quick Reference*.

## Conditional expressions

Conditional expressions are expressed like this:

```
(condition)?thenExpr : elseExpr
```

where condition is a ground condition with no decision variable. If condition is true, the condition evaluates to thenExpr; otherwise, it evaluates to elseExpr.

### Examples

```
int value = ...;
int signValue = (value>0) ? 1 : (value<0) ? -1 : 0;
int absValue = (value>=0) ? value : -value;
```

See the numeric functions in Summary table of OPL functions in the Language Quick Reference.

## **Counting expressions**

Among integer expressions, there are also some combinatorial expressions. For example, you can use the count function to count the number of times a particular value appears in an array of decision variables. You can use such an expression in modeling constraints only if the modeling constraints are part of a model that is solved by the CP Optimizer engine (that is, starting with the using CP; statement).

```
The constraint count(x, 2) == 3;
```

states that in the array of variables x, exactly three variables take the value 2.

For more information, see count in the Language Quick Reference.

# **Aggregate operators**

Describes the operators available for computing integer and float summations.

Integer and float expressions can also be constructed using aggregate operators for computing summations (sum), products (prod), minima (min), and maxima (max) of a collection of related expressions. For instance, the excerpt

```
int capacity[Routes] = ...;
int minCap = min(r in Routes) capacity[r];
```

uses the aggregate operator min to compute the minimum value in array capacity. The form of the formal parameters in these aggregate operators is very general and is discussed at length in "Formal parameters" on page 59.

For information on operators in general, see Operators in the Language Quick Reference.

## **Piecewise-linear functions**

Describes the use of piecewise-linear functions in OPL.

Piecewise-linear functions are important in many applications. They are often specified by giving a set of slopes, a set of breakpoints at which the slopes change, and the value of the functions at a given point. Consider, for instance, a transportation problem in which the transportation cost between two locations o and d depends on the size of the shipment ship[o][d]. The piecewise-linear expression

```
piecewise{10 -> 100;20 -> 200;40}(0,0) ship[o][d];
```

describes the piecewise-linear function of ship[0,d] depicted in Figure 1 The function has slopes 10, 20, and 40, breakpoints 100 and 200, and evaluates to 0 at point 0.

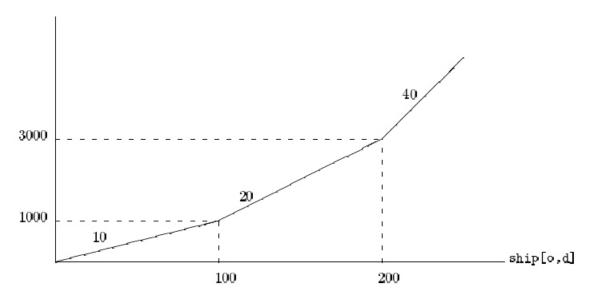


Figure 1. A Piecewise-linear function.

In other words, the piecewise-linear expression is equivalent to the expression: 10 \* ship[o][d]

```
when
ship[o,d] <= 100
equivalent to
(10 * 100) + 20 * (ship[o][d] - 100)
when
100 <= ship[o][d] <= 200
and equivalent to
(10 * 100) + (20 * 200) + 40 * (ship[o][d] - 200)
otherwise.</pre>
```

By default, OPL assumes that a piecewise-linear function evaluates to zero at the origin, so that the last piecewise-linear function could actually be written as

```
piecewise\{10 \rightarrow 100; 20 \rightarrow 200; 40\} ship[o][d];
The last piecewise-linear function has a fixed number of pieces, but OPL also
allows generic pieces. The number of pieces may then depend on the input data,
piecewise(i in 1..n) {
  slope[i] -> breakpoint[i];
  slope[n+1];
} ship[o][d];
If breakpoint[1] \ge 0 this piecewise-linear function is equivalent to
slope[1] * ship[o][d]
when
ship[o][d] <= breakpoint[1]</pre>
It is equivalent to
slope[1] * breakpoint[1] +
\sum_{i=2}^{n} slope [i] * (breakpoint[i] - breakpoint[i-1]) +
slope[k] * (ship[o][d] - breakpoint[k-1])
when
breakpoint[k-1] < ship[o][d] <= breakpoint[k] (1 < k <= n)
and equivalent to
slope[1] * breakpoint[1] +
```

 $\sum_{i=1}^{n} slope [i] * (breakpoint[i] - breakpoint[i-1]) +$ 

slope[n+1] \* (ship[o][d] - breakpoint[n])

otherwise.

Note that there may be several generic pieces in piecewise-linear functions. It is important to stress that breakpoints and slopes in piecewise-linear functions must always be grounded by a point on the piecewise-linear function. Such a point (called an anchor point) uniquely defines the function. Also, the breakpoints must be strictly increasing.

To sort your model data for this purpose, use sorted sets, as explained in "Sorted and ordered sets" on page 13.

Section Piecewise linear programming in the *Language User's Manual* discusses piecewise-linear functions applied to an inventory problem.

# Discontinuous piecewise-linear functions

Describes the use of discontinuous piecewise-linear functions in OPL.

OPL also allows you to write discontinuous piecewise-linear functions. This is the case when, in the syntax of a piecewise-linear function with slopes and breakpoints, two successive breakpoints are identical and the value associated with the second one is considered to be a "step" instead of a "slope". The CPLEX and

the CP Optimizer engines behave differently with respect to what limit they consider as the discontinuity value. Because CPLEX allows either of these limits, note that the anchor point used to ground the breakpoints and slopes must not reside at the discontinuity. Otherwise, the piecewise-linear function would not be uniquely defined.

## Behavior with the CPLEX engine

Example 1: the sign function

```
The following piecewise function: piecewise{0 -> 0; 2 -> 0; 0}(1,1) x;
```

has a slope of 0 up to breakpoint 0, then a step of 2 at this breakpoint, then a slope of 0. It takes the value 1 at point 1. This piecewise represents the function sign() which returns the sign (1 or -1) of its argument, as represented in Figure 1 below.

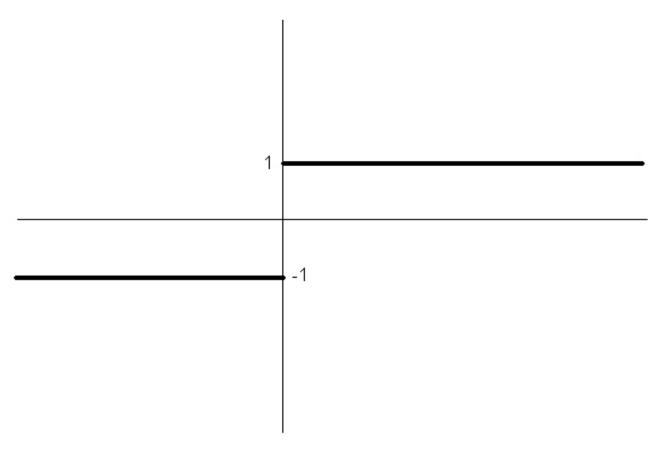


Figure 2. The discontinuous piecewise-linear function sign()

```
Then this model
dvar float x;
dvar float signx;
dvar float y;
dvar float signy;
maximize x;
subject to {
    x == 2;
```

```
signx == piecewise{0->0; 2->0; 0}(1,1) x;
y == -2;
signy == piecewise{0->0; 2->0; 0}(1,1) y;
}

gives the following output:
Final solution with objective 2.0000:
x = 2.0000;
signx = 1.0000;
y = -2.0000;
signy = -1.0000;
```

Figure 2 on page 41 shows that the value of the sign function at the breakpoint is either -1 (on the left-hand slope) or 1 (on the right-hand slope).

For example, this model takes this into account and sets the constraint x==y; on both values.

```
dvar float x;
dvar float signx;

dvar float y;
dvar float signy;

maximize signx-signy;
subject to {
    x == y;
    signx == piecewise{0->0; 2->0; 0}(1,1) x;
    signy == piecewise{0->0; 2->0; 0}(1,1) y;
}
```

This model solves with the following output:

```
Final solution with objective 2:
signx = 1;
signy = -1;
x = 0;
y = 0;
```

Example 2: discontinuous cost

The following piecewise function piecewise{0->0; 10->0; 0->10; 5->10; 0->20; 5->20; 0} (5,10) unit; represents a discontinuous cost.

This function is illustrated in Figure 2, *Discontinuous costs*, for the values summarized in Table 1.

Table 4. A discontinuous cost function

Values of Unit	Cost
&lt0	0
0 to 10	10
10 to 20	15
>20	20

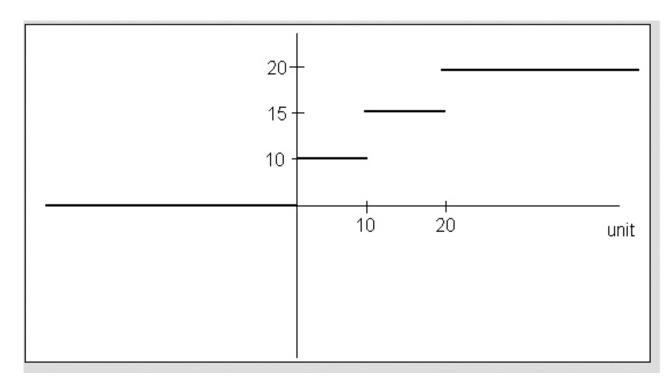


Figure 3. Discontinuous costs

## Different behavior with the CP Optimizer engine

Consider the following model:

```
//using CP;
dvar int x in -10..10;
dvar int signx;
dvar int y in -10..10;
dvar int signy;
maximize signx-signy;
    subject to {
        x == y;
        signx == piecewise{0->0; 2->0; 0}(1,1) x;
        signy == piecewise{0->0; 2->0; 0}(1,1) y;
    }
execute
{
    writeln(signx-signy);
}
```

Depending on which solving engine you write for, you get a different result because CPLEX and CP Optimizer do not handle limit values in the same way.

- If you comment out the using CP; statement, the model is solved by the CPLEX engine and the result is 2 because CPLEX handles symmetry in such a way that it interprets either limit as the discontinuity value.
- However, if you uncomment the using CP; statement, the model is solved by the CP Optimizer engine and the result is 0 because CP Optimizer considers the left limit as the discontinuity value.

# **Set expressions**

Describes the use of set expressions in OPL.

Set data can be initialized by set expressions, as mentioned in "Data types" on page 4. This section describes how these expressions are constructed and what functions are defined over sets.

- "Construction of set expressions"
- "Functions for sets"

## Construction of set expressions

Set expressions are constructed from previously defined sets and the set operations union, inter, diff, and symdiff. For instance:

```
{int} s1 = {1,2,3};
{int} s2 = {1,4,5};
{int} i = s1 inter s2;
{int} u = s1 union s2;
{int} d = s1 diff s2;
{int} sd = s1 symdiff s2;
```

initializes i to  $\{1\}$ , u to  $\{1,2,3,4,5\}$ , d to  $\{2,3\}$ , and sd to  $\{2,3,4,5\}$ . In addition, set expressions can be constructed from ranges. For instance, the excerpt  $\{int\}$  s = asSet(1..10);

initializes s to the finite set  $\{1,2,\ldots,10\}$ 

### Important:

The range 1..10 takes constant space, while the set s takes space proportional to the number of elements in the range.

#### **Functions for sets**

Table 5 shows the functions available over sets. These methods apply to all sets, including sets of tuples and other complex sets. In this section, we assume that:

- S is the set 3, 6, 7, 9
- item is an element of S
- n is an integer number

Table 5. Functions over sets

Function	Description
card	card(S) returns the size of S, that is, the number of items.
ord	ord(S,item) returns the position of item in S. Positions start at 0 and ord(S,item) produces an execution error if item is not in S.
	Example: ord(\$,6) evaluates to 1 and ord(\$,9) to 3.
	The order of items in an explicit set is by order of appearance in the initialization and is implementation-dependent when the sets are the results of a set operation.
first	first(S) returns the first item in S, 3 in this example.

Table 5. Functions over sets (continued)

Function	Description
item	<pre>item(\$,n) returns the n-th item in set S. Counting starts from 0. This is equivalent to next(first(\$),n)</pre>
	Example: $item(S,1) = 6$
last	last(S) returns the last item in S, 9 in this example.
next	next(S,item) returns the item in S that comes after item and produces an execution error if item is the last item.
	Example: next(\$,3) = 6
	<pre>next(S,item,n) returns the n-th next item. next(S,item) is equivalent to next(S,item,1).</pre>
nextc	A circular version of next. nextC(S,item) returns the first item in S if item is the last item.
	Example: nextc(\$,9) = 3
	<pre>nextc(S,item,n) returns the n-th circular next item. nextc(S,item) is equivalent to nextc(S,item,1).</pre>
prev	prev(S,item) returns the item in S that comes before item and produces an execution error if item is the first item.
	Example: prev(S,6) = 3
	<pre>prev(S,item,n) returns the n-th previous item. prev(S,item) is equivalent to prev(S,item,1).</pre>
prevc	A circular version of prev. prevc(S,item) returns the last item in S if item is the first item.
	Example: prev(S,3) = 9
	<pre>prevc(S,item,n) returns the n-th circular previous item. prevc(S,item) is equivalent to prevc(S,item,1).</pre>

# **Boolean expressions**

Describes the use of Boolean expressions in OPL.

Boolean expressions can have various operand types in OPL. They are constructed in different ways:

- from integer expressions using the traditional relational operators ==, != (not equal), >=, >, <, and <=.
- from float expressions using the same relational operators.
- from string expressions and support the same operators as well.

For convenience, OPL offers a range expression to express special combinations for constraints.

```
They are of the form a op1 x op2 b
```

#### where

- op1 and op2 are either of the relational operators <= or <</li>
- a and b are boundary expressions which need to be ground
- x is an expression

```
Those expressions are equivalent to a op1 x; x op2 b

which is itself equivalent to a op1 x && x op2 b
```

## **Constraints**

Specifies the constraints supported by OPL and discusses various subclasses of constraints to illustrate the support available for modeling combinatorial optimization applications.

## Introduction

Constraints are a subset of Boolean expressions.

The availability of certain constraints depends on their context. The contexts can be:

- · Data initialization when declared data is assigned
- Optimization model
  - a constraints block
  - a subject to block
- An expression that filters an iteration for aggregation or generation.

# **Using constraints**

Explains how to apply a constraint to a decision variable and, possibly, conditionalize it. Also explains why constraints are identified by a label, and why constraints are used for filtering purposes.

## **Modeling constraints**

Shows how to define modeling constraints in OPL.

Constraints passed to the algorithm, and which as such define the optimization problem, usually apply to decision variables; that is, they are Boolean expressions of some decision variables. To be taken into account by the solving algorithm, constraints must be stated using the optimization instruction:

```
constraints
or
subject to
```

as shown in the following example.

## Stating constraints by means of an optimization instruction

```
sum(p in Products) (insideCost[p]*inside[p] + outsideCost[p]*outside[p]);
subject to {
  forall(r in Resources)
    sum(p in Products) consumption[p,r] * inside[p] <= capacity[r];
  forall(p in Products)
    inside[p] + outside[p] >= demand[p];
}
```

#### Note:

- 1. Optimization instructions require an objective function of type integer or float.
- 2. That objective function <u>must be defined before the constraints.</u> Otherwise, a warning message is displayed.

## **Conditional constraints**

Shows how to define conditional constraints in OPL.

If-then-else statements make it possible to state constraints conditionally, as in:

```
if ( d>1 ) {
  abs(freq[f] - freq[g]) >= d;
} else {
  freq[f] == freq[g];
}
```

Conditions in if-else statements must be *ground*; that is, they must not contain decision variables. Also, the if block should not contain forall statements such as:

```
if (..) {
  forall(...)
   ...
}
```

Implications of constraints can be used instead when conditions contain decision variables. Conditionals can also be used in OPL to make different choices according to the truth value of a condition.

### Filtering with constraints

Describes the process of using constraints to filter decision variables or aggregates.

In addition to applying constraints to decision variables, you can also create constraints on formal parameters to filter aggregates, like this:

### Filtering with constraints

```
// The cities where we are doing business
{string} cities={"Paris","Berlin","Washington","Rio"};

{string} EuropeanMainCapitals = {"London","Paris","Berlin","Madrid","Roma"};

// Should we expand business in this city ?
dvar boolean x[cities];

// We want to expand business in Europe
maximize sum(c in cities: c in EuropeanMainCapitals) x[c];

subject to
{
    // We can expand business in 2 cities
    sum(c in cities) x[c]<=2;</pre>
```

```
{
string} expanded_cities = {c | c in cities : x[c] == 1};
execute
{
  for(c in expanded_cities) writeln("We should expand business in ",c);
}
The result is:
We should expand business in Paris
We should expand business in Berlin
```

In this context, filtering means placing a condition on an iteration to limit the iteration loop. See "Formal parameters" on page 59 for more examples.

## **Constraint labels**

Explains why label constraints, the benefits, costs, and limitations, how to label constraints, how to use indexed labels, and how to deal with compatibility between constraint names and labels.

### Why label constraints?

Explains why attaching labels to your constraints is the recommended practice.

You can identify constraints by attaching labels to them. It is the recommended practice but it has a performance cost.

#### **Benefits**

- Constraint labels enable you to benefit from the expand feature in the IDE Problem Browser to find which constraints are tight in a given application or to find dual variable values in linear programs. See <u>Understanding the Problem</u> Browser in *Getting Started with the IDE*.
- You can access the slack and dual values for labeled constraints when a solution is available. See the class in the *Reference Manual of IBM ILOG Script Extensions* for OPL.
- Only labeled constraints are considered by the relaxation and conflict search process in infeasible models (see Relaxing infeasible models in *IDE Tutorials*).

### Cost

However, labeling constraints has a performance and memory cost which can be significant, especially when a tuple pattern is used as the index. Therefore, you are encouraged to not use labels for large models or, if you do, at least use tuple indices instead of tuple patterns.

More precisely, constraint labels are used in three cases: <u>IDE</u> expand actions, in <u>slack and dual values with solutions</u>, and with relaxation and conflict detection. If you do not need these three use cases, you should get rid of the label to speed up the execution and lower memory consumption.

### Labeling constraints

Shows how to label constraints in OPL.

#### **Procedure**

Type the character string you want, followed by the colon (:) sign, before the constraint you want to label, as shown in **Labelling constraints (production.mod)**. If you used to declare constraint names in your existing OPL models, see "Compatibility between constraint names and labels" on page 52.

#### Note:

A constraint label or name cannot start with a number. Labeling constraints (production.mod)

```
sum(nimize
  sum( p in Products )
    ( InsideCost[p] * Inside[p] + OutsideCost[p] * Outside[p] );

subject to {
  forall( r in Resources )
    ctCapacity:
    sum( p in Products )
        Consumption[p][r] * Inside[p] <= Capacity[r];

forall(p in Products)
    ctDemand:
    Inside[p] + Outside[p] >= Demand[p];
}
```

This is equivalent to "Stating constraints by means of an optimization instruction" on page 47. The only difference is that the constraint on the production capacity has been labeled

```
ctCapacity:
```

and the constraint on the demand of products has been labeled

These labels can be used to display the data.

## Using indexed labels

Shows how to use indexed labels for your OPL constraints.

#### About this task

In some cases, it is more convenient to use indexed labels. Indexed labels enable you to control how a constraint is assigned to an array item.

The following example, Labeling constraints with indexed labels shows that the transp2.mod model identifies constraints using indexed labels following this syntax:

```
constraint ctDemand[Products];
...
ctDemand[p]:...

Indexed labels on constraints (transp2.mod)
  forall( p in Products , d in Dest[p] )
    ctDemand[p][d]:
    sum( o in Orig[p] )
    Trans[< p,o,d >] == Demand[<p,d>];
  ctCapacity: forall( o , d in Cities )
        sum( <p,o,d> in Routes )
        Trans[<p,o,d>] <= Capacity;</pre>
```

A case where you need indexed labels to reduce memory overhead is when you use forall iterations with variable sizes, as shown in the following code.

```
forall iterations with variable sizes
forall( p in Products , o in Orig[p] )
```

In this forall example, the second formal parameter o iterates on sets of potentially different sizes, depending on the value of the formal parameter p.

To use indexed labels:

### **Procedure**

- Declare the constraint array that will receive the labeled constraints. constraint ctSupply[Products][Cities];
- Add the indexing expressions to the label. ctSupply[p][o]:

### **Example**

The full code extract is shown here.

```
Labeling constraints with indexed labels
constraint ctSupply[Products][Cities];
constraint ctDemand[Products][Cities];

minimize
    sum(1 in Routes) Cost[1] * Trans[1];

subject to {
    forall( p in Products , o in Orig[p] )

        ctSupply[p][o]:
        sum( d in Dest[p] )
            Trans[< p,o,d >] == Supply[<p,o>];
    forall( p in Products , d in Dest[p] )
        ctDemand[p][d]:
        sum( o in Orig[p] )
        Trans[< p,o,d >] == Demand[<p,d>];
```

### Labeled assert statements

Shows how to label assertions in OPL.

Assertions can be labeled. When you label a constraint that is part of an assert statement, and if the assertion fails, the context of the failing assertions appears in the Issues output window. For example:

```
{int} vals = {1, 2, 3};
assert forall(i in vals) ct:i<2;</pre>
```

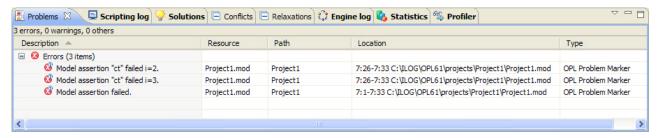


Figure 4. Labeled assert

## Limitations on constraint labeling

Explains what constraints cannot be labeled or at what cost they can be.

Not all constraints can be labeled. Limitations exist with respect to forall statements and to a variable size indexer.

### Labels and forall statements

You can label only constraints that are not nested within a forall statement (leaf constraints). However, you can label a forall constraint, if it is at the root level of constraints. For example, in the code sample Constraint label within forall statement, the constraint

```
ct1: forall(i in r1) forall(j in r2) X[i][j] <= i+j;
can also be written
forall(i in r1) forall(j in r2) ct1: X[i][j] <= i+j;</pre>
```

In both cases, the model executes correctly. However, if you execute: forall(i in r1) ct1: forall(j in r2) X[i][j] <= i+j;

the IDE reports "Element "ct1" has never been used" and the constraint does not appear in the Problem Browser.

#### Constraint label within forall statement

```
range r1 = 1..2;
range r2 = 1..3;

dvar int X[r1][r2] in 0..5;

constraints {
ct1: forall(i in r1) forall(j in r2) X[i][j] <= i+j;
}</pre>
```

### Labels and a variable size indexer

Constraint labels with a variable size indexer are not allowed. For example, this model generates an error message.

```
tuple RangeTuple
{
    int i;
    int j;
    string k;
};
{RangeTuple} RT = {<1, 2, "bla">};
```

```
minimize 1;
 subject to
     forall(<p1, p2, p3> in RT)
         forall(i in p1..p2)
     rangeLabel:
     1 == 1;
Write the following code instead.
tuple RangeTuple
    int i;
   int j;
   string k;
{RangeTuple} RT = {<1, 2, "bla">};
{int} s=\{1,2\};
constraint rangeLabel[RT][s]
minimize 1;
subject to
    forall(<p1, p2, p3> in RT)
        forall(i in p1..p2)
    rangeLabel[<p1, p2, p3>,i]:
    1 == 1;
}
```

Note the difference in rangeLabel.

## Compatibility between constraint names and labels

Explains compatibility issues for constraint names between OPL 4.x and subsequent versions.

OPL 4.x constraint names are deprecated in OPL 5.0 and later. They are still supported to maintain the compatibility of your OPL 4.x models and automatically implemented as labels internally, but they will be removed in future versions of IBM ILOG OPL. It is therefore strongly recommended that:

- you label constraints that are currently neither named nor labeled, as shown in "Labeling constraints" on page 48,
- you change possible existing constraint names in your models to labels. Compare Deprecated constraint names with Labeling constraints (production.mod). You do not need to previously declare the label as was the case with constraint names and you use the colon (:) sign instead of the equal (=) sign.

Deprecated constraint names

```
constraint capacityCons[r];
constraint demandCons[p];
minimize
   sum(p in Products) (insideCost[p]*inside[p] + outsideCost[p]*outside[p]);
subject to {
   forall(r in Resources)
```

```
capacityCons[r] = sum(p in Products) consumption[p,r] * inside[p] <= capacity[r];
forall(p in Products)
  demandCons[p] = inside[p] + outside[p] >= demand[p];
}
```

## Types of constraints

Classifies constraints according to their operand type.

### Float constraints

Float constraints are constraints involving float decision variables.

When **modeled** in OPL, they are restricted to be linear or piecewise linear. OPL has efficient algorithms for solving linear, piecewise linear, quadratic, or logical constraints, but in general these algorithms do not apply to nonlinear problems. Note that the linearity requirement precludes the use of relations with variables in constraints and the use of non-ground expressions as indices of float arrays. In addition, operators !=, <, and > are not allowed for float constraints. However, integers and integer variables may occur in a float constraint, provided that the constraint remains linear or quadratic. OPL supports all the expressions supported by CPLEX, provided that the constraint is of one of the types described at the beginning of this paragraph.

## **Discrete constraints**

Discrete constraints are arbitrary Boolean expressions with integer operands, possibly containing variables.

In OPL, these constraints must be well-typed, but no restrictions are imposed on them. It is, however, useful to review subclasses of these constraints to illustrate the functionalities of OPL. MP models, solved by the CPLEX engine, can contain only basic constraints (see the section *Basic constraints* below. Logical constraints can be set by filtering of forall and sum constraints.

#### **Basic constraints**

Basic discrete constraints are constructed from discrete data, discrete variables, and the arithmetic operators and functions defined in "Expressions" on page 36. For instance, the excerpt

```
range r=1..5;
dvar int x[1..5] in 0..10;
dvar int obj;

maximize obj;

subject to
{
  obj==sum(ordered i,j in r) abs(x[i]-x[j]);
  forall(ordered i,j in r) abs(x[i]-x[j])>=1;
}
```

generates distance constraints between integer variables.

Note that the following code creates an error because IBM ILOG CPLEX does not accept non linear constraints.

```
dvar int+ X in 0..1000;
minimize X;
subject to {
X mod 7 == 0;
};
```

## String constraints

Describes the use of string constraints in OPL.

String constraints cannot be used on decision variables or added to the model to be solved. They can only be used on indexers to filter aggregates (see "Filtering with constraints" on page 47). For example:

```
{string} s = {"a", "b"};
dvar int x[s] in 0..10;
minimize sum(i in s) x[i];
subject to {
   forall(i in s : i != "a")
   x[i] >= 5;
}
```

## Implicit constraints

Explains that implicit constraints may imply infeasibility.

Implicit constraints are implied by operators. For example:

```
using CP;
int a[1..4];

dvar int x in 1..5;

maximize a[x];

subject to {
    x==5;
}

makes the model infeasible even though
using CP;
int a[1..4];
```

dvar int x in 1..5;

subject to
{
 x==5;
}

is feasible.

## Logical constraints for CPLEX

Describes the use of logical constraints in OPL.

Logical constraints are one particular kind of discrete or numerical constraints. OPL and CPLEX can translate logical constraints automatically into their transformed equivalent that the discrete (MIP) or continuous (LP) optimizers of IBM ILOG CPLEX can process efficiently. This section describes all the available logical constraints, as well as the logical expressions that can be used in logical constraints. Logical constraints are available in constraint programming models without linearization.

For an example of how OPL uses logical constraints, see Tutorial: Using CPLEX logical constraints in the *Language User's Manual*.

In this section, you will learn:

• "What are logical constraints?" on page 55

- "What can be extracted from a model with logical constraints?"
- "Which nonlinear expressions can be extracted?"
- "Logical constraints for counting" on page 56
- "How are logical constraints extracted?" on page 56

## What are logical constraints?

For IBM ILOG CPLEX, a logical constraint combines linear constraints by means of logical operators, such as logical-and, logical-or, implication, negation (not), conditional statements (if ... then ...) to express complex relations between linear constraints. IBM ILOG CPLEX can also handle certain logical expressions appearing within a linear constraint. One such logical expression is the minimum of a set of variables. Another such logical expression is the absolute value of a variable. There's more about logical expressions in "Which nonlinear expressions can be extracted?."

### What can be extracted from a model with logical constraints?

The following table lists the logical constraints that CPLEX can extract.

Symbol	Meaning
&&	Logical AND
1 1	Logical OR
!	Logical NOT
=>	Imply
!=	Different from
==	Equivalence

All those constructs accept as their arguments other linear constraints or logical constraints, so you can combine linear constraints with logical constraints in complicated expressions in your application.

### Note:

With the C++ API you should use the symbol <= to express implication.

### Which nonlinear expressions can be extracted?

Some expressions are easily recognized as nonlinear, for example, a function such as

$$x^2 + y^2 = 1$$

However, other nonlinearities are less obvious, such as absolute value as a function. In a very real sense, MIP is a class of nonlinearly constrained problems because the integrality restriction destroys the property of convexity which any linear constraints otherwise might possess. Because of that characteristic, certain (although not all) nonlinearities are capable of being converted to a MIP formulation, and thus can be solved by IBM ILOG CPLEX. The following nonlinear expressions are accepted in an OPL model:

- min and min1: the minimum of several numeric expressions
- max and max1 : the maximum of several numeric expressions
- abs : the absolute value of a numeric expression

- piecewise: the piecewise linear combination of a numeric expression
- A linear constraint can appear as a term in a logical constraint.

In fact, ranges containing logical expressions can, in turn, appear in logical constraints. It is important to note here that only linear constraints can appear as arguments of logical constraints extracted by IBM ILOG CPLEX. That is, quadratic constraints are not handled in logical constraints. Similarly, quadratic terms can not appear as arguments of logical expressions such as min, max, abs, and piecewise.

## Logical constraints for counting

In many cases it is even unnecessary to allocate binary variables explicitly in order to gain the benefit of linear constraints within logical expressions. For example, optimizing how many items appear in a solution is often an issue in practical problems. Questions of counting (how many?) can be represented formally as cardinality constraints. Suppose that your application includes three variables, each representing a quantity of one of three products, and assume further that a good solution to the problem means that the quantity of at least two of the three products must be greater than 20. Then you can represent that idea in your application, like this:

$$(x[0] \ge 20) + (x[1] \ge 20) + (x[2] \ge 20) \ge 2;$$

### How are logical constraints extracted?

Logical constraints are transformed automatically into equivalent linear formulations when they are extracted by an IBM ILOG CPLEX algorithm. This transformation involves automatic creation by IBM ILOG CPLEX of new variables and constraints. For more details on this transformation, refer to the IBM ILOG CPLEX documentation.

### Constraints available in constraint programming

Lists the types of constraints and expressions available in OPL for constraint programming models. See the OPL *Language Quick Reference* and the *CP Optimizer User's Manual* for further details.

- "Arithmetic constraints and expressions"
- "Logical constraints for CP" on page 57
- "Element constraints" on page 57
- "Compatibility constraints" on page 58
- "Specialized constraints" on page 58

### Arithmetic constraints and expressions

The following arithmetic constraints and expressions are available for OPL CP models. The references point to the OPL *Language Quick Reference*.

- Arithmetic operations
  - addition
  - subtraction
  - multiplication
  - scalar products
  - integer division
  - floating-point division
  - modular arithmetic

- Arithmetic expressions for use in constraints.
  - standard deviation: see standardDeviation
  - minimum: see min
  - maximum: see max
  - counting: see count
  - absolute value: see abs
  - element or index: see element
- Arithmetic constraints
  - equal to
  - not equal to
  - strictly less than
  - strictly greater than
  - less than or equal to
  - greater than or equal to

## Logical constraints for CP

Logical operators, such as logical-and, logical-or, implication, negation (not), difference, and equivalence are available in constraint programming.

The following logical constraints are available for OPL CP models.

Symbol	Meaning
&&	Logical AND
1.1	Logical OR
!	Logical NOT
=>	Imply
!=	Different from
==	Equivalence

#### Note:

With the C++ API you should use the symbol <= to express implication.

### Element constraints

The following subscripting operators return constrained integer or floating-point expressions, also known as element expressions.

- A[X], where A is an array of integer values (with one or several dimensions) and X is a dexpr int, returns a dexpr int such that: when X is fixed to the value i, the value of the expression is A[i].
- A[X], where A is an array of floating-point values (with one or several dimensions) and X is a dexpr int, returns a dexpr float such that: when X is fixed to the value i, the value of the expression is A[i].

More generally, the domain of the expression is the set of values A[i], where i is in the domain of X.

In both of the previous cases, the expression can include several indices (A[X][k],A[X][Y], A[X][k][Y], ...).

- For constant indices, all the usual index types are available (int, float, string, tuple).
- For a decision expression index, only the integer type is available (dvar int or dexpr int).

## **Compatibility constraints**

The CP Optimizer engine supports allowed and forbidden assignments for OPL CP models. See allowedAssignments and forbiddenAssignments in the *Language Quick Reference*.

## Scheduling constraints

This section lists the scheduling constraints available for OPL CP scheduling models and provides links to the reference documentation for these constraints. For a more detailed description of these constraints, please refer to the "Scheduling" on page 63 section of this manual.

The following constraints are available for OPL CP scheduling models:

- · Precedence constraint: endAtEnd
- Precedence constraint: endAtStart
- · Precedence constraint: endBeforeEnd
- · Precedence constraint: endBeforeStart
- Precedence constraint: startAtEnd
- Precedence constraint: startAtStart
- Precedence constraint: startBeforeEnd
- Precedence constraint: startBeforeStart
- Interval grouping constraint: alternative
- Interval grouping constraint: span
- Interval grouping constraint: synchronize
- Interval presence constraint: presenceOf
- Sequence constraint: first (scheduling)
- Sequence constraint: last (scheduling)
- Sequence constraint: before
- Sequence constraint: prev
- Sequence constraint: noOverlap
- Cumulative or state function constraint: <= (scheduling), alwaysIn
- State function constraint: alwaysConstant
- State function constraint: alwaysEqual
- State function constraint: alwaysNoState

### Specialized constraints

The CP Optimizer engine also accepts some powerful combinatorial constraints known as specialized constraints. For these constraints, some powerful propagation algorithms are used to reduced the decision variable domains.

- allDifferent : constrains variables within a dvar array to all take different values
- allMinDistance: constrains variables within a dvar array to all take values that are one-to-one different by at least a given gap

- inverse: takes two arrays of integer variables that must be indexed by an integer and be one-dimensional
- lex : states that the first array of variables is less than or equal to the second array of variables in the alphabetical order
- · pack : represents some simple but powerful one-dimensional packing constraint

See the individual entries under OPL functions in *Language Quick Reference* for a complete description of each constraint.

## **Limitations on constraints**

There is a limitation on certain types of constraints.

You cannot change the bound of a top level inequality constraint unless that bound is a literal expression, such as  $x \ge 13$  or  $x \le 13$ .

As a workaround, you can declare your constraint either by using a range syntax, or by putting the inequality inside a forall statement.

```
Instead of
ct0: x ≤ a[0];
you can write
ct1: -infinity ≤ x ≤ a[0];
or
forall(i in 0..0)
   ct2: x ≤ a[0];
and change the upper (or lower) bound of the constraint with a script instruction such as:
ct1.UB = 1313;
or
ct2[0].UB = 1313;
```

# Formal parameters

Describes basic formal parameters, tuples of parameters, and filtering in tuples of parameters.

# **Basic formal parameters**

Formal parameters play a fundamental role in OPL; they are used in aggregate operators, generic sets, and forall statements.

The simplest formal parameter has the form

```
p in S
```

where p is the formal parameter and S is the set from which p takes its values.

The set *S* can be:

· an integer range, as in

```
int n=6;
int s = sum(i in 1..n) i*i;

• a string set, as in
    {string} Products = {"car", "truck"};
    float cost[Products] = [12000,10000];
    float maxCost = max(p in Products) cost[p];

• or a tuple set, as in
    {string} Cities = { "Paris", "London", "Berlin" };
    tuple Connection
    {
        string orig;
        string dest;
    }
    {Connection}    connections = { <"Paris", "Berlin">, <"Paris", "London">};
    float cost[connections] = [ 1000, 2000 ];
    float maxCost= max(r in connections) cost[r];
```

If you need to filter the range of the formal parameters using conditions, the formal parameter then takes the form

```
p in S: filtering condition
```

and assigns to *p* all elements of *S* according to the filter applied.

```
For instance, in the excerpt
int n=8;
dvar int a[1..n][1..n];
subject to
{
   forall(i in 1..8)
      forall(j in 1..8: i < j)
       a[i][j] >= 0;
}
```

the constraint a[i][j] >= 0 is modeled for all i and j such that  $1 \le i < j \le 8$ .

### Note:

OPL does not support aggregates in filter expressions. For example:

For an expression such as:
 {int} notFirst = {i | i in 1..10 : card({j | <i,j> in pairs}) == 0};
 the type check displays the error: "Aggregate set is currently not supported for

For an expression such as:
 {int} notFirst = {i | i in 1..10 : sum(<i,j>in pairs) i == 0};
 the type check displays the error: "Aggregate sum is currently not supported for filter expressions."

Several parameters can often be combined together to produce more compact statements. For instance, the declaration

```
int s = sum(i, j in 1..n: i < j) i*j;
is equivalent to
int s = sum(i in 1..n) sum(j in 1..n: i < j) i*j;</pre>
```

which is less readable.

filter expressions."

```
The declaration
int s = sum(i in 1..n, j in 1..m) i*j;

is equivalent to
int s = sum(i in 1..n) sum(j in 1..m) i*j;

These parameters can, of course, be subject to filtering conditions
The excerpt forall(i,j in 1..n : i < j)
    a[i][j] >= 0;

is equivalent to
forall(i in 1..n, j in 1..n : i < j)
    a[i][j] >= 0;

Here is an even more compact form:
forall(ordered i,j in 1..n)
    a[i][j] >= 0;
```

Indeed, in many applications one is interested, given a set S, in stating filters or conditions over all pairs (i, j) of elements of S satisfying i < j in the ordering associated with S. In this excerpt

```
{T} S = ...;
forall(ordered s, t in S)...;
forall(s in S, t in S: ord(S,s) < ord(S,t)) ...</pre>
```

the first forall line is equivalent to the second one and illustrates the functionality ordered, often useful in practical applications. T can be one of the types int, float, string, or a tuple type.

#### Important:

This ordering does not refer to the ordering associated with type T but to the order of the items within the set.

# **Tuples of parameters**

OPL allows tuples of formal parameters to be created in aggregate operators, forall statements, and generic sets.

The following code sample states the precedence constraints between tasks. The constraint declaration requires explicit accesses to the fields of the tuple to state the constraints. In addition, the field before is accessed twice. An alternative way to state the same constraint is to use a tuple of formal parameters as shown in the last line of the code sample, precluding the need to access the tuple fields explicitly. The tuple in the forall quantifier contains two components that are successively given the values of the fields of each tuple in Prec.

```
Tuple of formal parameters
int minTime=7*60;
int maxTime=9*60;

{string} Tasks = { "Make dinner", "Have dinner", "Clean post dinner" };
tuple Precedence {
    string pre;
    string post;
}

{Precedence} Prec = {
```

```
<"Make dinner","Have dinner">,
    <"Have dinner","Clean post dinner">
};
int Duration[Tasks] = [20,60,10];
dvar int Start[Tasks] in minTime..maxTime;
subject to {
  forall(p in Prec) Start[p.post] >= Start[p.pre] + Duration[p.pre];
}
```

More generally, an expression

```
p in S
```

where S is a set of tuples containing n fields, can be replaced by a formal parameter expression

```
<p1,...,pn> in S
```

that contains n formal parameters. Each time a tuple r is selected from S, its fields are assigned to the corresponding formal parameters. This functionality is often useful in producing more readable models.

## Filtering in tuples of parameters

OPL enables simple equality constraints to be factorized inside tuples, which is important in obtaining more readable and efficient models. In this context, slicing refers to nested iterations with filtering conditions.

Consider, for instance, a transportation problem where products must be shipped from one set of cities to another set of cities. The model may include a constraint specifying that the total shipments for all products transported along a connection may not exceed a specified limit. This can be expressed by a constraint

```
Explicit slicing
forall(c in connections)
   sum(<p,co> in routes: c == co) trans[<p,c>] <= limit;</pre>
```

This constraint states that the total products shipped along each connection c is not greater than limit. OPL must scan the entire set routes to select the tuples involving each connection. In this example, the expression c==co is used to make slicing *explicit*.

The constraint would be stated equivalently as follows:

```
Implicit slicing
forall(c in connections)
  sum(<p,c> in routes) trans[<p,c>] <= limit;</pre>
```

In this constraint, the tuple <p,c> contains one new parameter p and uses the previously defined parameter c. Since the value of c is known, OPL uses it to index the set routes, avoiding a complete scan of the set routes. In this example, slicing is said to be *implicit* because the formal parameter c is used to declare iteration in both the forall and sum loops. You can also use a constant as a tuple item, for example <p,2>, for implicit slicing.

In OPL 4.0 and later versions, models tend to be more readable when explicit slicing is used. Besides, there is no performance advantage in using implicit slicing over explicit slicing.

### More about implicit slicing

```
You should be aware that the following statement:
int array[i in set1] = ((sum(i in set2) 1 >= 1) ? 1:0);
is exactly equivalent to
int array[i in set1] = ((sum(j in set2) 1 >= 1) ? 1:0);
that is, the two "i" on either side of the "equal" sign =, in the first statement, are
not linked. This is called scope hiding because the second "i" hides the first one in a
nested scope.
```

```
In contrast, this statement
int array[<i,j> in set1] = ((sum(<i,j> in set2) 1 >= 1) ? 1:0);
codes implicit slicing, which is equivalent to:
int array[i in set1] = ((sum(j in set2 : j==i) 1 >= 1) ? 1:0);
In other words, there is no implicit slicing outside tuple patterns.
```

See also Modeling tips in the Language User's Manual.

## **Scheduling**

Describes how to model scheduling problems in OPL.

#### Introduction

OPL and CP Optimizer introduce a set of modeling features for applications that deal with scheduling over time.

In OPL and CP Optimizer, time points are represented as integers, but the possible very wide range of time points means that time is effectively continuous. A consequence of scheduling over effectively continuous time is that the evolution of some known quantities over time (for instance the instantaneous efficiency/speed of a resource or the earliness/tardiness cost for finishing an activity at a given date *t*) needs to be compactly represented in the model. To that end, CP Optimizer provides the notion of *piecewise linear* and *stepwise* functions.

Most scheduling applications consist in scheduling in time some activities, tasks or operations that have a start and an end time. In CP Optimizer, this type of decision variable is captured by the notion of the *interval variable*. Several types of constraints are expressed on and between interval variables:

- to limit the possible positions of an interval variable (forbidden start/end or "extent" values)
- to specify precedence relations between two interval variables
- to relate the position of an interval variable with one of a set of interval variables (spanning, synchronization, alternative).

An important characteristic of scheduling problems is that time intervals may be optional, and whether to execute a time-interval or not is a possible decision

variable. In CP Optimizer, this is captured by the notion of a boolean *presence* status associated with each interval variable. Logical relations can be expressed between the presence of interval variables, for example to state that whenever interval *a* is present then interval *b* must also be present.

Another aspect of scheduling is the allocation of scarce resources to time intervals. The evolution of a resource over time can be modelled by two types of variables:

- The evolution of a disjunctive resource over time can be described by the sequence of intervals that represent the activities executing on the resource. For that, CP Optimizer introduces the notion of an *interval sequence variable*. Constraints and expressions are available to control the sequencing of a set of interval variables.
- The evolution of a cumulative resource often needs a description of how the accumulated usage of the resource evolves over time. For that purpose, CP Optimizer provides the concept of the *cumulative function expression* that can be used to constrain the evolution of the resource usage over time.
- The evolution of a resource of infinite capacity, the state of which can vary over time, is captured in CP Optimizer by the notion of the *state function*. The dynamic evolution of a state function can be controlled with the notion of transition distance, and constraints are available for specifying conditions on the state function that must be satisfied during fixed or variable intervals.

Some classical cost functions in scheduling are earliness/tardiness costs, makespan, and activity execution or non-execution costs. CP Optimizer generalizes these classical cost functions and provides a set of basic expressions that can be combined together; this allows you to express a large spectrum of scheduling cost functions that can be efficiently exploited by the CP Optimizer search.

For the description of the symbolic notation used throughout this section, see "Notation" on page 84.

# Piecewise linear and stepwise functions

Describes piecewise linear and stepwise functions as related to scheduling.

In CP Optimizer, piecewise linear functions are typically used to model a known function of time, for instance the cost incurred for completing an activity after a known date *t*. Stepwise functions are typically used to model the efficiency of a resource over time.

A **piecewise linear function** F(t) is defined by a tuple  $F = piecewise(S, T, t_0, v_0)$  where:

- S is a vector of n+1 "change" values  $s_i \in \mathbb{R}$  for F
- T is a vector of n values  $t_i \in \mathbb{R}$  for F such that  $\forall i \in [1, n-1], t_i \leq t_{i+1} \land \forall i \in [1, n-2], t_i < t_{i+2}$
- (t<sub>0</sub>, v<sub>0</sub>) is a reference point of the function

The function is defined everywhere on  $(-\infty, +\infty)$  as follows:

- F(t<sub>0</sub>) = v<sub>0</sub>
- the slope F'(t) of F on the segment  $(-\infty, t_1)$  is equal to  $s_1$
- the slope F'(t) of F on the segment [t<sub>n</sub>, +∞) is equal to s<sub>n+1</sub>
- $\bullet \ \forall i \in [1,n-1]:$ 
  - if  $t_i \neq t_{i+1}$ , then the slope F'(t) of F on the segment  $[t_i, t_{i+1})$  is equal to  $s_{i+1}$ . That is,  $\forall t \in [t_i, t_{i+1})$ ,  $F(t) = F(t_i) + s_{i+1}(t t_i)$ . Note that the point  $t_{i+1}$  is not included in the segment.
  - if  $t_i = t_{i+1}$ , then F is discontinuous at point  $t_i$ .  $s_{i+1}$  represents the height of the step (positive or negative) at  $t_i$ . The value of  $F(t_i)$  is equal to the the value of the leftmost extremity of the segment immediately to the right of the discontinuity. Formally, this is given by:

```
* if i \neq n-1 then F(t_i) = F(t_{i+2}) + s_{i+2}(t_i - t_{i+2})

* else if i \neq 1 then F(t_i) = F(t_{i-1}) + s_i(t_i - t_{i-1}) + s_{i+1}

* { Otherwise, function has form slope / step / slope, and i = 1 }

* else if t_0 \geq t_i then F(t_i) = v_0 + s_3(t_i - t_0)

* else F(t_i) = v_0 + s_1(t_i - t_0) + s_2
```

For a complete description of the OPL syntax of a piecewise linear function, see piecewise and pwlFunction in the OPL Language Quick Reference.

A **stepwise function** is a special case of the piecewise linear function, where all slopes are equal to 0 and the domain and image of F are integer. A stepwise function F(t) is defined by a tuple F = stepwise(V, T) where:

- V is a vector of n+1 function values  $v_i \in \mathbb{Z}$  for F
- T is a vector of n values  $t_i \in \mathbb{Z}$  for F such that  $\forall i, t_i < t_{i+1}$

The function is defined everywhere on  $(-\infty, +\infty)$  and the function value at a point t is given by:

- $\forall t < t_1, F(t) = v_1$
- $\forall i \in [1, n-1], \forall t \in [t_i, t_{i+1}), F(t) = v_{i+1}$
- $\forall t \geq t_n, F(t) = v_{n+1}$

For a complete description of the OPL syntax of a stepwise linear function, see stepwise and stepFunction in the OPL Language Quick Reference.

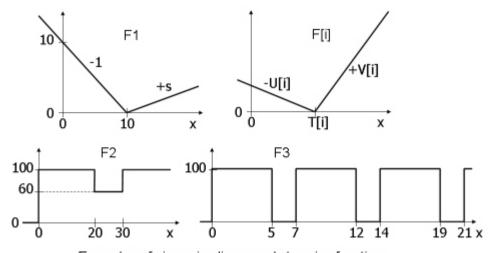
#### **Examples**

- A V-shape function with value 0 at x = 10, slope -1 before x = 10 and slope s afterwards:
  - pwlFunction F1 = piecewise $\{-1->10; s\}$  (10, 0);
- An array of V-shaped functions indexed by i in [1..n] with value 0 at T[i], slope -U[i] before T[i] and slope V[i] afterwards (T, U and V are data integer arrays): pwlFunction  $F[i \text{ in } 1..n] = piecewise\{ -U[i] -> T[i]; V[i] \} (T[i], \emptyset);$

- A stepwise function with value 0 before 0, 100 on [0, 20), value 60 on [20, 30), and value 100 later on:
  - stepFunction F2 = stepwise{ 0->0; 100->20; 60->30; 100 };
- A stepwise function with value 0 everywhere except on intervals [7*i*, 7*i*+5) for *i* in [0, 51] where the value is 100:

```
stepFunction F3 = stepwise(i in 0..51, p in 0..1) { 100*p -> (7*i)+(5*p); 0 };
```

The previous piecewise and stepwise function are depicted in the following diagram.



Examples of piecewise linear and stepwise functions

### Interval variables

Describes a basic building block of scheduling, the *interval*.

Informally speaking, an interval variable represents an interval of time during which something happens (a task, an activity is carried out) and whose position in time is an unknown of the scheduling problem. An interval is characterized by a start value, an end value and a size. An important feature of interval variables is the fact that they can be optional; that is, one can decide not to consider them in the solution schedule. This concept is crucial in applications that present at least some of the following features:

- optional activities (operations, tasks) that can be left unperformed (with an impact on the cost); examples include externalized, maintenance or control tasks
- activities that can execute on a set of alternative resources (machines, manpower) with possibly different characteristics (speed, calendar) and compatibility constraints
- operations that can be processed in different temporal modes (for instance in series or in parallel)
- alternative modes for executing a given activity, each mode specifying a particular combination of resources
- alternative processes for executing a given production order, a process being specified as a sequence of operations requiring resources
- hierarchical description of a project as a work-breakdown structure with tasks decomposed into sub-tasks, part of the project being optional (with an impact on the cost if unperformed), and so forth.

Formally, an interval variable a is a variable whose domain dom(a) is a subset of  $\{\bot\} \cup \{[s,e)|s,e\in\mathbb{Z},s\leq e\}$ . An interval variable is said to be *fixed* if its domain is reduced to a singleton; that is, if  $\underline{a}$  denotes a fixed interval variable:

• interval is absent:  $\underline{a} = \bot$ ; or • interval is present:  $\underline{a} = [s,e)$ 

Absent interval variables have special meaning. Informally speaking, an absent interval variable is not considered by any constraint or expression on interval variables it is involved in. For example, if an absent interval variable is used in a noOverlap constraint, the constraint will behave as if the interval was never specified to the constraint. If an absent interval variable a is used in a precedence constraint between interval variables a and b this constraint does not impact interval variable b. Each constraint specifies how it handles absent interval variables.

The semantics of constraints defined over interval variables is described by the properties that fixed intervals must have in order the constraint to be true. If a fixed interval  $\underline{a}$  is present and such that  $\underline{a} = [s, e)$ , we will denote  $s(\underline{a})$  its integer start value s,  $e(\underline{a})$  its integer end value e and  $l(\underline{a})$  its positive integer length defined as  $e(\underline{a})-s(\underline{a})$ . The presence status  $x(\underline{a})$  will be equal to 1. For a fixed interval that is absent,  $x(\underline{a}) = 0$  and the start, end and length are undefined.

Until a solution is found it may not be known whether an interval will be present or not. In this case we say that the interval is optional. To be precise, an interval is said to be absent when  $dom(a) = \{\bot\}$ , present when  $\bot \not\in dom(a)$  and optional in all other cases.

## Intensity and size

Sometimes the intensity of "work" is not the same during the whole interval. For example let's consider a worker who does not work during weekends (his work intensity during weekends is 0%) and on Friday he works only for half a day (his intensity during Friday is 50%). For this worker, 7 man-days of work will span for longer than just 7 days. In this example 7 man-days represents what we call the *size* of the interval; that is, what the length of the interval would be if the intensity function was always at 100%.

To model such situations, you can specify a range for the *size* of an interval variable and an integer *stepwise intensity function F*. For a fixed present interval  $\underline{a}$  the following relation will be enforced at any solution between the start, end, size sz of the interval and the integer granularity G (by default, the intensity function is expressed as a percentage so the granularity G is 100):

$$sz(\underline{a}) \le \int_{s(\underline{a})}^{\epsilon(\underline{a})} \frac{F(t)}{G} dt - \langle sz(\underline{a}) + 1 \rangle$$

That is, the length of the interval will be at least long enough to cover the work requirements given by the interval size, taking into account the intensity function. However, any over-estimation is always strictly less than one work unit.

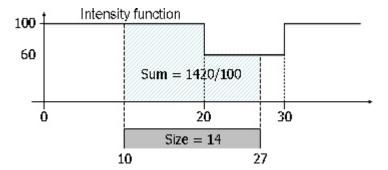
If no intensity is specified, it is supposed to be the constant full intensity function  $\forall t$ , F(t) = 100% so in that case sz(a) = l(a). Note that the size is not defined for absent intervals.

#### Important:

The intensity step function *F* should be a stepwise function with integer values and is not allowed to exceed the granularity (100 by default).

The following figure depicts an interval variable of size 14 with its intensity function. A valid solution is represented where the interval starts at 10 and ends at 27. Indeed in this case:

$$\int_{s(a)}^{e(\underline{a})} \frac{F(t)}{G} dt = \frac{1420}{100} = 14.2$$



#### **OPL** formulation

Typically, the problem structure will indicate if an interval can be optional or not, and the keyword optional is used (or not) in the definition of the interval variable. In the case where the optionality depends on input data, you can specify a boolean parameter to the optionality field: optional(true) being equivalent to optional and optional(false) being equivalent to the omission of optional.

A window [StartMin,EndMax] can be specified to restrict the position of the interval variable. By default, an interval variable will start after 0 and end before maxint/2. The fixed size or the size range for the interval is specified with the size keyword. Note that these bounds are taken into account only when the interval variable is present in the final schedule, that is, they allow specifying conditional bounds on the interval variable would the interval be present in the final schedule. For absent intervals, they are just ignored.

Where:

```
int IsOptional, StartMin, EndMax, SZ, SZMin, SZMax;
stepFunction F;
-maxint/2 + 1 <= StartMin <= maxint/2 - 1</pre>
```

```
-maxint/2 + 1 <= EndMax <= maxint/2 - 1
0 <= SZ <= maxint/2 - 1
0 <= SZMin <= maxint/2 - 1
0 <= SZMax <= maxint/2 - 1
```

### **Examples**

For examples of using *interval*, see the CP keywords interval, optional, size, and intensity in the OPL Language Quick Reference.

## Display of interval variable domain

The domain of an interval variable is displayed as shown in this example:

```
A1[0..1: 10..990 -- (5..10)5..990 --> 25..1000]
```

After the name of the interval variable (here A1), the first range (here 0..1) represents the domain of the boolean presence status of the interval variable. Thus 0..1 represents an optional interval variable whose status has still not been fixed, 0 an absent interval variable and 1 a present interval variable.

The remaining fields describe the position of the interval variable, it is omitted if the interval variable is absent as this information is not relevant in this case. Thus, an absent interval variable is displayed as:

A1[0]

When the interval variable is possibly present:

- the first range in the remaining fields represents the domain of the interval start
- the second range (between parenthesis) represents the domain of the interval size
- the third range represents the domain of the interval length
- the fourth and last range represents the domain of the interval end

Note that the second range may be omitted in case the size and length of the interval variable are necessarily equal.

When the values are fixed, the ranges *min..max* are replaced by a single value. For instance, the following display represents a fixed interval variable of size 5 that is present, starts at 10 and ends at 35:

-元约束

size length

## Unary constraints on interval variables

CP Optimizer provides constraints for modeling restrictions that an interval cannot start, cannot end or cannot overlap a set of fixed dates.

Let  $\underline{a}$  denote a fixed interval and F an integer stepwise function.

- **Forbidden start**. The constraint  $forbidStart(\underline{a}, F)$ , states that whenever the interval is present, it cannot start at a value t where F(t) = 0.
- **Forbidden end**. The constraint  $forbidEnd(\underline{a}, F)$ , states that whenever the interval is present, it cannot end at a value t where F(t 1) = 0.
- **Forbidden extent**. The constraint  $forbidExtent(\underline{a}, F)$ , states that whenever the interval is present, it cannot overlap a point t where F(t) = 0.

More formally:

$$\begin{split} &forbidStart(\underline{a},F) &\Leftrightarrow & \left(\underline{a}=\bot\right) \vee \left(F(s(\underline{a})) \neq 0\right) \\ &forbidEnd(\underline{a},F) &\Leftrightarrow & \left(\underline{a}=\bot\right) \vee \left(F(e(\underline{a})-1) \neq 0\right) \\ &forbidExtent(\underline{a},F) &\Leftrightarrow & \left(\underline{a}=\bot\right) \vee \left(\forall t \in [s(\underline{a}),e(\underline{a})),F(t) \neq 0\right) \end{split}$$

For syntax and examples of these constraints, see forbidEnd, forbidExtent, and forbidStart in the OPL Language Quick Reference. Note that none of these constraints can be used in meta-constraints.

### Precedence constraints between interval variables

Precedence constraints are common scheduling constraints used to restrict the relative position of interval variables in a solution.

For example a precedence constraint can model the fact that an activity a must end before activity b starts (optionally with some minimum delay z). If one or both or the interval variables of the precedence constraint is absent, then the precedence is systematically considered to be true; therefore it does not impact the schedule.

More formally, the semantics of the relation  $TC(\underline{a}, \underline{b}, z)$  on a pair of fixed intervals  $\underline{a}$ ,  $\underline{b}$  and for a value z depending on the constraint type TC is given in the following table.

Relation	Semantics
startBeforeStart	$x(\underline{a}) \wedge x(\underline{b}) \Rightarrow s(\underline{a}) + z \le s(\underline{b})$
startBeforeEnd	$x(\underline{a}) \wedge x(\underline{b}) \Rightarrow s(\underline{a}) + z \le e(\underline{b})$
endBeforeStart	$x(\underline{a}) \wedge x(\underline{b}) \Rightarrow e(\underline{a}) + z \le s(\underline{b})$
endBeforeEnd	$x(\underline{a}) \wedge x(\underline{b}) \Rightarrow e(\underline{a}) + z \le e(\underline{b})$
startAtStart	$x(\underline{a}) \wedge x(\underline{b}) \Rightarrow s(\underline{a}) + z = s(\underline{b})$
startAtEnd	$x(\underline{a}) \wedge x(\underline{b}) \Rightarrow s(\underline{a}) + z = e(\underline{b})$
endAtStart	$x(\underline{a}) \wedge x(\underline{b}) \Rightarrow e(\underline{a}) + z = s(\underline{b})$
endAtEnd	$x(\underline{a}) \wedge x(\underline{b}) \Rightarrow e(\underline{a}) + z = e(\underline{b})$

Precedence relations semantics

For syntax and examples, see the following functions described in the OPL Language Quick Reference. Note that none of these constraints may be used in a meta-constraint.

- endAtEnd
- endAtStart
- endBeforeEnd
- endBeforeStart
- startAtEnd
- startAtStart
- startBeforeEnd
- startBeforeStart

# Constraints on groups of interval variables

Describes constraints that act to encapsulate a group of intervals together.

The main purpose of a group constraint is to encapsulate a group of interval variables into one effective higher level interval.

Three "interval grouping" constraints are available: span, alternative, and synchronize.

**Span constraint**: The constraint  $span(a, \{b_1, ..., b_n\})$  states that the interval a spans over all present intervals from the set  $\{b_1, ..., b_n\}$ . That is, interval a starts together with the first present interval from  $\{b_1, ..., b_n\}$  and ends together with the last one.

**Alternative constraint:** The constraint *alternative*(a, { $b_1$ , ...,  $b_n$ }) models an exclusive alternative between { $b_1$ , ...,  $b_n$ }. If interval a is present then exactly one of intervals { $b_1$ , ...,  $b_n$ } is present and a starts and ends together with this chosen one. The alternative constraint can also be specified by a non-negative integer cardinality c,  $alternative(a, \{b_1, ..., b_n\}, c)$ . In this case, it is not 1 but c interval variables that will be selected among the set { $b_1$ , ...,  $b_n$ } and those c selected intervals will have to start and end together with interval variable a.

**Synchronize constraint**: The constraint *synchronize*(a, { $b_1$ , ...,  $b_n$ }) makes intervals  $b_1$  . . .  $b_n$  start and end together with interval a (if a is present).

More formally, let  $\underline{a},\underline{b_1},...,\underline{b_i},...,\underline{b_n}$  be a set of fixed intervals and  $\underline{c}$  a fixed integer. The **span constraint**  $span(\underline{a},\{\underline{b_1},...,\underline{b_i},...,\underline{b_n}\})$  holds if and only if:

$$\begin{aligned} \neg x(\underline{a}) &\Leftrightarrow & \forall i \in [1, n], \neg x(\underline{b_i}) \\ x(\underline{a}) &\Leftrightarrow & \begin{cases} \exists i \in [1, n], x(\underline{b_i}) \\ s(\underline{a}) = \min_{i \in [1, n], x(\underline{b_i})} s(\underline{b_i}) \\ e(\underline{a}) = \max_{i \in [1, n], x(\underline{b_i})} e(\underline{b_i}) \end{cases}$$

The alternative interval constraint  $alternative(\underline{a}, \{\underline{b_1}, ..., \underline{b_i}, ..., \underline{b_n}\}, \underline{c})$  holds if and only if

When parameter c is ommitted, we assume c=1. Note that when  $c\leq 0$ , as there cannot be any subset  $K\subset [1,n]$  such that  $K\neq\emptyset$  and |K|=c, it means that necessarily, interval a is absent. And the contrapositive: if interval a is present, then necessarily, for the constraint to be satisfied it must be that  $c\geq 1$ .

The synchronization constraint  $synchronize(\underline{a}, \{\underline{b_1},...,\underline{b_i},...,\underline{b_n}\})$  holds if and only if:  $x(\underline{a}) \Rightarrow \forall i \in [1,n] \, | \, x(\underline{b_i}), \, \big(s(\underline{a}) = s(\underline{b_i})\big) \land \big(e(\underline{a}) = e(\underline{b_i})\big)$ 

Note that the alternative, span, and synchronize constraints cannot be used in meta-constraints.

For syntax and examples, see the functions as described in the *OPL Language Quick Reference*.

- alternative
- span
- synchronize

## A logical constraint between interval variables: presenceOf

The presence constraint states that a certain interval must be present in the solution.

The semantics of the presence constraint on a fixed interval  $\underline{a}$  is simply:  $presenceOf(\underline{a}) \Rightarrow x(\underline{a})$ . The truth value of this constraint can be used in arithmetical expressions, and thereby restricted by logical constraints.

This constraint can be used in meta-constraints to indicate, for example, that there may be two optional intervals a and b; if interval a is present then b must be present as well. This is modelled by the constraint  $presenceOf(a) \rightarrow presenceOf(b)$ .

#### **OPL** formulation

The constraint to express that the interval variable must be present:

```
presenceOf(a);
where:
dvar interval a;
```

For an example of the presenceOf constraint see presenceOf in the OPL Language Quick Reference.

## **Expressions on interval variables**

Integer and numerical expressions are available to access or evaluate different attributes of an interval variable.

These expressions can be used, for example, to define a term for the cost function or to connect interval variables to integer and floating point expressions.

The integer expressions are startOf, endOf, lengthOf, and sizeOf and they provide access to the different attributes of an interval variable. Special care must be taken for optional intervals, as an integer value dval must be specified which represents the value of the expression when the interval is absent. If this value is omitted, it is supposed to be 0. For the syntax of integer expressions, see endOf, lengthOf, sizeOf, and startOf in the OPL Language Quick Reference.

The numerical expressions are startEval, endEval, lengthEval, and sizeEval, and they allow evaluation of a piecewise linear function on a given bound of an interval. As with integer expressions, in the case of optional intervals an integer value dval must be specified which represents the value of the expression when the interval is absent. If this value is omitted, it is supposed to be 0. For the syntax and examples of the use of a numerical expression, see endEval, lengthEval, sizeEval, and startEval in the OPL Language Quick Reference.

Let <u>a</u> denote a fixed interval variable. The semantics of these expressions is shown in the table.

Expression	Semantics
$\mathit{startOf}(\underline{a}, dval)$	$expr = \begin{cases} s(\underline{a}) & \text{if } x(\underline{a}) \\ dval & \text{otherwise} \end{cases}$
$endOf(\underline{a}, dval)$	$expr = \begin{cases} e(\underline{a}) & \text{if } x(\underline{a}) \\ dval & \text{otherwise} \end{cases}$
$lengthOf(\underline{a}, dval)$	$expr = \begin{cases} l(\underline{a}) & \text{if } x(\underline{a}) \\ dval & \text{otherwise} \end{cases}$
$\mathit{sizeOf}(\underline{a}, dval)$	$expr = \begin{cases} sz(\underline{a}) & \text{if } x(\underline{a}) \\ dval & \text{otherwise} \end{cases}$
$startEval(\underline{a}, F, dval)$	$expr = \begin{cases} F(s(\underline{a})) & \text{if } x(\underline{a}) \\ dval & \text{otherwise} \end{cases}$
$endEval(\underline{a},F,dval)$	$expr = \begin{cases} F(e(\underline{a})) & \text{if } x(\underline{a}) \\ dval & \text{otherwise} \end{cases}$
$lengthEval(\underline{a}, F, dval)$	$expr = \begin{cases} F(l(\underline{a})) & \text{if } x(\underline{a}) \\ dval & \text{otherwise} \end{cases}$
$\mathit{sizeEval}(\underline{a}, F, dval)$	$expr = \begin{cases} F(sz(\underline{a})) & \text{if } x(\underline{a}) \\ dval & \text{otherwise} \end{cases}$

Expression semantics

## Sequencing of interval variables

Describes a basic building block of scheduling, the interval sequence.

 $\subseteq \pi \forall$ 

An *interval sequence variable* is defined on a set of interval variables *A*. Informally speaking, the value of an interval sequence variable represents a total ordering of the interval variables of *A*. Note that any absent interval variables are not considered in the ordering.

More formally, an interval sequence variable p on a set interval variables A represents a decision variable whose possible values are all the permutations of the intervals of A. Let  $\underline{A}$  be a set of fixed intervals and n denote the cardinality of  $\underline{A}$ .

A permutation  $\pi$  of  $\underline{A}$  is a function  $\pi : \underline{A} \to \{\bot\} \cup [1, n]$  such that if we denote  $length(\pi) = |\{\underline{a} \in \underline{A}, x(\underline{a})\}|$  then the number of present intervals:

- 1.  $\forall \underline{a} \in \underline{A}, x(\underline{a}) \Leftrightarrow (\pi(\underline{a}) \neq \bot)$
- 2.  $\forall \underline{a} \in \underline{A}, x(\underline{a}) \Rightarrow (\pi(\underline{a}) \leq length(\pi))$
- 3.  $\forall \underline{a}, \underline{b} \in \underline{A}, (x(\underline{a}) \land x(\underline{b}) \land \underline{a} \neq \underline{b}) \Rightarrow (\pi(\underline{a}) \neq \pi(\underline{b}))$

Note that the sequence alone does not enforce any constraint on the relative position of intervals end-points. For instance, an interval variable a could be sequenced before an interval variable b in a sequence p without any impact on the relative position between the start/end points of a and b (a could still be fixed to start after the end of b). This is because different semantics can be used to define

how a sequence constrains the positions of intervals. We will see later how the *noOverlap* constraint implements one of these possible semantics.

The sequence variable also allows associating a fixed non-negative integer *type* with each interval variable in the sequence. In particular, these integers are used by the *noOverlap* constraint. T(p, a) denotes the fixed non-negative integer type of interval variable a in the sequence variable p.

### Constraints on sequence variables

The following constraints are available on sequence variables:

- first (scheduling)(p,a) states that if interval a is present, then it will be the first interval of the sequence p.
- last (scheduling)(p,a) states that if interval a is present, then it will be the last interval of the sequence p.
- before(*p*, *a*, *b*) states that if both intervals *a* and *b* are present then *a* will appear before *b* in the sequence *p*.
- prev (scheduling)(p, a, b) states that if both intervals a and b are present then a will be just before b in the sequence p, that is, it will appear before b and no other interval will be sequenced between a and b in the sequence.

The formal semantics of these basic constraints is shown in the following table.

Relation	Semantics
$first(\pi, \underline{a})$	$x(\underline{a}) \Rightarrow (\pi(\underline{a}) = 1)$
$last(\pi, \underline{a})$	$x(\underline{a}) \Rightarrow (\pi(\underline{a}) = length(s))$
$before(\pi, \underline{a}, \underline{b})$	$(x(\underline{a}) \land x(\underline{b})) \Rightarrow (\pi(\underline{a}) \leq \pi(\underline{b}))$
$prev(\pi, \underline{a}, \underline{b})$	$(x(\underline{a}) \land x(\underline{b})) \Rightarrow (\pi(\underline{b}) = \pi(\underline{a}) + 1)$

Sequence relation semantics

#### The no overlap constraint

The *no overlap* constraint on an interval sequence variable *p* states that the sequence defines a chain of non-overlapping intervals, any interval in the chain being constrained to end before the start of the next interval in the chain. This constraint is typically useful for modelling disjunctive resources.

More formally, the condition for a permutation value  $\pi$  for sequence p to satisfy the noOverlap constraints is defined as:

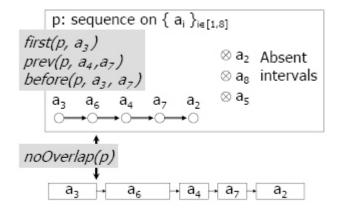
$$\begin{array}{ll} noOverlap(\pi) & \Leftrightarrow & \forall \underline{a},\underline{b} \in \underline{A}, \neg x(\underline{a}) \vee \neg x(\underline{b}) \vee \\ & \left(\left(\pi(\underline{a}) < \pi(\underline{b})\right) \Leftrightarrow \left(e(\underline{a}) \leq s(\underline{b})\right)\right) \end{array}$$

If a transition distance matrix M is specified, it defines the minimal non-negative distance that must separate two consecutive intervals in the sequence.

More formally, if T(p,a) denotes the non-negative integer type of interval a in the sequence variable p:

$$noOverlap(\pi, M) \Leftrightarrow \forall \underline{a}, \underline{b} \in \underline{A}, \neg x(\underline{a}) \lor \neg x(\underline{b}) \lor ((\pi(\underline{a}) < \pi(\underline{b})) \Leftrightarrow (e(\underline{a}) + M[T(\pi, \underline{a}), T(\pi, \underline{b})] \le s(\underline{b})))$$

A sequence variable together with a no-overlap constraint using it are illustrated in this figure:



Example of sequence variable and no-overlap constraint

### The sameSequence and sameCommonSubsequence constraints

The sameSequence and sameCommonSubsequence constraints are binary constraints on a pair of sequence variables  $p_1$  and  $p_2$ . These constraints state that the relative position of some related subsets of interval variables is the same in both sequences. Typically, if a is before b in sequence  $p_1$  then the interval related to a is before the interval related to b in sequence b. Here are two examples of use-cases for these constraints.

- First in/first out and no-bypass constraints. In some physical systems, like trains on a single line railway, or items on a conveyor belt, bypassing is not possible and items must enter and exit a given section of the system in the same order. If entering and exiting the sections or the junctions of the system is modeled by two related interval variables, those constraints can be modeled by sameSequence constraints (or sameCommonSubsequence constraints if the items do not follow the same path) on the sequences of entering and exiting intervals. A classic example of such a constraint is the permutation flow-shop scheduling problem.
- Scenario-based approaches for scheduling with uncertainties. In presence of uncertainties (for instance, uncertain activity durations) one may be interested in building sequences of activities on unary resources that optimize some robustness or statistical criterion (for instance the expected makespan). A scenario is a sub-model that defines a particular realization of the uncertainties in the environment. As one must find robust sequences that optimize some criterion over all scenarios, the different sequences of a given unary resource across all scenarios are linked with sameSequence constraints.

A constraint  $sameCommonSubsequence(p_1, p_2, B_1, B_2)$  is defined on two sequence variables  $p_1$  (defined on a set of interval variables  $A_1$ ) and  $p_2$  (defined on a set of interval variables  $A_2$ ), with two arrays of interval variables  $B_1$  and  $B_2$  of identical size and such that  $B_1 \subset A_1$ ,  $B_2 \subset A_2$ . The constraint states that the sub-sequences

defined by  $p_1$  and  $p_2$  by only considering the pairs of present intervals ( $B_1[i]$ ,  $B_2[i]$ ) are identical modulo the mapping between intervals  $B_1[i]$  and  $B_2[i]$ .

For instance, let's suppose sequence variable  $p_1$  defined on interval variables  $\{a,b,c,d,e,f\}$  and sequence variable  $p_2$  defined on interval variables  $\{u,v,w,x,y,z\}$  with a constraint  $sameCommonSubsequence(p_1, p_2, [a,c,d,e,f], [u,w,v,x,y])$ . Assuming that, in a solution, interval variables d and y are absent while all the other interval variables are present. The pair of permutations  $\pi_1$ =(c,f,a,e,b) and  $\pi_2$ =(w,v,u,x) satisfies the sameCommonSubsequence constraints. Indeed if one considers only pairs of present intervals, the mapping  $B_1$ ,  $B_2$  maps [a,c,e] with [u,w,x]. Sub-sequence of  $\pi_1$  over  $\{a,c,e\}$  is (c,a,e), sub-sequence of  $\pi_2$  over  $\{u,w,x\}$  is (w,u,x) and one can see that the two sub-sequences are the same modulo the mapping.

More formally, let n denote the size of both interval variable arrays  $B_1$  and  $B_2$ . The condition for permutations  $\pi_1$  and  $\pi_2$  of sequencing variables  $p_1$  and  $p_2$  to satisfy the constraint  $sameCommonSubsequence(\pi_1, \pi_2, \underline{B_1}, \underline{B_2})$  is defined as:

$$\forall i,j \in [1,n], \ i \neq j, \ x(\underline{B_1}[i]), \ x(\underline{B_2}[i]), \ x(\underline{B_2}[i]), \ x(\underline{B_2}[i]), \ (\pi_1(\underline{B_1}[i]) < \pi_1(\underline{B_1}[i]) < \pi_2(\underline{B_2}[i]) < \pi_2(\underline{B_2}[i])$$

A constraint  $sameSequence(p_1, p_2, B_1, B_2)$  is defined on two sequence variables of identical size  $p_1$  (defined on a set of interval variables  $A_1$ ) and  $p_2$  (defined on a set of interval variables  $A_2$ ) with two arrays of interval variables  $B_1$  and  $B_2$  that are an ordering of sets  $A_1$  and  $A_2$ . The constraint states that sequences  $p_1$  and  $p_2$  are identical modulo a mapping between intervals  $B_1[i]$  and  $B_2[i]$ . A sameSequence constraint is stronger than a sameCommonSubsequence constraint; it additionally imposes that arrays  $B_1$  and  $B_2$  contain all the definition intervals of their relative sequence variables and that the presence status of mapped interval variables  $B_1[i]$  and  $B_2[i]$  is the same.

More formally, let n denote the size of both sequence variables. The condition for permutations  $\pi_1$  and  $\pi_2$  of the sequence variables  $p_1$  and  $p_2$  to satisfy constraint  $sameSequence(\pi_1, \pi_2, \underline{B_1}, \underline{B_2})$  is defined as:

$$\forall i \in [1, n], \, \pi_1(\underline{B_1}[i]) = \pi_2(\underline{B_2}[i])$$

When parameters  $B_1$  and  $B_2$  are omitted in the definition of the above constraints, we assume  $B_1$  and  $B_2$  are the definition intervals of  $p_1$  and  $p_2$ , following the definition order of interval variables in the sequence.

#### Syntax and examples

For syntax and examples of use of the sequence interval variable, see sequence in the *OPL Language Quick Reference*.

For the syntax and examples of use of the no overlap constraint, which needs to be defined as a set of integer triples, see noOverlap in the *OPL Language Quick Reference*.

For the syntax and examples of the other constraints available on an interval sequence variable, see first, last (scheduling), prev (scheduling), before, sameSequence, and sameCommonSubsequence in the *OPL Language Quick Reference*. (Note that there are similarly-named constraints available for set operations in OPL.)

Note that none of the constraints mentioned in this section can be used in a meta-constraint.

Sequence variables can be viewed in a Gantt chart in the CPLEX Studio IDE. A simple example is given in Viewing the results of scheduling problems.

## **Cumulative functions**

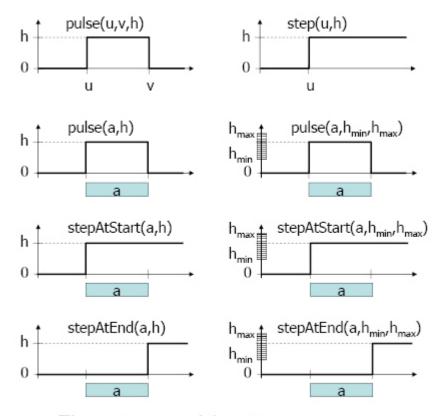
Describes the cumulative function.

In scheduling problems involving cumulative resources (also known as renewable resources), the cumulated usage of the resource by the activities is usually represented by a function of time. An activity usually increases the cumulated resource usage function at its start time and decreases it when it releases the resource at its end time (pulse function). For resources that can be produced and consumed by activities (for instance the content of an inventory or a tank), the resource level can also be described as a function of time; production activities will increase the resource level whereas consuming activities will decrease it. In these type of problems, the cumulated contribution of activities on the resource can be represented by a function of time and constraints can be modeled on this function, for instance a maximal or a safety level.

CP Optimizer introduces the notion of the *cumulative function expression*, which is a function that represents the sum of individual contributions of intervals. A panel of elementary cumul function expressions is available to describe the individual contribution of an interval variable (or a fixed interval of time) which cover the main use-cases mentioned previously: *pulse* for usage of a cumulative resource, *step* for resource production/consumption. When the elementary cumulative function expressions that define a cumul function expression are fixed (and thus, so are their related intervals), the expression is fixed. CP Optimizer provides several constraints over cumul function expressions. These constraints allow restricting the possible values of the function over the complete horizon or over some fixed or variable interval. For applications where the actual quantity of resource that is used, produced or consumed by intervals is an unknown of the problem, expressions are available for constraining these quantities.

### **Cumul function expressions**

Let  $\mathcal{F}^+$  denote the set of all functions from  $\mathbb{Z}$  to  $\mathbb{Z}^+$ . A cumul function expression f is an expression whose value is a function of  $\mathcal{F}^+$  and thus, whose domain dom(f) is a subset of  $\mathcal{F}^+$ . Let  $u, v \in \mathbb{Z}$  and  $h, h_{min}, h_{max} \in \mathbb{Z}^+$  and a be an interval variable, we consider the following elementary cumul function expressions illustrated in the following figure: pulse(u, v, h), step(u, h), pulse(a, h),  $pulse(a, h_{min}, h_{max})$ , stepAtStart(a, h),  $stepAtStart(a, h_{min}, h_{max})$ , stepAtEnd(a, h), and  $stepAtEnd(a, h_{min}, h_{max})$ .



Elementary cumul function expressions

More formally, let  $u,v\in\mathbb{Z}$  and  $h\in\mathbb{Z}^+$  and we define the following particular functions of  $\mathcal{F}^+$ 

- 0 is the null function that is, the function F such that ∀t ∈ Z, F(t) = 0
- pulse(u, v, h) is the function F such that F(t) = h if t ∈ [u, v), F(t) = 0 otherwise
- $step(u, h) = pulse(u, +\infty, h)$

The semantics of the elementary function expressions is listed in the following table, together with the formal definition of their domain. The function set  $0_a$  is equal to the singleton  $\{0\}$  if  $\bot \in dom(a)$ ; that is, if interval variable a is possibly absent, and equal to the empty set otherwise.

Function expression	Domain
pulse(u, v, h)	$\{ pulse(u, v, h) \} $ (singleton)
step(u,h)	$\{\underline{step}(u,h)\}$ (singleton)
pulse(a, h)	$0_a \cup \{ pulse(s(\underline{a}), e(\underline{a}), h)   \underline{a} \in dom(a) \setminus \bot \}$
$pulse(a, h_{min}, h_{max})$	$0_a \cup \{ \underline{pulse}(s(\underline{a}), e(\underline{a}), h) \mid h \in [h_{min}, h_{max}] \land \underline{a} \in dom(a) \setminus \bot \}$
stepAtStart(a, h)	$0_a \cup \{ \underline{step}(s(\underline{a}), h)   \underline{a} \in dom(a) \setminus \bot \}$
$stepAtStart(a, h_{min}, h_{max})$	$0_a \cup \{ \underline{step}(s(\underline{a}), h)   h \in [h_{min}, h_{max}] \land \underline{a} \in dom(a) \setminus \bot \}$
stepAtEnd(a, h)	$0_a \cup \{ \ step(e(\underline{a}), h) \   \underline{a} \in dom(a) \setminus \bot \ \}$
$stepAtEnd(a, h_{min}, h_{max})$	$0_a \cup \{ step(e(\underline{a}), h)   h \in [h_{min}, h_{max}] \land \underline{a} \in dom(a) \setminus \bot \}$

Elementary cumul function expressions

A cumul function expression f is an expression built as the algebraical sum of the elementary function expressions in the table (or negations). More formally, it is a construct of the form  $f = \sum_i \epsilon_i \cdot f_i$  where  $\epsilon_i \in \{-1, +1\}$  and  $f_i$  is an elementary function expression.

When all elementary function expressions  $f_i$  related with a cumul function expression f are fixed (that is, their domain is reduced to a singleton  $\underline{f}_i$ ), the value of the cumul function expression f is fixed and equal to the function  $\underline{f} = \sum_i \epsilon_i \cdot \underline{f}_i$ .

### **Constraints on cumul function expressions**

The following constraints can be expressed on a cumul function expression f. Let  $u, v \in \mathbb{Z}$  and  $h, h_{min}, h_{max} \in \mathbb{Z}^+$  and a be an interval variable:

- $alwaysIn(f, u, v, h_{min}, h_{max})$  means that the values of function f must remain in the range  $[h_{min}, h_{max}]$  everywhere on the interval [u, v).
- $alwaysIn(f, a, h_{min}, h_{max})$  means that if interval a is present, the values of function f must remain in the range  $[h_{min}, h_{max}]$  between the start and the end of interval variable a.
- $f \le h_{\max}$  means that function f cannot take values greater than  $h_{\max}$ . It is semantically equivalent to  $alwaysIn(f, -\infty, +\infty, 0, h_{\max})$ .
- $f \ge h_{\min}$  means that function f cannot take values lower than  $h_{\min}$ . It is semantically equivalent to  $alwaysIn(f, -\infty, +\infty, h_{\min}, +\infty)$

More formally:

$$alwaysIn(\underline{f}, u, v, h_{min}, h_{max}) \Leftrightarrow \forall t \in [u, v), \ \underline{f}(t) \in [h_{min}, h_{max}]$$

$$alwaysIn(\underline{f}, \underline{a}, h_{min}, h_{max}) \Leftrightarrow \forall t \in [s(\underline{a}), e(\underline{a})), \ f(t) \in [h_{min}, h_{max}]$$

## **Expressions on cumulative functions**

The following elementary cumul function expressions are based on an interval variable a: pulse(a, h),  $pulse(a, h_{\min}, h_{\max})$ , stepAtStart(a, h),  $stepAtStart(a, h_{\min}, h_{\max})$ , stepAtEnd(a, h), and  $stepAtEnd(a, h_{\min}, h_{\max})$ .

Some of these expressions define a range  $[h_{\min}, h_{\max}]$  of possible values for the actual height of the function when the interval variable a is present. The actual height is an unknown of the problem. CP Optimizer provides some integer expressions to control this height. These expressions are based on the notion of the contribution of a given interval variable a to the (possibly composite) cumul function expression f. This contribution is defined as the sum of all the elementary

cumul function expressions based on a in f. This contribution is a discrete function that can change value only at the start and at the end of interval a and is equal to 0 before the start of a.

For instance, let a and b be two interval variables and a cumul function expression f defined by: f = pulse(a, 3) + pulse(a, 2) - stepAtEnd(a, 1) + stepAtStart(b, 2) - stepAtEnd(b, 3). The contribution of a to f is the function pulse(a, 3) + pulse(a, 2) - stepAtEnd(a, 1) and the contribution of b to f is the function stepAtStart(b, 2) - stepAtEnd(b, 3).

If interval a is present, the expression heightAtStart(a, f) returns the value of the contribution of a to f evaluated at the start of a that is, it measures the contribution of interval a to cumul function expression f at its start point. Similarly, the expression heightAtEnd(a, f) returns the value of the contribution of a to f evaluated at the end of a that is, it measures the contribution of interval a to cumul function expression f at its end point. An additional integer value dval can be specified at the construction of the expression, which will be the value returned by the expression when the interval is absent. Oherwise, if no value is specified, the expression will be equal to 0 when the interval is absent.

In the previous example, assuming both interval a and b to be present we would get: heightAtStart(a, f) = 5, heightAtEnd(a, f) = 4, heightAtStart(b, f) = 2, heightAtEnd(b, f) = -1. Of course, in general when using ranges for the height of elementary cumul function expressions, the heightAtStart/End expressions will not be fixed until all the heights have been fixed by the search.

More formally, if an elementary cumul function expression  $f_i$  is based on an interval variable, we denote  $ivar(f_i)$  this interval variable. Let f be a cumul function expression defined as  $f = \sum_i \epsilon_i \cdot f_i$  where  $\epsilon_i \in \{-1, +1\}$  where  $f_i$  are elementary cumul function expressions. The contribution of an interval variable a to f is defined as  $f|_a = \sum_{i \mid ivar(f_i) = a} \epsilon_i \cdot f_i$  and the expressions heightAtStart/End are defined as follows:

$$\bullet \ \ heightAtStart(\underline{a},\underline{f},dval) = \left\{ \begin{array}{ll} \underline{f}|_{\underline{a}}(s(\underline{a})) & \text{if } x(\underline{a}) \\ \overline{dval} & \text{otherwise} \end{array} \right.$$

$$\bullet \ \ heightAtEnd(\underline{a},\underline{f},dval) = \left\{ \begin{array}{ll} \underline{f|_{\underline{a}}(e(\underline{a}))} & \text{if } x(\underline{a}) \\ dval & \text{otherwise} \end{array} \right.$$

## Syntax and examples

For the syntax and examples of use of a cumulative function see cumulFunction, pulse, step, stepAtEnd, and stepAtStart in the *OPL Language Quick Reference*.

The results of a solved scheduling model can be viewed in the CPLEX Studio IDE. An example of results containing a cumulative function is provided in Viewing the results of cumulative functions in the IDE.

These are the constraints available on cumulative function expressions:

- f <= hmax;
- hmin <= f:
- alwaysIn(f, u, v, hmin, hmax)
- alwaysIn(f, a, hmin, hmax)

Note that these constraints cannot be used in meta-constraints.

The following expressions are available on cumulative functions:

- dexpr int h = heightAtStart(a,f[,dval]);
- dexpr int h =heightAtEnd(a,f[,dval]);

More information on these constraints and expressions is available in the OPL Language Quick Reference.

### State functions

Describes the state function.

Some scheduling problems involve reasoning with resources whose state may change over time. The state of the resource can change because of the scheduled activities or because of exogenous events; yet some activities in the schedule may need a particular condition on the resource state to be true in order to execute. For instance, the temperature of an oven may change due to an activity that sets the oven temperature to a value v, and a cooking activity may follow that requires the oven temperature to start at and maintain a temperature level v' throughout its execution. Furthermore, the transition between two states is not always instantaneous and a transition time may be needed for the resource to switch from a state v to a state v'.

CP Optimizer introduces the notion of state function which is used to describe the evolution of a given feature of the environment. The possible evolution of this feature is constrained by interval variables of the problem. The main difference between state functions and cumulative functions is that interval variables have an incremental effect on cumul functions (increasing or decreasing the function value) whereas they have an absolute effect on state functions (requiring the function value to be equal to a particular state or in a set of possible states).

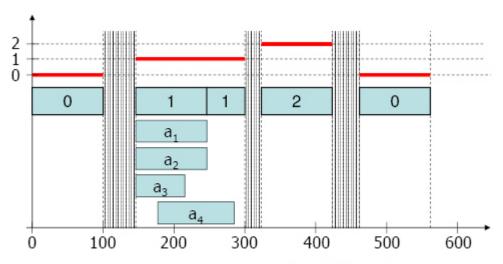
Informally speaking, a state function is a set of non-overlapping intervals over which the function maintains a particular non-negative integer state. In between those intervals, the state of the function is not defined, typically because of an ongoing transition between two states. For instance for an oven with three possible temperature levels identified by indexes 0, 1 and 2 we could have:

- [start=0, end=100): state=0,
- [start=150, end=250): state=1,
- [start=250, end=300): state=1,
- [start=320, end=420): state=2,
- [start=460, end=560): state=0, ...

Constraints are available to restrict the evolution of a state function. These constraints allow you to specify:

- That the state of the function must be defined and should remain equal to a given non-negative state everywhere over a given fixed or variable interval (alwaysEqual).
- That the state of the function must be defined and should remain constant (no matter its value) everywhere over a given fixed or variable interval (alwaysConstant).
- That intervals requiring the state of the function to be defined cannot overlap a given fixed or variable interval (alwaysNoState).
- That everywhere over a given fixed or variable interval, the state of the function, if defined, must remain within a given range of non-negative states  $[v_{\min}, v_{\max}]$ (alwaysIn).

Additionally, the two first constraints can be complemented to specify that the given fixed or variable interval should have its start and/or end point synchronized with the start and/or end point of the interval of the state function that maintains the required state. This is the notion of *start* and *end alignment* which is particularly useful for modelling parallel batches. For instance in the oven example above, all interval variables that would require an oven temperature of level 1 and specify a start and end alignment, if executed over the interval [150, 250) would have to start exactly at 150 and end at 250. This is depicted in the following figure where  $a_1$  and  $a_2$  are two start and end aligned interval variables,  $a_3$  is start aligned only and  $a_4$  is not aligned at all.



State function and interval variable alignments

#### State functions and transition distance

A state function f is a decision variable whose value is a set of non-overlapping intervals, each interval  $[s_i, e_i]$  being associated a non-negative integer value  $v_i$  that represents the state of the function over the interval.

Let f be a fixed state function, we will denote  $\underline{f} = (\ [s_i, e_i) : v_i\ )_{i \in [1, n]} :$ 

- $\bullet \ \forall i \in [1,n], s_i \in \mathbb{Z}, e_i \in \mathbb{Z}, v_i \in \mathbb{Z}^+, s_i < e_i$
- $\forall i \in [1, n-1], e_i < s_{i+1}$

We denote  $D(\underline{f}) = \bigcup_{i \in [1,n]} [s_i, e_i)$  the definition domain of  $\underline{f}$ , that is, the set of points where the state function is associated a state.

For a fixed state function  $\underline{f}$  and a point  $t \in D(\underline{f})$ , we will denote  $[s(\underline{f},t),e(\underline{f},t))$  the unique interval of the function that contains t and f(t) the value of this interval:

$$\begin{split} [s(\underline{f},t),e(\underline{f},t)) &= [s_i,e_i), t \in [s_i,e_i) \\ &\underline{f}(t) = v_i, t \in [s_i,e_i) \end{split}$$

For instance, in the example of the oven introduced previously, we would have  $\underline{f}$  (200) = 1,  $s(\underline{f}$  , 200) = 150 and e(f, 200) = 250.

A state function can be associated with a *transition distance*. The transition distance defines the minimal distance that must separate two consecutive states in the state function. More formally, if M[v, v'] represents a transition distance matrix between state v and state v', the transition distance means that:

$$\forall i \in [1, n-1], e_i + M[v_i, v_{i+1}] \le s_{i+1}$$

The transition distance matrix *M* must satisfy the triangular inequality. For instance, in the example of the oven, the state function depicted in the previous figure is consistent with the following transition distance:

$$\mathbf{M} = \left(\begin{array}{ccc} 0 & 50 & 60 \\ 20 & 0 & 20 \\ 40 & 30 & 10 \end{array}\right)$$

#### Constraints on state functions

If f is a state function, a an interval variable, v,  $v_{\min}$ ,  $v_{\max}$  non-negative integers and  $algn_s$ ,  $algn_e$  two boolean values:

• The constraint  $alwaysEqual(f, a, v, algn_s, algn_e)$  specifies that whenever a is present, state function f must be defined everywhere between the start and the end of interval a and be constant and equal to non-negative value v over this interval. If  $algn_s$  is true, it means that interval a is start-aligned with f: Interval a must start at the beginning of the interval where f is maintained in state s. If  $algn_e$  is true, it means that interval a is end-aligned with f: Interval a must end at the end of the interval where f is maintained in state s. More formally:

$$- \forall t \in [s(\underline{a}), e(\underline{a})), t \in D(\underline{f}) \land \underline{f}(t) = v$$

$$- algn_s \Rightarrow s(\underline{a}) = s(\underline{f}, s(\underline{a}))$$

$$- algn_e \Rightarrow e(\underline{a}) = e(\underline{f}, e(\underline{a}))$$

- The constraint alwaysNoState(f, a) specifies that whenever a is present, state function f cannot provide any valid state between the start and the end of interval a. As a consequence, any interval constrained with alwaysEqual or alwaysConstant on this function and thus requiring the function to be defined cannot overlap interval a. Formally:  $[s(\underline{a}), e(\underline{a})) \cap D(\underline{f}) = \emptyset$
- The constraint  $alwaysIn(f, a, v_{\min}, v_{\max})$  specifies that whenever a is present, everywhere between the start and the end of interval a the state of function f, if defined, must belong to the range  $[v_{\min}, v_{\max}]$  where  $0 \le V_{\min} \le V_{\max}$ . Formally:  $\forall t \in [s(\underline{a}), e(\underline{a})) \cap D(\underline{f}), \underline{f}(t) \in [v_{\min}, v_{\max}]$

### Syntax and examples

For the syntax and examples of use of a state function see stateFunction.

The results of a solved scheduling model can be viewed in the CPLEX Studio IDE. An example of results containing a state function is provided in Viewing the results of state functions in a Gantt chart.

Another example featuring a transition and the *alwaysEqual* constraint is shown in the following list.

The following list includes the constraints available on a state function. A full description and example for each syntax is available in the *OPL Language Quick Reference*.

```
alwaysEqual(f,s,e,v[,aligns,aligne]);
alwaysEqual(f,a,v[,aligns,aligne]);
alwaysConstant(f,s,e,[,aligns,aligne]);
alwaysConstant(f,a,[,aligns,aligne]);
alwaysNoState(f,s,e);
alwaysNoState(f,a);
alwaysIn(f,u,v,hmin,hmax);
alwaysIn(f,a,hmin,hmax);
```

Note that these constraints cannot be used in meta-constraints.

Example with stateFunction, transition, and alwaysEqual.

A machine can be equipped with a tool among a set of n possible tools. Each operation o executed on the machine needs a specific tool RequiredTool[o]. The machine can process several operation simultaneously provided these operations are compatible with the tool currently installed on the machine. Changing the tool installed on the machine needs some constant set-up time which is supposed to be independent from the tools.

```
int nbTools = ...;
int nbOps = ...;
int setupTime = ...;
range Tools = 1..nbTools;
range Operations = 1..nbOps;
int Duration [Operations] = ...;
int RequiredTool [Operations] = ...;

dvar interval op[o in Operations] size Duration[o];

tuple triplet { int tl1; int tl2; int value; };
{ triplet } Transition = { <tl1,tl2,setupTIme> } tl1, tl2 in Tools };
stateFunction machineTool with Transition;

constraints {
  forall(o in Operations) {
    alwaysEqual(machineTool, op[o], RequiredTool[o]);
  }
}
```

### **Notation**

The main types of notation used throughout the scheduling section are defined here.

Vectors are denoted by capital letters, for example *Y*. The size of a vector *Y* is denoted |Y|. If n = |Y|, the vector is denoted  $Y = (y_1,...,y_n)$ .

Notation	Туре	Description
a, b	interval variable	
x(a)	boolean expression	execution status of a
s(a), e(a), l(a)	integer expression	start, end, length of $a$
sz(a)	integer expression	size of a
A, B	vector of interval variables	
p	interval sequence variable	
f, g, h	cumul function expression	
W C 9786C)	state function	
$\underline{a}, \underline{b}$	fixed interval	
$x(\underline{a})$	boolean	presence status of $\underline{a}$
$s(\underline{a}), e(\underline{a}), l(\underline{a})$	integer	start, end, length of $\underline{a}$
$sz(\underline{a})$	integer	size of <u>a</u>
$\pi$	interval variable permutation	
z	integer (expression)	delay of a precedence
t	integer	date
T	vector of integers	array of dates
S	vector of floats	array of slopes
v, w	integer or float	miscellaneous integer or float value
V, W	vector of integers or floats	vector of miscellaneous values
align	boolean	alignment flag
F,G,H	piecewise linear function $\mathbb{R} \to \mathbb{R}$	
	vector of cumul function expressions	
T(p, a)	integer	type of interval $a$ in sequence $p$
M	integer matrix	transition distance matrix
$\mathcal{F}^+$	set of functions	set of all functions $f : \mathbb{Z} \to \mathbb{Z}^+$

# Chapter 3. IBM ILOG Script for OPL

Describes the structure and built-in values and functions of the scripting language.

## Language structure

Presents the structure of the IBM ILOG Script language for OPL: the language constructs, the elements from which expressions can be constructed, and the possible types of statement.

## **Syntax**

What composes a scripting statement, compound statements, comments, identifiers.

#### General

Provides a general overview of OPL syntax.

A script comprises a sequence of statements. An expression can also be used whenever a statement is expected, in which case its value is ignored and only its side effect is taken into account.

You can put multiple statements or expressions on a single line if you separate them with a semi-colon (;), for example, the following two scripts are equivalent:

## Script1

```
writeln("Hello, world")
x = x+1
if (x > 10) writeln("Too big")
```

#### Script2

```
writeln("Hello, world"); x = x+1; if (x > 10) writeln("Too big")
```

#### Compound statements

Explains the use of compound statements in OPL syntax.

A compound statement is a sequence of statements and expressions enclosed in curly brackets ({}). It can be used to perform multiple tasks whenever a single statement is expected, for example, in the following conditional statement, the three statements and expressions in curly brackets are executed when the condition a > b is true:

```
if (a > b) {
  var c = a
   a = b
   b = c
}
```

The last statement or expression before a closing curly bracket does not need to be followed by a semicolon, even if it is on the same line. For example, the following program is syntactically correct and is equivalent to the previous one:

```
if (a > b) { var c = a; a = b; b = c }
```

### **Comments**

Explains the syntax of comments in OPL.

Script supports two different styles of comments:

• Single line comments: A single line comment starts with // and stops at the end of the line.

#### Example:

```
x = x+1 // Increment x,

y = y-1 // then decrement y.
```

• Multiple line comments: To span on more than one line, comments must start with a /\* and ends with a \*/; Nested multiple line comments are not allowed.

#### Example:

```
/* The following statement
increments x. */
x = x+1
/* The following statement
decrements y. */
y = y /* A comment can be inserted here */ -1
```

#### **Identifiers**

Shows how to use identifiers in OPL syntax.

Identifiers are used to name script variables and functions. An identifier starts with either a letter or an underscore, and is followed by a sequence of letters, digits, and underscores.

These are examples of identifiers:

```
car
x12
main_window
_foo
```

The language is case-sensitive, so that the uppercase letters A-Z are distinct from the lowercase letters a-z. For example, the identifiers car and Car are distinct.

# **Expressions in IBM ILOG Script**

Expressions are a combination of literals, script variables, special keywords, and operators.

#### Literals

Explains the use of literals in IBM ILOG Script.

Literals can represent the following elements:

- Numbers, for example: 12 14.5 1.7e-100
- Strings, for example, "Ford" "Hello world\n"
- Booleans, either true or false
- The null value: null.

For further details about number and string literal syntax, see "Numbers" on page 99 and "IBM ILOG Script strings" on page 106.

#### **Operators**

Explains the use of and precedence of operators in IBM ILOG Script.

The precedence of operators determines the order in which they are applied when an expression is evaluated. You can override operator precedence using parentheses.

For a list of IBM ILOG Script operators, see Operators.

### Syntax of different types of expression

Describes the syntax of several types of expression in OPL.

The following types of expression are described:

- A reference to a script variable
- Access to a property
- Assignment operators
- Function calls
- Special keywords
- Special operators
- Other operators

#### Note:

**For C/C++ programmers:** The syntax of Script expressions is very close to the C and C++ syntax. Expressions include assignments, function calls, property access, and so on.

## Script variable reference

Provides a reference for the syntax of script variables in OPL.

Table 6. Reference syntax of script variables

Syntax	Effect
variable	Returns the value of variable. See "Identifiers" on page 88 for the syntax of script variables. If variable does not exist, an error is signalled. This is not the same as referencing an existing script variable whose value is the undefined value, which is legal and returns the undefined value. When used in the body of a with statement, a variable reference is first looked up as a property of the current default value.

#### Property access

Provides a reference for the syntax used to access properties in OPL.

There are two ways of accessing a property value.

Table 7. Property access syntax

Syntax	Effect
value.name	Returns the value of the name property of value, or the undefined value if this property is not defined. See "Identifiers" on page 88 for the syntax of name.
	Examples:str.length getCar().name
	Because name must be a valid identifier, this form cannot be used to access properties which do not have a valid identifier syntax. For example, the numeric properties of an array cannot be accessed this way:myArray.10 // Illegal syntax  For these properties, use the second syntax.
value[name]	Same as the previous syntax, except that this time name is an evaluated expression which gives the property name.
	Examples:str["length"] // Same as str.length getCar()[getPropertyName()] myArray[10] myArray[i+1]

## **Assignment operators**

Provides a reference for the syntax of assignment operators in OPL.

The equals (=) operator can be used to assign a new value to a script variable or a property.

Table 8. Assignment operator syntax

Syntax	Effect
variable = expression	In scripting, all objects are assigned by reference, except strings, numbers and Booleans, which are assigned by value. See the ECMA standard for details.  Example:x = y+1
	The whole expression returns the value of expression.
value.name = expression value[name] = expression	Assigns the value of expression to the given property.
	If value does not have such a property, then if it is either an array or an object, the property is created; otherwise, an error is signalled.
	Example:car.name = "Ford" myArray[i] = myArray[i]+1
	The whole expression returns the value of expression.

In addition, shorthand operators are also defined.

## **Function calls**

Provides a reference for the shorthand syntax used with function calls in OPL.

Table 9. Syntactic shorthand

Syntax	Shorthand for
++X	X = X+1
X++	Same as ++X, but returns the initial value of X instead of its new value.
X	X = X-1
X	Same asX, but returns the initial value of X instead of its new value.
X += Y	X = X + Y
X -= Y	X = X - Y
X *= Y	X = X * Y
X /= Y	X = X / Y
X %= Y	X = X % Y
Χ <<= Υ	X = X << Y
X >>= Y	X = X >> Y
X >>>= Y	X = X >>> Y
X &= Y	X = X & Y
X ^= Y	X = X ^ Y
X  = Y	X = X   Y

Table 10. Function call syntax

Syntax	Effect
function (arg1,, argn)	Calls function with the given arguments, and returns the result of the call.
	Examples:parseInt(field) writeln("Hello ", name) doAction() str.substring(start, start+length)
	The function is typically either a script variable reference or a property access, but it can be any expression; the expression must yield a function value, or an error is signalled.
	Examples:callbacks[i](arg) // Calls the function in callbacks[i] "foo"() // Error: a string is not a function

## Special keywords

Lists special reserved keywords in OPL.

Table 11. Special keyword syntax

Syntax	Effect
this	When referenced in a method, returns the current calling object; when referenced in a constructor, returns the object currently being initialized. Otherwise, returns the global object. See "this as a keyword" on page 120 for examples.
arguments	Returns an array containing the arguments of the current function. When used outside a function, an error is signalled.
	For example, the following function returns the sum of all its arguments:function sum() { var res = 0 for (var i=0; i <arguments.length; i++)="" res="" return="" td="" }<=""></arguments.length;>
	The call sum(1, 3, 5) returns 9.

## **Special operators**

Lists the special operators in OPL and their syntax.

Table 12. Special operator syntax

Syntax	Effect
new constructor(arg1,, argn)	Calls the constructor with the given arguments, and returns the created value.
	Examples:new Array() new MyCar("Ford", 1975)
	The constructor is typically a script variable reference, but it can be any expression.
	Example:new ctors[i](arg) // Invokes constructor ctors[i]
typeof value	Returns a string representing the type of value, as follows:Array "object" Boolean "boolean" Date "date" Function "function" Null "object" Number "number" Object "object" String "string" Undefined "undefined"

Table 12. Special operator syntax (continued)

Syntax	Effect
delete variable	Deletes the global script variable variable. This does not mean that the value in variable is deleted, but that it is removed from the global environment.
	Example:myVar = "Hello, world" // Create the global variable myVar delete myVar writeln(myVar) // Signals an error because myVar is undefined
	If variable is a local variable, an error is signalled; if variable is not a known variable, nothing happens.
	The whole expression returns true.
	For C/C++ programmers: The delete operator has a radically different meaning in IBM ILOG Script; in C++, it is used to delete objects, not script variables and properties.
delete value.name  delete value[name]	Remove the property name from the object value.
	If value does not contain the name property, this expression does nothing. If the property does exist but cannot be deleted, an error is signalled. If value is not an object, an error is signalled.
	The whole expression returns the true value.
expression1, expression2	Evaluates expression1 and expression2 sequentially, and returns the value of expression2. The value of expression1 is ignored.
	The most common use for this operator is inside for loops, where it can be used to evaluate several expressions where a single expression is expected:
	for (var i=0, j=0; i<10; i++, j+=2) { writeln(j, " is twice as big as ", i);

# Other operators

Provides a reference of other operators used in OPL.

Other operators are described in detail in the section dedicated to the data type they operate on.

Table 13. Other operator syntax

Syntax	Effect
- X	Arithmetic operators.
X + Y X - Y	These operators perform the usual arithmetic operations. In addition, the + operator can be used to concatenate strings.
X * Y	See "Numeric operators" on page 104 and "String operators" on page 112.
X / Y	
X % Y	
X == Y	Equality operators.
X != Y	These operators can be used to compare numbers and strings; see "Numeric operators" on page 104 and "String operators" on page 112.
	For other types of values, such as dates, arrays, and objects, the == operator is true if, and only if, X and Y are the exact same value. For example:
	new Array(10) == new Array(10) false var a = new Array(10); a == a true
X > Y	Relational operators.
X >= Y X < Y	These operators can be used to compare numbers and strings. See "Numeric operators" on page 104 and "String
X <= Y	operators" on page 112.
~ x	Bitwise operators.
X & Y	See "Numeric operators" on page 104.
X   Y	
x ^ Y	
X << Y	
X >> Y	
X >>> Y	
! X	Logical operators.
X     Y	See "Logical operators" on page 114.
X && Y	
condition ? X : Y	

## **Statements**

A statement can be a conditional statement, a loop statement, a local script variable declaration, a function definition, or a default value.

## **Conditional statement**

Provides a reference for the syntax of conditional statements in OPL.

## Loops

Table 14. Conditional statement syntax

Syntax	Effect
if (expression) statement1	Evaluate expression ; if it is true, execute
[else statement2]	statement1; otherwise, if statement2 is provided, execute statement2.
	If expression gives a non-Boolean value, this value is converted to a Boolean value.
	Examples:
	<pre>if (a == b) writeln("They are equal") else writeln("They are not equal") if (s.indexOf("a") &lt; 0) { write("The string ", s)</pre>
	writeln(" doesn't contains the letter a") }

Table 15. Loop syntax

Syntax	Effect
while (expression) statement	Execute statement repeatedly as long as expression is true. The test takes place before each execution of statement.
	If expression gives a non-Boolean value, this value is converted to a Boolean value.
	Examples:
	while (a*a < b) a = a+1 while (s.length) { $r = s.charAt(0)+r s = s.substring(1)$ }
for ([initialize];	Evaluate initialize once, if present. Its value is ignored. If it has the form:var
[ condition ] ;	variable = expression
[ update ] )	then variable is declared as a local script
statement	variable and initialized as in the var statement.
where <i>condition</i> and <i>update</i> are expressions, and <i>initialize</i> is either an expression or has the form:  var variable = expression	Then, execute statement repeatedly as long as condition is true. If condition is omitted, it is taken to be true, which results in an infinite loop. If condition gives a non-Boolean value, this value is converted to a Boolean value.
	If present, update is evaluated at each pass through the loop, after statement and before condition. Its value is ignored.
	Example:for (var i=0; i < a.length; i++) { sum = sum+a[i] prod = prod*a[i] }

Table 15. Loop syntax (continued)

Syntax	Effect
for ([var] variable in expression) statement	Iterate over the properties of the value of expression: for each property, variable is set to a string representing this property, and statement is executed once.
	If the var keyword is present, variable is declared as a local script variable, as with the var statement.
	For example, the following function takes an arbitrary value and displays all its properties and their values:function printProperties(v) { for (var p in v) writeln(p, " -> ", v[p]) }
	Properties listed by the forin statement include method properties, which are merely regular properties whose value is a function value. For example, the call printProperties("foo") would display:length -> 3 toString -> [primitive method toString] substring -> [primitive method substring] charAt -> [primitive method charAt] etc
	The only properties which are not listed by forin loops are the numeric properties of arrays.
break	Exit the current while, for or fori n loop, and continue the execution at the statement immediately following the loop. This statement cannot be used outside a loop.
	Example:while (i < a.length) { if (a[i] == "foo") { foundFoo = true break } i = i+1 } // Execution continues here
continue	Stop the current iteration of the current while, for or fori n loop, and continue the execution of the loop with the next iteration. This statement cannot be used outside a loop.
	Example: for (var i=0; i < a.length; i++) { if (a[i] < 0) continue writeln("A positive number: ", a[i]) }

## **Declaration of script variables**

Provides a reference for the syntax of script variable declarations in OPL.

Table 16. Declaration syntax for script variables

Syntax	Effect
var decl1,, decln	Declares each script variable as a local variable. If an expression is provided, it is
where each decli	evaluated and its value is assigned to the
has the form	variable as its initial value. Otherwise, the variable is set to the undefined value.
variable [ = expression ]	Examples:var x var name = "Joe" var average = (a+b)/2, sum, message="Hello"

### Inside a function definition

Script variables declared with var are local to the function, and they hide any global variables with the same names; they have the same status as function arguments.

For example, in the following program, the script variables sum and res are local to the average function, as well as the arguments a and b; when average is called, the global variables with the same names, if any, are temporarily hidden until exit from the function:

```
function average(a, b) {
  var sum = a+b
  var res = sum/2
  return res
}
```

Script variables declared with var at any place in a function body have a scope which is the entire function body. This is different from local variable scope in C or C++. For example, in the following function, the variable res declared in the first branch of the if statement is used in the other branch and in the return statement:

```
function max(x, y) {
   if (x > y) {
      var res = x
   } else {
      res = y
   }
return res
}
```

### Outside a function definition

At the same level as function definitions, script variables declared with var are local to the current program unit. A program unit is a group of statements which is considered a whole; the exact definition of a program unit depends on the application in which the script is embedded. Typically, a script file loaded by the application is treated as a program unit. In this case, variables declared with var at the file top level are local to this file, and they hide any global variables with the same names.

For example, suppose that a file contains the following program:

```
var count = 0
function NextNumber() {
  count = count+1
  return count
}
```

When this file is loaded, the function NextNumber becomes visible to the whole application, while count remains local to the loaded program unit and is visible only inside it.

It is an error to declare the same local variable twice in the same scope. For example, the following program is incorrect because res is declared twice:

```
function max(x, y) {
  if (x > y) {
    var res = x
} else {
    var res = y // Error }
  return res
}
```

## **Function definitions**

Provides a reference for the syntax of function definitions in OPL.

Table 17. Function definition syntax

Syntax	Effect
[ static ] function name(v1,, vn)	Defines a function name with the given parameters and body. A function definition can only take place at the top level; function definitions cannot be nested.
{statements}	When the function is called, the script variables v1,, vn are set to the corresponding argument values, then the statements are executed. If a return statement is reached, the function returns the specified value; otherwise, after the statements are executed, the function returns the undefined value.  The number of actual arguments does not need to match the number of parameters: if there are fewer arguments than parameters, the remaining parameters are set to the undefined value; if there are more arguments than parameters, the excess arguments are ignored.  Independently of the parameter mechanism, the function arguments keyword described in Table 11 on page 92.
return [ expression ]	Returns the value of expression from the current function. If expression is omitted, returns the undefined value. The return statement can only be used in the body of a function.

Defining a function name is operationally the same as assigning a specific function value to the variable name; thus a function definition is equivalent to:

```
var name = some function value
```

The function value can be retrieved from the script variable and manipulated like any other type of value. For example, the following program defines a function add and assigns its value to the variable sum, which makes add and sum synonyms for the same function:

```
function add(a, b) {
  return a+b
}
sum = add
```

Without the keyword static, the defined function is global and can be accessed from the whole application. With the keyword static, the function is local to the current program unit, exactly as if name was declared with the keyword var:

```
var name = some function value
```

### **Default values**

Lists the default values used in OPL.

Table 18. Default value syntax

Syntax	Effect
with	Evaluate expression, then execute statement with the value of expression temporarily
(expression)	installed as the default value.
statement	When a reference to an identifier name in statement is evaluated, this identifier is first looked up as a property of the default value; if the default value does not have such a property, name is treated as a regular variable.
	For example, the following program displays "The length is 3", because the identifier length is taken as the length property of the string "abc".
	<pre>with ("abc") { writeln("The length is ", length) }</pre>
	You can nest with statements; in this case, references to identifiers are looked up in the successive default values, from the innermost to the outermost with statement.

# **Built-in values and functions**

Presents the built-in values and functions of the IBM ILOG Script language for OPL: numbers, strings, Booleans, arrays, objects, dates, the null and undefined values, functions.

## **Numbers**

Number representations and functions.

### Introduction

Provides an overview of how numbers are expressed in OPL.

Numbers can be expressed in decimal (base 10), hexadecimal (base 16), or octal (base 8.) There are also special numbers.

#### Note:

**For C/C++ programmers:** Numbers have the same syntax as C and C++ integers and doubles. They are internally represented as 64-bit double-precision floating-point numbers.

### **Decimal numbers**

Explains the use of decimal numbers in OPL.

A decimal number consists of a sequence of digits, followed by an optional fraction, followed by an optional exponent. The fraction consists of a decimal point (.) followed by a sequence of digits; the exponent consists of an e or E followed by an optional + or - sign and a sequence of digits. A decimal number must have at least one digit.

Here are some examples of decimal number literals:

```
15
3.14
4e100
.25
5.25e-10
```

### **Hexadecimal numbers**

Shows how hexadecimal numbers are used in OPL.

A hexadecimal number consists of a 0x or 0X prefix, followed by a sequence of hexadecimal digits, which include digits 0-9 and the letters a-f or A-F. For example: 0x3ff 0x0

## Octal numbers

Explains the use of octal numbers in OPL.

An octal number consists of a 0 followed by a sequence of octal digits, which include the digits 0-7. For example:

0123 0777

### Special numbers

Explains NaN, Infinity, -Infinity, used in OPL.

There are three special numbers: NaN (Not-A-Number), Infinity (positive infinity), and -Infinity (negative infinity).

#### NaN

The special number NaN is used to indicate errors in number manipulations. For example, the square root function Math.sqrt applied to a negative number returns NaN. There is no representation of NaN as a number literal, but the global script variable NaN contains its value.

The NaN value is contagious, and a numeric operation involving NaN always returns NaN. A comparison operation involving NaN always returns false — even the NaN == NaN comparison.

```
Examples of NaN
Math.sqrt(-1) NaN
Math.sqrt(NaN) NaN
NaN + 3 NaN
NaN == NaN false
NaN <= 3 false
NaN >= 3 false
```

Note that, according to the ECMA specification http://www.ecmainternational.org/publications/standards/Ecma-262.htm, the operation ToInt32 (NaN) returns the value 0. For example:

```
int i;
float f;
execute {
i = NaN:
 f = NaN;
writeln(i);
writeln(f);
returns:
NaN
```

# Infinity, -Infinity

The special numbers Infinity and -Infinity are used to indicate infinite values and overflows in arithmetic operations. The global script variable Infinity contains the positive infinity. The negative infinity can be computed using the negation operator (-Infinity).

```
Examples of Infinity
1/0 Infinity
-1/0 -Infinity
1/Infinity 0
Infinity == Infinity true
```

### Automatic conversion to a number

Describes the automatic conversion of numbers in OPL functions.

When a function or a method which expects a number as one of its arguments is passed a nonnumeric value, it tries to convert this value to a number using the following rules:

- A string is parsed as a number literal. If the string does not represent a valid number literal, the conversion yields NaN.
- The Boolean true yields the number 1.
- The Boolean false yields the number 0.
- The null value yields the number 0.
- A date yields the corresponding number of milliseconds since 00:00:00 UTC, January 1, 1970.

For example, if the Math.sqrt function is passed a string, this string is converted to the number it represents:

```
Math.sqrt("25") 5
```

Similarly, operators which take numeric operands attempt to convert any nonnumeric operands to a number:

```
"3" * "4" 12
```

For operators that can take both strings (concatenation) and numbers (addition), such as +, the conversion to a string takes precedence over the conversion to a number (See "Automatic conversion to a string" on page 107.). In other words, if at least one of the operands is a string, the other operand is converted to a string; if none of the operands is a string, the operands are both converted to numbers. For example:

```
"3" + true "3true"
3 + true 4
```

For comparison operators, such as == and >=, the conversion to a number takes precedence over the conversion to a string. In other words, if at least one of the operands is a number, the other operand is converted to a number. If both operands are strings, the comparison is made on strings. For example:

```
"10" > "2" false
"10" > 2 true
```

# **Number methods**

Provides a reference for the number method in OPL.

There is only one number method.

Table 19. Number method

Syntax	Effect
number.toString()	Returns a string representing the number as a literal.
	For example:(14.3e2).toString() "1430"

### **Numeric functions**

Provides a reference for numeric functions in the OPL language.

### Note:

For C/C++ programmers: Most of the numeric functions are wrap-ups for standard math library functions.

Table 20. Numeric functions

Syntax	Effect
Math.abs(x)	Returns the absolute value of x.
Math.max(x,y)	Math.max(x, y) returns the larger of x and y.
Math.min(x,y)	Math.max $(x, y)$ returns the smaller of x and y.
Math.random()	Returns a pseudo-random number between 0, inclusive, and 1, exclusive.

Table 20. Numeric functions (continued)

Syntax	Effect
Math.ceil(x)	Math.ceil(x) returns the smallest integer value greater than or equal to x.
Math.floor(x)	Math.floor(x) returns the greatest integer value less than or equal to x.
Math.round(x)	Math.round(x) returns the nearest integer value to x.
Math.sqrt(x)	Returns the square root of x.
Math.sin(x)	Math.sin(x) returns the trigonometric function sine of a radian argument.
Math.cos(x)	Math.cos(x) returns the trigonometric function cosine of a radian argument.
Math.tan(x)	Math.tan(x) returns the trigonometric function tangent of a radian argument.
Math.asin(x)	Math.asin(x) returns the arcsine of x in the range $-pi/2$ to $pi/2$ .
Math.acos(x)	Math.acos(x) returns the arc cosine of x in the range $\theta$ to pi.
Math.atan(x)	Math.atan(x) returns the arc tangent of x in the range $-pi/2$ to $pi/2$ .
Math.atan2(y,x)	Math.atan2( $y$ , $x$ ) converts rectangular coordinates ( $x$ , $y$ ) to polar coordinates ( $r$ , a) by computing a as an arc tangent of $y/x$ in the range -pi to pi.
Math.exp(x)	Math.exp(x) computes the exponential function.
Math.log(x)	Math.log(x) computes the natural logarithm of x.
Math.pow(x,y)	Math.pow(x, y) computes x raised to the power y.

# **Numeric constants**

Provides a reference for numeric constants in OPL.

The following numeric constants are defined.

Table 21. Numeric constants

Syntax	Value
NaN	Contains the NaN value.
Infinity	Contains the Infinity value.
Number.NaN	Same as NaN.
Number.MAX_VALUE	The maximum representable number, approximately 1.79E+308.
Number.MIN_VALUE	The smallest representable positive number, approximately 2.22E-308.
Math.E	Napier's constant, e, and the base of natural logarithms, approximately 2.718.

Table 21. Numeric constants (continued)

Syntax	Value
Math.LN10	The natural logarithm of 10, approximately 2.302.
Math.LN2	The natural logarithm of 2, approximately 0.693.
Math.LOG2E	The base 2 logarithm of e, approximately 1.442.
Math.LOG10E	The base 10 logarithm of e, approximately 0.434.
Math.PI	The ratio of the circumference of a circle to its diameter, approximately 3.142.
Math.SQRT1_2	The square root of one-half, approximately 0.707.
Math.SQRT2	The square root of 2, approximately 1.414.

# **Numeric operators**

Provides a reference for numeric operators in OPL.

### Note:

**For C/C++ programmers:** The numeric operators are the same as in C and C++.

Table 22. Numeric operators

Syntax	Effect
x+y	The usual arithmetic operations.
х-у	Examples:
x*y	3 + 4.2 -> 7.2
x/y	100 - 120 -> -20
	4 * 7.1 -> 28.4
	6 / 5 -> 1.2
-x	Negation.
	Examples:
	- 142
	-142
x%y	Returns the floating-point remainder of dividing x by y.
	Examples:
	12 % 5 -> 2
	12.5 % 5 -> 2.5

Table 22. Numeric operators (continued)

Syntax	Effect
x==y x!=y	The operator == returns true if x and y are equal, and false otherwise. The operator != is the converse of ==.
	Examples:
	12 == 12 -> true
	12 == 12.1 -> false
	12 != 12.1 -> true
x <y x&lt;=y</y 	The operator < returns true if x is less than y, and false otherwise. The operator <= returns true if x is less than or equal to y,
x>y	and false otherwise; and so on.
x>=y	Examples:
	-1 < 0 -> true
	1 < 1 -> false
	1 <= 1 -> true
x&y	The bitwise operations & (AND),
x y	(inclusive OR), and ^ (exclusive OR), where x and y must be integers in the range
x^y	-2**31+1 to 2**31-1 (-2147483647 to 2147483647.)
	Examples:
	14 & 9 -> 8 (because 1110 & 1001 -> 1000)
	14   9 -> 15 (because 1110   1001 -> 1111)
	14 ^ 9 -> 7 (because 1110 ^ 1001 -> 111)
~ x	The bitwise NOT operation, where x must be an integer in the range -2**31+1 to 2**31-1 (-2147483647 to 2147483647.)
	Examples:
	~ 14 -> 1 (because ~ 1110 -> 0001)

Table 22. Numeric operators (continued)

Syntax	Effect
x< <y< td=""><td>Binary shift operations, where x and y must be integers in the range -2**31+1 to 2**31-1</td></y<>	Binary shift operations, where x and y must be integers in the range -2**31+1 to 2**31-1
x>>y	(-2147483647 to 2147483647.) The operator <<
x>>>y	shifts to the left, >> shifts to the right (maintaining the sign bit). The left operand specifies the value to be shifted. The right operand specifies the number of positions that the bits in the value are to be shifted.
	>>> shifts to the right, shifting in zeros from the left.
	Examples:
	9 << 2 -> 36 (because 1001 << 2 -> 100100)
	9 >> 2 -> 2 (because 1001 >> 2 -> 10)
	-9 >> 2 -> -2 (because 111001 >> 2 -> 111110)
	-9 >>> 2 -> 1073741821 (because 111001 >>> 2 -> 0111110)

# **IBM ILOG Script strings**

String representation and functions.

### Introduction

Provides an overview of the use of strings in IBM ILOG Script.

A string literal is zero or more characters enclosed in double (") or single (') quotes.

## Note:

**For C/C++ programmers:** Except for the use of single quotes, string literals have the same syntax as in C and C++.

Here are examples of string literals:

```
"My name is Hal"
'My name is Hal'
'"Hi there", he said'
"3.14"
"Hello, world\n"
```

In these examples, the first and second strings are identical.

The backslash character \ can be used to introduce an escape sequence, which stands for a character which cannot be directly expressed in a string literal.

Table 23. Escape sequences in strings

Escape sequence	Stands for
\n	Newline
\t	Tab

Table 23. Escape sequences in strings (continued)

Escape sequence	Stands for
\\	Backslash character (\)
\"	Double quote (")
\'	Single quote (')
\b	Backspace
\f	Form feed
\r	Carriage return
\xhh	The character whose ASCII code is hh, where hh is a sequence of two hexadecimal digits.
\000	The character whose ASCII code is ooo, where ooo is a sequence of one, two, or three octal digits.

Here are examples of string literals using escape sequences:

Table 24. Examples of string literals using escape sequences

String literal	Stands for
"Read 'The Black Bean'"	Read 'The Black Bean'
'"Hello", he said'	"Hello", he said
"c:\\temp"	c:\temp
"First line\nSecond line\nThird line"	First line
	Second line
	Third line
"\xA9 1995-1997"	© 1995-1997

When a string is converted to a number, an attempt is made to parse it as a number literal. If the string does not represent a valid number literal, the conversion yields NaN.

### Automatic conversion to a string

Explains how nonstrings are automatically converted to strings.

When a function or a method that expects a string as one of its arguments is passed a nonstring value, this value is automatically converted to a string. For example, if the string method index0f is passed a number as its first argument, this number is treated like its string representation:

Similarly, operators that take string operands automatically convert nonstring operands to strings:

The conversion to a string uses the toString method of the given value. All built-in values have a toString method.

<sup>&</sup>quot;The 10 commandments".index0f(10) -> 4

<sup>&</sup>quot;The " + 10 + " commandments" -> "The 10 commandments"

# **String properties**

Provides a reference for the properties of strings.

There is a single, read-only string property.

Table 25. String property

Syntax	Value
string.length	Number of characters in string. This is a read-only property.
	Examples:
	"abc".length -> 3
	"".length -> 0

# String methods

Provides a reference for string methods.

Characters in a string are indexed from left to right. The index of the first character in a string is 0, and the index of the last character is string.length-1.

Table 26. String methods

Syntax	Effect
string.substring(start [ , end ] )	Returns the substring of string starting at the index start and ending at the index end-1. If end is omitted, the tail of string is returned.
	Examples:
	"0123456".substring(0, 3) -> "012"
	"0123456".substring(2, 4) -> "23"
	"0123456".substring(2) -> "23456"
string.charAt(index)	Returns a one-character string containing the character at the specified index of string. If index is out of range, an empty string is returned.
	Examples:
	"abcdef".charAt(0) -> "a"
	"abcdef".charAt(3) -> "d"
	"abcdef".charAt(100) -> ""
string.charCodeAt(index)	Returns the ASCII code of the character at the specified index of string. If index is out of range, returns NaN.
	Examples:
	"abcdef".charCodeAt(0) -> 97
	"abcdef".charCodeAt(3) -> 100
	"abcdef".charCodeAt(100) -> NaN

Table 26. String methods (continued)

Syntax	Effect
string.indexOf(substring [ , index ] )	Returns the index in string of the first occurrence of substring. The string is searched starting at index. If index is omitted, string is searched from the beginning. This method returns -1 if substring is not found.
	Examples:
	"abcdabcd".indexOf("bc") -> 1
	"abcdabcd".indexOf("bc", 1) -> 1
	"abcdabcd".indexOf("bc", 2) -> 5
	"abcdabcd".indexOf("bc", 10) -> -1
	"abcdabcd".indexOf("foo") -> -1
	"abcdabcd".indexOf("BC") -> -1
string.lastIndexOf(substring [ , index ] )	Returns the index in string of the last occurrence of substring, when string is searched backwards, starting at index. If index is omitted, string is searched from the end. This method returns -1 if substring is not found.
	Examples:
	"abcdabcd".lastIndexOf("bc") -> 5
	"abcdabcd".lastIndexOf("bc", 5) -> 5
	"abcdabcd".lastIndexOf("bc", 4) -> 1
	"abcdabcd".lastIndexOf("bc", 0) -> -1
	"abcdabcd".lastIndexOf("foo") -> -1
	"abcdabcd".lastIndexOf("BC") -> -1
string.toLowerCase()	Returns string converted to lowercase.
	Example: "Hello, World".toLowerCase() "hello, world"
string.toUpperCase()	string.toUpperCase() Returns string converted to uppercase.
	Example:"Hello, World".toUpperCase() "HELLO, WORLD"

Table 26. String methods (continued)

Syntax	Effect
string.split(separator)	Returns an array of strings containing the substrings of string that are separated by separator. See also the array method join.
	Examples:
	"first name, last name, age".split(",") -> an array a such that .length is 3, a[0] is "first name", a[1] is "last name", and a[2] is "age".
	If string does not contain separator, an array with one element containing the whole string is returned.
	Example:
	"hello".split(",") -> an array a such that a.length is 1 and a[0] is "hello",
string.toString()	Returns the string itself.

# **String functions**

Provides a reference of string functions.

Table 27. String functions

Syntax	Effect
String.fromCharCode(code)	Returns a single-character string containing the character with the given ASCII code.
	Examples:
	String.fromCharCode(65) ->"A"
	writeln(String.fromCharCode(0x30)); -> "0"

Table 27. String functions (continued)

Syntax	Effect
parseInt(string [ , base ] )	Parses string as an integer written in the given base, and returns its value. If the string does not represent a valid integer, NaN is returned.
	Leading white space characters are ignored. If parseInt encounters a character that is not a digit in the specified base, it ignores it and all succeeding characters and returns the integer value parsed up to that point.
	If base is omitted, it is taken to be 10, unless string starts with 0x or 0X, in which case it is parsed in base 16, or with 0, in which case it is parsed in base 8.
	Examples:
	parseInt("123") -> 123
	parseInt("-123") -> -123
	parseInt("123.45") -> 123
	parseInt("1001010010110", 2) -> 4758
	parseInt("a9", 16) -> 169
	parseInt("0xa9") -> 169
	parseInt("010") -> 8
	parseInt("123 poodles") -> 123
	parseInt("a lot of poodles") -> NaN
parseFloat(string)	Parses string as a floating-point number and return its value. If the string does not represent a valid number, NaN is returned.
	Leading white space characters are ignored. The string is parsed up to the first unrecognized character. If no number is recognized, the function returns NaN.
	Examples:
	parseFloat("-3.14e-15") -> -3.14e-15
	parseFloat("-3.14e-15 poodles") -> -3.14e-15
	parseFloat("a fraction of a poodle") -> NaN

# **String operators**

Provides a reference of string operators.

Table 28. String operators

Syntax	Effect
string1 + string2	Returns a string containing the concatenation of string1 and string2.
	Examples:"Hello," + " world" -> "Hello, world"
	When the operator + is used to add a string to a nonstring value, the nonstring value is first converted to a string.
	Examples:
	"Your age is " + 23 -> "Your age is 23"
	23 + " is your age" -> "23 is your age"
string1 == string2 string1 != string2	The operator == returns the Boolean true if string1 and string2 are identical, and false otherwise. Two strings are identical if they have the same length and contain the same sequence of characters. The operator != is the converse of ==.
	Examples:
	"a string" == "a string" -> true
	"a string" == "another string" -> false
	"a string" == "A STRING" -> false
	"a string" != "a string" -> false
	"a string" != "another string" -> true
	When the operators == and != are used to compare a string with a number, the string is first converted to a number and the two numbers are compared numerically.
	Examples:
	"12" == "+12" -> false
	12 == "+12" -> true

Table 28. String operators (continued)

Syntax	Effect
string1 < string2  string1 <= string2  string1 > string2  string1 >= string2	The operator < returns true if string1 strictly precedes string2 lexicographically, and false otherwise. The operator <= returns true if string1 strictly precedes string2 lexicographically or is equal to it, and false otherwise; and so on.
string1 >= string2	Examples:
	"abc" < "xyz" -> true
	"a" < "abc" -> true
	"xyz" < "abc" -> false
	"abc" < "abc" -> false
	"abc" > "xyz" -> false
	"a" > "abc" -> false
	"xyz" > "abc" -> true
	Etc.
	When one of these operators is used to compare a string with a number, the string is first converted to a number and the two numbers are compared numerically. In all other cases, the other argument is first converted to a string.
	Examples:
	"10" > "2" -> false
	10 > "2" -> true
	123 < "2" -> false
	Hint: Autocasting may cause unexpected behavior.

# **IBM ILOG Script Booleans**

Boolean representation and functions.

### Introduction

Provides an overview of the use of Boolean literals in IBM ILOG Script.

There are two Boolean literals: true, which represents the Boolean value true, and false, which represents the Boolean value false.

When converted to a number, true yields 1 and false yields 0.

### **Automatic conversion to Boolean**

Describes the automatic conversion of non-Booleans to Booleans.

When a function, method or statement which expects a Boolean value as one of its arguments is passed a non-Boolean value, this value is automatically converted to a Boolean value as follows:

- The number 0 yields false.
- The empty string "" yields false.
- The null value yields false.
- The undefined value yields false.
- Any other non-Boolean values yield true.

## For example:

```
if ("") writeln("True"); else writeln("False");
if (123) writeln("True"); else writeln("False");
```

This displays "False", then "True".

### **Boolean methods**

Provides a reference of Boolean methods.

There is only one Boolean method.

Table 29. Boolean method

Syntax	Effect
boolean.toString()	Returns a string representing the Boolean value, either "true" or "false".
	Example:
	true.toString -> "true"
	false.toString -> "false"

### Logical operators

Provides a reference of logical operators.

The following Boolean operators are available.

### Note:

**For C/C++ programmers:** These operators are the same as in C and C++.

Table 30. Logical operators

Syntax	Effect
! boolean	Logical negation.
	Examples:
	! true -> false
	! false -> true

Table 30. Logical operators (continued)

Syntax	Effect
exp1 && exp2	Logical AND. Returns true if both Boolean expressions exp1 and exp2 are true. Otherwise, returns false.
	If exp1 is false, this expression immediately returns false without evaluating exp2, so any side effects of exp2 are not taken into account.
	Examples:
	true && true -> true
	true && false -> false
	false && whatever -> false (whatever is not evaluated)
exp1    exp2	Logical OR. Returns true if either Boolean expression exp1 or exp2 (or both) is true. Otherwise, returns false.
	If exp1 is true, this expression immediately returns true without evaluating exp2, so any side effects of exp2 are not taken into account.
	Examples:
	false    true -> true
	false    false -> false
	true    whatever -> true (whatever is not evaluated)
condition ? exp1 : exp2	If condition is true, this expression returns exp1; otherwise, it returns exp2.
	When condition is true, the expression exp2 is not evaluated, so any side effects it may contain are not taken into account. Similarly, when condition is false, exp1 is not evaluated.
	Examples:
	true ? 3.14 : whatever -> 3.14
	false ? whatever : "Hello" -> "Hello"

# **IBM ILOG Script arrays**

Array representation and functions.

# Introduction

Provides an overview of the use of arrays in IBM ILOG Script.

Arrays provide a way of manipulating ordered sets of values referenced through an index starting from zero (0). Unlike arrays in other languages, IBM ILOG Script arrays do not have a fixed size and are automatically expanded as new elements are added. For example, in the following program, an array is created empty, and new elements are then added.

```
a = new Array() // Create an empty array
a[0] = "first" // Set the element 0
a[1] = "second" // Set the element 1
a[2] = "third" // Set the element 2
```

Arrays are internally represented as sparse objects, which means that an array where only the element 0 and the element 10000 have been set occupies just enough memory to store these two elements, not the 9999 which are between 0 and 10000.

## Array constructor

Provides a reference for array constructors in IBM ILOG Script.

The array constructor has two distinct forms.

Table 31. Array constructor

Syntax	Effect
new Array(length)	Returns a new array a of length length with its elements from 0 to length-1 set to null.
	If length is not a number, and its conversion to a number yields NaN, the second syntax is used.
	Examples:
	new Array(12) -> an array a with length 12 and a[0] to a[11] containing null
	new Array("5") -> an array a with length 5 and a[0] to a[4] containing null
	new Array("foo") see second syntax
new Array(element1,, elementn)	Returns a new array a of length n with a[0] containing element1, a[1] containing element2, and so on. If no argument is given, that is n=0, an empty array is created. If n=1 and element1 is a number or can be converted to a number, the first syntax is used.
	Examples:
	new Array(327, "hello world") -> an array a of length 2 with a[0] == 327 and a[1] == "hello world"
	new Array() -> an array with length 0
	new Array("327") see first syntax

# **Array properties**

Provides a reference of the properties of arrays.

Table 32. Array properties

Syntax	Effect
array[index]	If index can be converted to a number between 0 and 2e31-1 (see "Automatic conversion to a number" on page 101), array[index] is the value of the index-th element of the array.
	Otherwise, it is considered as a standard property access.
	If this element has never been set, null is returned.
	Example:
	Suppose that the array a has been created witha = new Array("foo", 12, true)
	Then:
	a[0] -> "foo"
	a[1] -> 12
	a[2] -> true
	a[3] -> null
	a[1000] -> null
	When an element of an array is set beyond the current length of the array, the array is automatically expanded:a[1000] = "bar" // the array is automatically expanded.
	Unlike other properties, the numeric properties of an array are not listed by the forin statement.

Table 32. Array properties (continued)

Syntax	Effect
array.length	The length of array, which is the highest index of an element set in array, plus one. It is always included in 0 to 2e31 -1.
	When a new element is set in the array, and its index is greater than or equal to the current array length, the length property is automatically increased.
	Example: Suppose that the array a has been created witha = new Array("a", "b", "c")
	Then:a.length -> 3 a[100] = "bar"; a.length -> 101
	You can also change the length of an array by setting its length property.
	a = new Array(); a[4] = "foo"; a[9] = "bar" a.length -> 10
	a.length = 5 a.length -> 5 a[4] -> "foo" a[9] -> null

# **Array methods**

Provides a reference of array methods.

Table 33. Array methods

Syntax	Effect
array.join([ separator ])	Returns a string that contains the elements of the array converted to strings, concatenated together and separated with <i>separator</i> . If <i>separator</i> is omitted, it is taken to be ",". Elements that are not initialized are converted to the empty string. See also the string method split.  Example: Suppose that the array a has been created with
	a = new Array("foo", 12, true)
	Then:
	a.join("//") -> "foo//12//true"
	a.join() -> "foo,12,true"

Table 33. Array methods (continued)

Syntax	Effect
array.sort([ function ])	Sorts the array. The elements are sorted in place; no new array is created.
	If <i>function</i> is not provided, <i>array</i> is sorted lexicographically: Elements are compared by converting them to strings and using the < operator. With this order, the number 20 would come before the number 5, since "20" < "5" is true.
	If <i>function</i> is supplied, the array is sorted according to the return value of this function. This function must take two arguments x and y and return:
	-1 if x is smaller than y
	0 if x is equal to y
	1 if x is greater than y
	Example: Suppose that the function compareLength is defined as
	function compareLength(x, y) {if (x.length < y.length) return -1; else if (x.length == y.length) return 0; else return 1; }
	and that the array a has been created with:
	a = new Array("giraffe", "rat", "brontosaurus")
	Then a.sort() will reorder its elements as follows:"brontosaurus" "rat" "giraffe"
	while a.sort(compareLength) will reorder them as follows:"rat" "giraffe" "brontosaurus"
array.reverse()	Transposes the elements of the array: the first element becomes the last, the second becomes the second to last, and so on. The elements are reversed in place; no new array is created.
	Example: Suppose that the array a has been created with
	a = new Array("foo", 12, "hello", true, false)
	Then a.reverse() changes a so that:
	a[0] false
	a[1] true
	a[2] "hello"
	a[3] 12
	a[4] "foo"

Table 33. Array methods (continued)

Syntax	Effect
array.toString()	Returns the string "[object Object]".

# **Objects**

Object representation and functions.

### Introduction

Provides an overview of the use of objects in IBM ILOG Script.

Objects are values which do not contain any predefined properties or methods (except the toString method), but where new ones can be added. A new, empty object can be created using the Object constructor. For example, the following program creates a new object, stores it in the variable myCar, and adds the properties "name" and "year" to it:

```
myCar = new Object() // o contains no properties
myCar.name = "Ford"
myCar.year = 1985

Now:
myCar.name -> "Ford"
myCar.year -> 1985
```

# **Defining methods**

Explains how methods are defined in IBM ILOG Script.

Since a method is really a property which contains a function value, defining a method simply consists in defining a regular function, then assigning it to a property.

For example, the following program adds a method start to the myCar object defined in the previous section:

```
function start_engine() {
  writeln("vroom vroom")
}
myCar.start = start_engine
```

Now, the expression myCar.start() will call the function defined as start\_engine. Note that the only reason for using a different name for the function and for the method is to avoid confusion; we could have written:

```
function start() {
  writeln("vroom vroom")
}
myCar.start = start
```

# this as a keyword

Describes the use of the this keyword in IBM ILOG Script.

Inside methods, the this keyword can be used to reference the calling object. For example, the following program defines a method getName, which returns the value of the name property of the calling object, and adds this method to myCar:

```
function get name() {
  return this.name
myCar.getName = get_name
```

Inside constructors, this references the object created by the constructor.

When used in a nonmethod context, this returns a reference to the global object. The global object contains script variables declared at top level, and built in functions and constructors.

## Object constructor

Provides a reference of object constructors in IBM ILOG Script.

Table 34. Object constructor

Syntax	Effect
new Object()	Returns a new object with no properties.

### **User-defined constructors**

Provides a reference of user-defined constructors.

In addition to the Object constructor, any user-defined function can be used as an object constructor.

Table 35. User-defined constructors

Syntax	Effect
new function(arg1,, argn)	Creates a new object, then calls function(arg1,, argn) to initialize it.

Inside the constructor, the keyword this can be used to make reference to the object being initialized.

For example, the following program defines a constructor for cars:

```
function Car(name, year) {
  this.name = name;
  this.year = year;
  this.start = start engine;
Now, calling
new Car("Ford", "1985")
```

creates a new object with the properties name and year, and a start method.

### **Built-in methods**

Provides a reference of the built-in methods of IBM ILOG Script.

There is only one object built-in method.

Table 36. Built-in method

Syntax	Effect
	Returns the string "[object Object]". This method can be overridden by assigning the toString property of an object.

## **Dates**

Date representation and functions

### Introduction

Provides an overview of dates and date functions in IBM ILOG Script.

Date values provide a way of manipulating dates and times. Dates can be best understood as internally represented by a number of milliseconds since 00:00:00 UTC, January 1, 1970. This number can be negative, to express a date before 1970.

#### Note:

- 1. For C/C++ programmers: Unlike dates manipulated by the standard C library, date values are not limited to the range of 1970 to 2038, but span approximately 285,616 years before and after 1970.
- 2. IBM ILOG Script measures time in wall-clock time (and not system time).

When converted to a number, a date yields the number of milliseconds since 00:00:00 UTC, January 1, 1970.

## **Date constructor**

Explains the different forms of the date constructor.

The date constructor has four distinct forms.

Table 37. Date constructor

Syntax	Effect
new Date()	Returns the date representing the current time.
new Date(milliseconds)	Returns the date representing 00:00:00 UTC, January 1, 1970, plus milliseconds milliseconds. The argument can be negative, to express a date before 1970. If the argument cannot be converted to a number, the third constructor syntax is used.  Examples:  new Date(0) -> a date representing 00:00:00 UTC, January 1, 1970.  new Date(1000*60*60*24*20) -> a date representing twenty days after 00:00:00 UTC, January 1, 1970.  new Date(-1000*60*60*24*20) -> a date representing twenty days before 00:00:00 UTC, January 1, 1970.

Table 37. Date constructor (continued)

Syntax	Effect
new Date(string)	Returns the date described by string, which must have the form:
	month/day/year hour:minute:second msecond
	The date expressed in string is taken in local time.
	Example:
	new Date("12/25/1932 14:35:12 820")
	A date representing December 25th, 1932, at 2:35 PM plus 12 seconds and 820 milliseconds, local time.
new Date(year, month,	Returns a new date representing the given year, month, day, and so on, taken in local time. The arguments are:
[ , day	year: any integer.
[ , hours	month: range 0-11 (where 0=January, 1=February, and so on)
[, seconds	day: range 1-31, default 1
[ , mseconds ]]]]])	hours: range 0-23, default 0
	minutes: range 0-59, default 0
	seconds: range 0-59, default 0
	mseconds: range 0-999, defaults to 0
	Examples:
	new Date(1932, 11, 25, 14, 35, 12, 820)
	A date representing December 25th, 1932, at 2:35 PM plus 12 seconds and 820 milliseconds, local time.
	new Date(1932, 11, 25)
	A date representing December 25th, 1932, at 00:00, local time.

## **Date methods**

Provides a reference of date methods in IBM ILOG Script.

Table 38. Date methods

Syntax	Effect		
date.getTime()	Returns (or sets) the number of milliseconds since 00:00:00 UTC, January 1, 1970.		
date.setTime(milliseconds)	Example: Suppose that the date d has been created with:		
	d = new Date(3427)		
	Then:		
	d.getTime() -> 3427		
date.toLocaleString() date.toUTCString()	Returns a string representing the date in local time or in UTC respectively.		
	Example: Suppose that the date d has been created with:		
	d = new Date("3/12/1997 12:45:00 0")		
	Then:		
	d.toLocaleString() -> "03/12/1997 12:45:00 000"		
	d.toUTCString() -> "03/12/1997 10:45:00 000"		
	assuming a local time zone offset of +2 hours with respect to Greenwich Mean Time.		
date.getYear()	Returns (or sets) the year of date.		
date.setYear(year)			
date.getMonth()	Returns (or sets) the month of date.		
date.setMonth(month)			
date.getDate()	Returns (or sets) the day of date.		
date.setDate(day)			
date.getHours()	Returns (or sets) the hours of date.		
date.setHours(day)			
date.getMinutes()	Returns (or sets) the minutes of date.		
date.setMinutes(minutes)			
date.getSeconds()	Returns (or sets) the seconds of date.		
date.setSeconds(seconds)			
date.getMilliseconds()	Returns (or sets) the milliseconds of date.		
date.setMilliseconds(millisecs)			
date.toString()	Returns the same value as date.toLocaleString()		

## Related reference:

"Tips on date formats"

Provides some tips on using ILOG Script to improve the date format.

### **Date functions**

Provides a reference of date functions.

Table 39. Date functions

Syntax	Effect
Date.UTC(year, month, [ , day [ , hours [ , minutes [ , seconds [ , mseconds ]]]]])	Returns a number representing the given date taken in UTC. The arguments are:
	year: any integer
	month: range 0-11, where 0 = January, 1 = February, and so on
	day: range 1-31, default 1
	hours: range 0-59, default 0
	minutes: range 0-59, default 0
	seconds: range 0-59, default 0
	mseconds: range 0-999, default 0
Date.parse(string)	Same as new Date(string), but the result is returned as a number rather than as a date object.

# **Date operators**

Explains the use of date operators in IBM ILOG Script.

There are no specific operators for dealing with dates, but, since numeric operators automatically convert their arguments to numbers, these operators can be used to compute the time elapsed between two dates, to compare dates, or to add a given amount of time to a date. For example:

```
date1 - date2 -> the number of milliseconds elapsed between date1 and date2.
date1 < date2 -> true if date1 is before date2, false otherwise.
new Date(date+10000) ->
a date representing 10000 milliseconds after date.
```

The following program displays the number of milliseconds taken to execute the statement <do something>:

```
before = new Date();
<do something>;
after = new Date();
writeln("Time for doing something: ", after-before, " milliseconds.");
```

# Tips on date formats

Provides some tips on using ILOG Script to improve the date format.

### Setting the date

The set methods on dates perform safety checks. For example, when you try to set the day, the corresponding month must have the correct number of days. If the date is 2011/ February/15 and you try to change the day to 31, an error will occur. Therefore, to avoid errors when you create a new date, you must respect the order: Year / Month / Day.

### Modifying the date format

If your date is not in the format you want, you need to use ILOG Script to modify it. See the following example.

```
execute {
function converter(dateIn) {
dateTime = dateIn.split(" ");
date = dateTime[0]; time = dateTime[1];

dateSplit = date.split("-");
dateReverse = dateSplit.reverse();
dateNew = dateReverse.join("/") + " " + time;
return Date.parse(dateNew);
}

var d1 = converter("2011-07-06 07:00:00.0");
writeln(d1);
}
```

This code will return a new date in the format day/month/year. So in the example, the date returned is:

06/07/2011 07:00:00.0

## The null value

Explains the use of the null value.

The null value is a special value used in some places to specify an absence of information.

For example, an array element which has not yet been set has a default value of null. The null value is not to be confused with the undefined value, which also specifies an absence of information in some contexts. See section "The undefined value" below.

The null value can be referenced in programs with the keyword null: null -> the null value

When converted to a number, null yields zero (0).

### Methods of null

There is only one method of null.

Table 40. Methods of null

Syntax	Effect
null.toString()	Returns the string "null".

# The undefined value

Explains the use of the undefined value.

The undefined value is a special value used in some places to specify an absence of information. For example, accessing a property of a value which is not defined, or a local variable which has been declared but not initialized, yields the undefined value.

There is no way of referencing the undefined value in programs. Checking if a value is the undefined value can be done using the typeof operator:

```
typeof(value) == "undefined" -> true if value is undefined,
                                 false otherwise.
```

### Methods of undefined

There is only one method of undefined.

Table 41. Methods of undefined

Syntax	Effect
undefined.toString()	Returns the string "undefined".

# IBM ILOG Script functions

Describes the use of functions in IBM ILOG Script.

In IBM ILOG Script, functions are regular values (also known as "first class" values) which can be manipulated like any other type of value: They can be passed to functions, returned by functions, stored into script variables or into object properties, and so on.

For example, the function parseInt is a function value which is stored in the parseInt variable:

```
parseInt ->
a function value
```

This function value can be, for example, assigned to another variable: myFunction = parseInt

and then called through this variable: myFunction("-25") -> -25

### **Function methods**

There is only one method of functions.

Table 42. Function methods

Syntax	Effect
function.toString()	Returns a string which contains some information about the function.
	Examples:
	"foo".substring.toString() "[primitive method substring]"
	eval.toString() "[primitive function eval]"

# **Miscellaneous functions**

Provides a reference of miscellaneous functions in IBM ILOG Script.

Table 43. Miscellaneous functions

Syntax	Effect	
stop()	Stops the execution of the program at the current statement and, if the debugger is enabled, enters debug mode.	
write (arg1,, argn) writeln (arg1,, argn)	Converts the arguments to strings and prints them to the current debug output. The implementation depends on the application in which IBM ILOG Script is embedded. The function writeln prints a newline at the end of the output, while write does not.	
loadFile(string)	Loads the script file whose path is string. The path can be either absolute or relative. If this path does not designate an existing file, the file is looked up using a method that depends on the application in which the script is embedded. Typically, a file with the name string is searched for in a list of directories specified in the application setup.	
includeScript(string)	Loads the script file whose path is string.	
For use in OPL.	The path can be either absolute or relative. If this path does not designate an existing file, the file is looked up using a method that depends on the application in which the script is embedded. Typically, a file with the name string is searched for in a list of directories specified in the application setup.  The effect of this function is the same as for loadFile. In OPL, you have to use	
	includeScript, as the function loadFile does not recognize OPL objects, such as thisOplModel.	
eval(string)	Executes string as a program, and returns the value of the last evaluated expression. The program in string can use all the features of the language, except that it cannot define functions; in other words, the function statement is not allowed in string.	
	Examples:	
	eval("2*3") -> 6	
	eval("var i=0; for (var j=0; j<100; j++) i=i+j; i") -> 4950 n=25; eval("Math.sqrt(n)") -> 5 eval("function foo(x) { return x+1 }") -> error	

Table 43. Miscellaneous functions (continued)

Syntax	Effect
fail()	Stops the execution of the scripting block at the current statement, reports an error, and goes on.
	Example:
	execute b1 { writeln("A"); fail(); writeln("B"); } execute b2 {writeln("C"); }
	gives
	A C
	as the output
	and reports an error line 4
	Scripting runtime error: fail() called.

# Index

A	conditional statements	data structures (continued)
abs, OPL function 37	in IBM ILOG Script for OPL 87, 95	ranges 7
accessing	connection	sets 12
value of an IBM ILOG Script	to a spreadsheet 24	data types
property 89	consistency of model data 27	arrays 8
accessing named ranges	constants 103	floats 4
in Excel 25	dynamic collection 34	integers 4
aggregate operators 38, 61	constraint labeling	piecewise linear functions 5
all, OPL keyword 34	limitations 51	ranges 7
and, logical constraint 55	constraints	sets 12
arguments	basic 53	stepwise functions 6
IBM ILOG Script keyword 92	conditional 47	strings 5
arrays	declaration 46 discrete 53	data types allowed in tuples 11 data, pushing and pulling 23
appending 35	element expressions (CP) 57	data, pushing and punning 25 date format, tips using script 125
constructor (IBM ILOG Script) 116	float 53	dates in IBM ILOG Script 122
explicit 35	for compatibility (CP) 58	constructor 122
in IBM ILOG Script 115	for filtering 47	functions 125
initialization 17	labeling 48	methods 124
methods (IBM ILOG Script) 118	limitations 59	operators 125
multidimensional 8	logical (CPLEX) 54	decimal numbers 100
of decision variables, initialization 36	logical, for CP 57	decision expressions 34
one-dimensional 8	nonlinear, rejected by CPLEX 53	and tuple patterns/tuple indices 11
properties (IBM ILOG Script) 117	scheduling 58	decision variables 32
assert, OPL keyword 29	specialized (CP) 58	and integer ranges 7
processing order of statements 30	string 54	arrays of, initialization 36
assertions 29	types 53	definition 32
assignment operators 90	using 46	dynamic collection 34
	constraints, limitations 59	reusable (dexpr) 34
_	constructors	declaration
В	for IBM ILOG Script arrays 116	of constraints 46
Boolean expressions 45	for IBM ILOG Script dates 122	of script variables
constraints 46	for IBM ILOG Script objects 121	inside a function definition 97
Boolean literals 113	user-defined in IBM ILOG Script 121	outside a function definition 97
conversion to 114	continue, IBM ILOG Script keyword 95	of tuples, using keys 10
Boolean method toString 114	conventions	default values
brackets, delimiters in IBM ILOG Script	in models 3	of statements in IBM ILOG Script 99
for OPL 87	conversion (IBM ILOG Script)	delimiters
break, IBM ILOG Script keyword 95	of non-Boolean value to a	and internal initialization
breakpoint 40	Boolean 114	tuples 20
breakpoints	of nonnumeric value to a	curly brackets 87
in piecewise-linear functions 39	number 101	for comments 88
building blocks 3	of nonstring value to a string 107	for tuples 19
	costs	quotes 106, 113
^	discontinuous (pwl) 41	semi-colon 87
C	count, OPL function 37	dexpr, OPL keyword 34
card, OPL function 44		diff, OPL keyword 44 discontinuous piecewise-linear
cardinality constraints 56	n	functions 40
case-sensitivity of the scripting	ט	discrete
language 88	data	constraints 53
collections	assertions for consistency 29	data 53
not sorted in tuple sets without	consistency 27	variables 53
keys 14	initializing 15	dvar, OPL keyword
comments	input/output	vs. var 32
delimiters 88	to/from a spreadsheet 24	
compatibility constraints in CP 58	preprocessing 30	
compound statements	reading from an Excel	E
in IBM ILOG Script for OPL 87	spreadsheet 24	<del></del>
conditional constraints 47	data initialization, lazy 16, 23, 31	efficient models 48, 61, 62
conditional expressions	data structures	element expressions in CP 57
for float and integers 37	arrays 8	

else	functions in ILOG Script	keywords (continued)
See if-then-else 47	miscellaneous 128	diff 44
else, IBM ILOG Script keyword 95	functions, numeric 102	dvar 32
equivalence, logical constraint 55		else 95
errors	•	float 4
tuples with same names 10	G	for 95
eval function 128	generic sets 21, 61	function 98
Excel	ground	if 95
accessing named ranges 25	breakpoints and slopes in	infinity 4, 37
Excel spreadsheet	piecewise-linear functions 39	inter 44
what data can be read 24	conditions in if-then-else	max 38
Excel, R1C1 style not supported 24, 26	statements 47	maxint 37
execute, IBM ILOG Script block	expressions and relations 36	min 38
for preprocessing 30	1	null 126
expressions 88		ordered 59
Boolean 45	Н	piecewise 39
counting 37		prod 38
float 37	hexadecimal numbers 100	return 98
in IBM ILOG Script for OPL 87		setof 12
in logical constraints 54	1	SheetConnection 24
integer 37	1	SheetRead 24
set 44	IBM ILOG Script	special 92
syntax 89	compound statements 87	static 98
external data 16	default values of statements 99	string 5
and memory allocation 24	syntax 87	sum 38
sets 21	identifiers 88	symdiff 44
	conventions 3	this 92, 120, 121
E	for data and variables 36	tuple 9
F	if-then-else	union 44 var 97
fail function 128	conditional constraints 47	
filtering	if, IBM ILOG Script keyword 95	with 27
formal parameters 59	implication	
in tuples of parameters 62	logical constraint (CP) 57	1
with constraints 47	logical constraint (CPLEX) 55	L
first, OPL function 44	implicit slicing 62	labeled assertions 50
float constraints 53	includeScript function 128	labeled constraints 48
float, OPL keyword 4	indexed labels 49	last, OPL function 44
floats 4	infinity 100	lazy initialization of data 16, 23, 30
expression 37	infinity, OPL keyword 4, 37	limitations on constraint labeling 51
functions in OPL 37	initializing	limitations on constraints 59
for, IBM ILOG Script keyword	arrays 17	limitations on tuples 11
loop syntax 95	arrays of decision variables 36	literals 88
forall, statements	data 15, 16	loadFile function 128
and constraint labels 51	set of tuples 16	logical constraints
formal parameters	sets 21	definition and extraction 54
basic 59	tuples 19	for CP 57
filter expressions 59	input/output	logical expressions 54
tuples 61	data to/from a spreadsheet 24	logical operators 114
function calls 91	integer constant maxint 4	loops
function definitions 98	integer expressions 37	in IBM ILOG Script 95
function, IBM ILOG Script keyword 98	integer ranges 7	
functions 127	integers 4	
card 44	inter, OPL keyword 44	M
first 44	internal data 16	max, OPL keyword 38
float 37	and memory allocation 23	maxint, OPL keyword 4, 37
for dates 125	sets 21	memory allocation and management
for strings 110	item function 44	and data initialization 23
item 44		memory consumption
last 44	17	and multidimensional arrays 8
next 44	K	methods
nextc 44	keys in tuple declarations 10	built-in, for objects 121
ord 44	keywords	defining for objects 120
over set expressions 44	arguments 92	for arrays 118
prev 44	assert 29	for Booleans 114
prevc 44	break 95	for dates 124
sign 41	continue 95	for functions 127
	-	

methods (continued)	P	SheetRead, OPL keyword 24
for numbers 102	-	shorthand operators 90
for strings 108	piecewise linear functions	sign function 41
for the null value 126	pwlFunction 5	slicing
for the undefined value 127	piecewise-linear functions 39	explicit/implicit 62
methods for sets 13	discontinuous 40	using key fields 10
min, OPL keyword 38	piecewise, OPL keyword 39	slopes in piecewise-linear functions 39
models	preprocessing	sorted sets 13
building 3	data 30	sorted tuple sets 14
conventions 3	prev, OPL function 44	sparsity
efficiency 48	prevc, OPL function 44	and multidimensional arrays 8
readability 61, 62	processing order	and one-dimensional arrays 8
multidimensional arrays 8	preprocessing items 30	and sets 21
,	prod, OPL keyword 38	special numbers 100
	program unit, and local variables	special ordered sets, not supported 12
N	(scripting) 97	specialized constraints 58
	properties	spreadsheets
NaN (Not-a-Number) 100	accessing value of 89	accessing named ranges 25
next, OPL function 44	for IBM ILOG Script arrays 117	connecting to 24
nextc, OPL function 44	for strings 108	data input/output 24
nonlinear constraints 53	pushing and pulling data 23	reading from 24
nonlinear expressions in logical	pwlFunction, OPL keyword 5	writing to 26
constraints 55		statements 95
Not-a-Number (NaN) 100		conditional 95
not, logical constraint 55	Q	in IBM ILOG Script for OPL 87
null value 126	<del></del>	last 87
null, IBM ILOG Script keyword 126	quotes 106, 113	static, IBM ILOG Script keyword 98
numbers 100		stepFunction, OPL keyword 6
automatic conversion to 101	<b>D</b>	stepwise functions
methods 102	R	stepFunction 6
numeric constants 103	R1C1 style not supported in Excel 24, 26	stop function 128
numeric functions 102	range float, data type 7	string constraints 54
numeric operators 104	ranges 7	9
	and set expressions 44	string, OPL keyword 5
	reading	strings 5, 106
0	from a spreadsheet 24	automatic conversion to 107
	return, IBM ILOG Script keyword 98	functions 110
objective function	,,	length 99
and decision variables 32		methods 108
objects 120	S	operators 112
constructor 121	_	properties 108
octal numbers 100	scheduling constraints in CP 58	structs in C, tuples in OPL 9
one-dimensional arrays 8	scope hiding 63	sum, OPL keyword 38
operations on sets 12	scope of script variables 97	symdiff, OPL keyword 44
operators	script functions	syntax, in IBM ILOG Script 87
aggregate 38	miscellaneous 128	accessing property values 89
assigning a value 90	script variables	assignment operators 90
for IBM ILOG Script dates 125	declaration 97	conditional statements 95
for strings 112	inside a function definition 97	default values 99
logical 114	outside a function definition 97	expressions 89
numeric 104	reference to 89	function definition 98
precedence 89	scope 97	identifiers 88
shorthand 90	semi-colon 87	loops 95
special 92	set expressions 44	other operators 93
syntax 93	and ranges 44	reference to a script variable 89
or, logical constraint 55	construction 44	shorthand 91
ord, OPL function 44	functions 44	special keywords 92
order	setof, OPL keyword 12	special operators 92
for processing script blocks 30	sets 12	variable declaration 97
ordered sets 12	allowed operations 12	
special ordered sets, not	and data consistency 27	-
supported 12	and sparsity 21	Т
ordered, OPL keyword 59	generic 21	then
overflow	initializing 21	See if-then-else 47
and integer expressions 37	methods for sets 13	this, IBM ILOG Script keyword 92, 120,
	of tuples, initialization 16	121
	ordered versus sorted 13	time representation in IBM ILOG
	SheetConnection, OPL keyword 24	Script 122
		1

```
tuple indices in dexpr 11
tuple indices in labeled constraints 12
tuple patterns
   in decision expressions 11
tuple sets
   external initialization 16
   referring to other sets with keys 28
   sorted 13
tuple sets, sorted 14
tuple, OPL keyword 9
tuples 9
  and data consistency 27
   data types allowed, not allowed 11
   duplicates in collections not
    detected 11
   initialization 19
   keys in declaration 10
   of parameters 61
      filtering 62
tuples, limitations 11
U
undefined value 126
union, OPL keyword 44
V
values
   and functions (IBM ILOG Script) 115
values, in IBM ILOG Script
   and functions
      Booleans 113
      dates 122
      decimal numbers 100
      hexadecimal numbers 100
      numbers 100
      objects 120
      special numbers 100
      strings 106
   assignment operators 90
   default 99
   null 126
   of properties, accessing 89
   undefined 126
values, in ILOG Script
   and functions
      octal numbers 100
var, IBM ILOG Script keyword 97
variables
   See script variables
W
while, IBM ILOG Script keyword
   loop syntax 95
with, IBM ILOG Script keyword 99
with, OPL keyword 27
```

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