

# Lecture 3-4

Pre-condition, Post-condition, Partial Correctness of an Algorithm, Total Correctness of Algorithm, and Illustrative Example





## *Pre-Condition, Post-Condition, Loop Invariant*

A **precondition** ( $P$ ) is a statement placed before the segment. It must be true prior to entering the segment for it to work correctly. **Preconditions are often placed either before loops or at the entry point of functions and procedures**

A **postcondition** ( $Q$ ) is a statement placed after the end of the segment. It should be true when the execution of the segment is complete. **Postconditions are often placed either after loops or at exit points of functions and procedures.**

An **invariant** ( $I$ ) is a statement placed in a code segment that is repeated, should be true each time the loop execution reaches that point. **Invariants are often used in a loops and recursions.**



## *Checking Loops*

- › The correctness of a loop can be ascertained by making the following sequence of checks:
  - Check that the **precondition** implies that the invariant is initially true.
  - Check that the **invariant** is preserved by the loop body.
  - Check that the **loop invariant**, together with the terminating condition, implies that the **postcondition** is true.
- › If the loop satisfies the above checks, then the **postcondition** must be true whenever the **precondition** is true, provided that the loop terminates. Since it is possible for a loop to run forever, one more check is needed:
  - › Check that the **loop terminates**.

# Proof of Correctness

How to Specify Computational Problem?





# Correctness Proof

- › A **computational problem** is specified by one (or more) pairs of *preconditions* and *postconditions*.
  - **Precondition**: A condition that one might expect to be satisfied when the execution of a program begins. This generally involves the algorithm's **inputs** as well as **initial values of global variables**.
  - **Postcondition**: A condition that one might want to be satisfied when the execution of a program ends. This might be
    - › A set of relationships between the values of **inputs** (and the values of global variables when execution started) and the **values of outputs** (and the values of global variables on a program's termination), or
    - › A description of **output generated**, or **exception(s) raised**.



## Pre-Condition and Post-Condition

- › **Precondition:** what's true before a block of code
- › **Postcondition:** what's true after a block of code
- › **Example:** computing the square root of a number
  - › Precondition: the input  $x$  is a positive real number
  - › Postcondition: the output is a number  $y$  such that  $y^2=x$



## Example

```
def factorial (n) :  
    '''  
    precondition:  n >= 0  
    postcondition:  return value equals n!  
    '''
```



## Enforcing preconditions

```
def factorial(n) :  
    '''  
    precondition:  n >= 0  
    postcondition: return value equals n!  
    '''  
    if (n < 0) : raise ValueError
```





# What about Post-conditions?

```
def factorial(n) :  
    '''  
    precondition:  n >= 0  
    postcondition: return value equals n!  
    '''  
    if (n < 0) : raise ValueError
```

- › `assert factorial(5)==120`
- › Can use **assertions** to verify that postconditions hold

An assertion is a **claim about the state of the program each time execution reaches a particular point in the program text (or step in the algorithm)**



# Loop invariants as a way of reasoning about the state of your program

<pre-condition:  $n > 0$

$i = 0$

while ( $i < n$ ) :

$i = i + 1$

<post-condition:  $i == n$ >

We want to prove the post-condition:  
 $i == n$  right after the loop



## Example: loop index value after a loop

```
// precondition:  $n \geq 0$ 
i = 0
//  $i \leq n$  loop invariant
while (i < n) :
    //  $i < n$  test passed
    // AND
    //  $i \leq n$  loop invariant
    i = i + 1
    //  $i \leq n$  loop invariant
//  $i \geq n$  WHY?
// AND
//  $i \leq n$ 
//  $\rightarrow i = n$ 
```

So we can conclude the obvious:

$i = n$  right after the loop

But what if the body were:

$i = i + 2$  ?



## Example: Search Problem Specification

- › *Precondition*  $P_1$ : Inputs include
  - $n$ : a positive integer
  - $A$ : an integer array of length  $n$ , with entries  
 $A[0], A[1], \dots, A[n-1]$
  - $key$ : An integer found in the array (i.e., such that  $A[i] = key$  for at least one integer  $i$  between  $0$  and  $n-1$ )
- › *Postcondition*  $Q_1$ :
  - Output is the integer  $i$  such that  $0 \leq i < n$ ,  $A[j] \neq key$  for every integer  $j$  such that  $0 \leq j < i$ , and such that  $A[i] = key$
  - Inputs (and other variables) have not changed
- › This describes what should happen for a “successful search.”



## Example: Search Problem Specification

### › Precondition $P_2$ : Inputs include

- $n$ : a positive integer
- $A$ : an integer array of length  $n$ , with entries  
 $A[0], A[1], \dots, A[n-1]$
- $key$ : An integer not found in the array (i.e., such that  $A[i] \neq key$  for every integer  $i$  between 0 and  $n-1$ )

### › Postcondition $Q_2$ :

- A *notFoundException* is thrown
  - Inputs (and other variables) have not changed
- This describes what should happen for an “unsuccessful search.”



## Examples: Pre-Condition and Post Condition

- › Let's start with a simple code example:
  - ›  $x = 17;$   
 $y = 42;$   
 $z = x + y;$
  - › We annotate the code to show this information:
  - ›  $\{ \text{true} \}$   
 $x = 17;$   
 $\{ x = 17 \}$   
 $y = 42;$   
 $\{ x = 17 \wedge y = 42 \}$   
 $z = x + y;$   
 $\{ x = 17 \wedge y = 42 \wedge z = 59 \}$   
 $\{ \text{true} \}$
- › An **assertion** is an assumption that something is  $\{\text{true}\}$ . An **assertion** is a logical formula inserted at some point in a program. There are two special assertions: the **precondition** and the **postcondition**.
- ›  $\{\text{true}\}$  is the **precondition**
- ›  $\{ x = 17 \wedge y = 42 \wedge z = 59 \}$  is the **postcondition**

## Additional Examples: Weakest Pre-Condition and Post Condition

```
x = x - 2;  
z = x + 1;  
{ z != 0 }
```

```
x = 2 * y;  
z = x + y;  
{ z > 0 }
```

```
w = 2 * w;  
z = -w;  
y = v + 1;  
x = min(y, z);  
{ x < 0 }
```

The solutions are:

```
{ x != -1 }  
x = x - 2;  
{ x != -1 }  
z = x + 1;  
{ z != 0 }
```

```
{ y > 0 }  
x = 2 * y;  
{ x + y > 0 }  
z = x + y;  
{ z > 0 }
```

```
{ v < -1  $\vee$  w > 0 }  
w = 2 * w;  
{ v < -1  $\vee$  w > 0 }  
z = -w;  
{ v < -1  $\vee$  z < 0 }  
y = v + 1;  
{ y < 0  $\vee$  z < 0 }  
x = min(y, z);  
{ x < 0 }
```



## Weakest Pre-Condition (IF – ELSE statement)

- ›  $wp(IF, Q)$ , We write the statement like  $if(B) S_1 \text{ else } S_2$
- › Suppose B is true. Because  $S_1$  is executed and Q must be true afterward, the weakest precondition for the entire IF statement will be the weakest precondition for  $S_1$  and Q, i.e.  $wp(S_1, Q)$
- › Analogously, if B is false the weakest precondition will be  $wp(S_2, Q)$
- › The weakest precondition for the entire if/else statement is  $wp(S_1, Q)$  when B is true and  $wp(S_2, Q)$  when B is false
- › 
$$wp(IF, Q) = (B \Rightarrow wp(S_1, Q) \wedge !B \Rightarrow wp(S_2, Q))$$
$$= (B \wedge wp(S_1, Q)) \vee (!B \wedge wp(S_2, Q))$$



## Example: Weakest Pre-Condition (IF – ELSE statement)

$$\begin{aligned} \triangleright \text{wp}(\text{IF}, Q) &= (B \Rightarrow \text{wp}(S_1, Q) \wedge !B \Rightarrow \text{wp}(S_2, Q)) \\ &= (B \wedge \text{wp}(S_1, Q)) \vee (!B \wedge \text{wp}(S_2, Q)) \end{aligned}$$

Using the formula above

- (1)  $\begin{array}{l} \text{if } (x < 5) \\ \quad x = x * x; \\ \text{else} \\ \quad x = x + 1; \\ \{ x \geq 9 \} \end{array}$
- $$\begin{aligned} \text{wp}(\text{IF}, x \geq 9) &= (x < 5 \wedge \text{wp}(x = x * x, x \geq 9)) \vee (x \geq 5 \wedge \text{wp}(x = x + 1, x \geq 9)) \\ &= (x < 5 \wedge x * x \geq 9) \vee (x \geq 5 \wedge x + 1 \geq 9) \\ &= (x \leq -3) \vee (x \geq 3 \wedge x < 5) \vee (x \geq 8) \end{aligned}$$
- (2)  $\begin{array}{l} \text{if } (x \neq 0) \\ \quad z = x; \\ \text{else} \\ \quad z = x + 1; \\ \{ z > 0 \} \end{array}$
- $$\begin{aligned} \text{wp}(\text{IF}, z > 0) &= (x \neq 0 \wedge \text{wp}(z = x, z > 0)) \vee (x == 0 \wedge \text{wp}(z = x + 1, z > 0)) \\ &= (x \neq 0 \wedge x > 0) \vee (x == 0 \wedge x + 1 > 0) \\ &= (x > 0) \vee (x == 0) \\ &= (x \geq 0) \end{aligned}$$



# Loop Invariant

- › A loop invariant is a statement about an algorithm's loop that:
  - is true before the first iteration of the loop and
  - if it's true before an iteration, then it remains true before the next iteration.
- › **If we can prove that those two conditions hold for a statement, then it follows that the statement will be true before each iteration of the loop.**

Let's say that we want to sum an array of real numbers

---

**Algorithm 1:** An algorithm to sum an array

---

**Data:**  $a = [a_1, a_2, \dots, a_n]$ : an array of  $n$  real numbers

**Result:** The sum of all elements of  $a$

$s \leftarrow 0$

**for**  $i = 1, 2, \dots, n$  **do**

$s \leftarrow s + a_i$

**end**

**return**  $s$

---

To prove working of algorithm, we must prove that algorithm works after the loop end, **s** is equal to the sum of the numbers in **a**.

$$s = \sum_{i=1}^n a_i$$



# Loop Invariant

- › At beginning of the  $i^{\text{th}}$  iteration,  $s$  is equal to the sum of the first  $i-1$  elements of  $a$ .
- › In this example, loop ends when  $i = n + 1$ , so invariant states that at the end of the loop

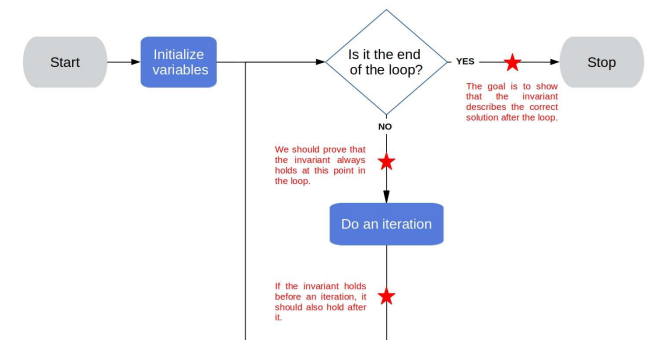
$$s = \sum_{i=1}^n a_i$$

- › The invariant hold before the first iteration corresponds to the base case of induction.
- › The second condition is similar to the inductive step.
- › But, unlike induction that goes on infinitely, a loop invariant needs to hold only until the loop has ended.



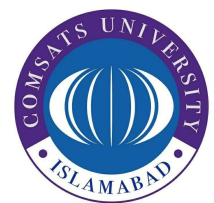
# Loop Invariant

- › For the given algorithm, we should prove loop invariant in two steps.
  - At the beginning of the loop,  $i=1$  and  $s=0$ . The sum  $\sum_{j=1}^0 a_j$  is the sum of no numbers. Thus, **loop invariant holds at the start of the  $i$ -th iteration**  $s = \sum_{j=1}^{i-1} a_j$
  - During the iteration, we add  $a_i$  to  $s$ , we get
$$s = a_i + \sum_{j=1}^{i-1} a_j = \sum_{j=1}^i a_j$$
The end of iteration  $i$  is the same as the beginning of the iteration  $i + 1$ , So the second condition is also satisfied.
- › As we have shown that  $s = \sum_{i=1}^n a_i$ , at the end of the loop, proving the invariant also verified and the algorithm was correct.



# Partial Correctness of An Algorithm

(1) Partial correctness – if the algorithm terminates then the output is guaranteed to be correct.





# Algorithm Correctness

## › *Partial Correctness*: If

- inputs satisfy the precondition  $P$ , and
- algorithm or program  $S$  is executed,  
then *either*
- $S$  halts and its inputs and outputs satisfy the postcondition  $Q$   
*or*
- $S$  does not halt, at all.

Generally written as

$$\{P\} \quad S \quad \{Q\}$$

**Note:** Detailed proofs rely heavily on discrete math and logic

To Prove Partial Correctness?



› Consider algorithm  $S$ :

– Divide  $S$  into sections  $S_1; S_2; \dots ; S_K$

- › assignment statements
- › Loops
- › control statements (i.e., if-then-else)
- › (other programming constructs)

– Identify intermediate assertions  $R_i$  so that

- ›  $\{P\} \ S_1 \ \{R_1\}$
- ›  $\{R_1\} \ S_2 \ \{R_2\}$
- .....
- ›  $\{R_{K-1}\} \ S_K \ \{Q\}$

– After proving each of these, we can then conclude that

- ›  $\{P\} \ S_1; S_2; \dots ; S_K \ \{Q\}$
- › equivalently,  $\{P\} \ S \ \{Q\}$



## Example: Proof of Partial Correctness

- › *Problem Definition:* Finding the largest entry in an integer array.

*Precondition P:* Inputs include

- $n$ : a positive integer
- $A$ : an integer array of length  $n$ , with entries  $A[0]; \dots ; A[n-1]$

*Postcondition Q:*

- Output is the integer  $i$  such that  $0 \leq i < n$ ,  $A[i] \geq A[j]$  for every integer  $j$  such that  $0 \leq j < n$
- Inputs (and other variables) have not changed

```
int FindMax(A, n)
  i = 0
  j = 1
  while (j < n) do
    if A[j] > A[i] then
      i = j
    end if
    j = j + 1
  end while
  return i
```



# Total Correctness of An Algorithm

Illustrative Examples





# Total Correctness of Algorithm

(2) Total correctness – the algorithm will terminate and produce the correct output for all possible inputs.

- › An algorithm is correct if for every legal input given to an algorithm, the **algorithm produces the correct output**. **For example**
  - **Max(a, b) is correct** if it always returns the larger value out of **a, b**.
  - **Max(int[] A) is correct** if it always returns the largest integer found in array A. That is, it returns **A[i]** such that **0 ≤ i < length(A)** and there is no **j** such that **0 ≤ j < length(A)** and **A[j] > A[i]**.



# Total Correctness of Algorithm

Example

```
int Maximum(int[] values)
{
    int max = values[0];

    for (int i = 1;
         i < values.Length;
         i++)
    {
        if (values[i] > max)
        {
            max = values[i];
        }
    }

    return max;
}
```



# Total Correctness of Algorithm

Example

```
int Maximum(int[] values)
{
    int max = values[0];

    for (int i = 1;
        i < values.Length;
        i++)
    {
        if (values[i] > max)
        {
            max = values[i];
        }
    }

    return max;
}
```

◀ Finds greatest value in the array



# Total Correctness of Algorithm

Example

```
int Maximum(int[] values)
{
    int max = values[0];

    for (int i = 1;
        i < values.Length;
        i++)
    {
        if (values[i] > max)
        {
            max = values[i];
        }
    }

    return max;
}
```

- ◀ Finds greatest value in the array  
Assume values array is non-null  
Assume values array contains at least one element



# Total Correctness of Algorithm

Example

```
int Maximum(int[] values)
{
    int max = values[0];

    for (int i = 1;
        i < values.Length;
        i++)
    {
        if (values[i] > max)
        {
            max = values[i];
        }
    }

    return max;
}
```

- ◀ **Finds greatest value in the array**  
Assume values array is non-null  
Assume values array contains at least one element
- ◀ **Can we offer a formal proof that this function works correctly?**  
Yes, we can offer a proof based on induction



# Total Correctness of Algorithm

Example

```
int Maximum(int[] values)
{
    int max = values[0];

    for (int i = 1;
        i < values.Length;
        i++)
    {
        if (values[i] > max)
        {
            max = values[i];
        }
    }

    return max;
}
```

$max \leftarrow \text{Max}\{k = 0 | \text{values}[k]\}$



# Total Correctness of Algorithm

Example

```
int Maximum(int[] values)
{
    int max = values[0];

    for (int i = 1;
         i < values.Length;
         i++)
    {
        if (values[i] > max)
        {
            max = values[i];
        }
    }

    return max;
}
```

$max \leftarrow \text{Max}\{k = 0 | \text{values}[k]\}$

Entire loop skipped if  
 $\text{values.Length} = 1$

$N = \text{values.Length}$

$max \leftarrow \text{Max}\{0 \leq k < N | \text{values}[k]\}$





# Total Correctness of Algorithm

Example

```
int Maximum(int[] values)
{
    int max = values[0];

    for (int i = 1;
        i < values.Length;
        i++)
    {
        if (values[i] > max)
        {
            max = values[i];
        }
    }

    return max;
}
```

$max \leftarrow \text{Max}\{k = 0 | \text{values}[k]\}$   
 $N = \text{values.Length} > 1$



# Total Correctness of Algorithm

Example

```
int Maximum(int[] values)
{
    int max = values[0];

    for (int i = 1;
         i < values.Length;
         i++)
    {
        if (values[i] > max)
        {
            max = values[i];
        }
    }

    return max;
}
```

$max \leftarrow \text{Max}\{k = 0 | \text{values}[k]\}$

$N = \text{values.Length} > 1$

Loop invariant (which must hold true):



# Total Correctness of Algorithm

Example

```
int Maximum(int[] values)
{
    int max = values[0];

    for (int i = 1;
        i < values.Length;
        i++)
    {
        if (values[i] > max)
        {
            max = values[i];
        }
    }

    return max;
}
```

$max \leftarrow \text{Max}\{k = 0 | \text{values}[k]\}$

$N = \text{values.Length} > 1$

Loop invariant (which must hold true):

$max \leftarrow \text{Max}\{0 \leq k < i | \text{values}[k]\}$

True when we enter the loop for  $i = 1$



# Total Correctness of Algorithm

Example

```
int Maximum(int[] values)
{
    int max = values[0];

    for (int i = 1;
        i < values.Length;
        i++)
    {
        if (values[i] > max)
        {
            max = values[i];
        }
    }

    return max;
}
```

$max \leftarrow \text{Max}\{k = 0 | \text{values}[k]\}$

$N = \text{values.Length} > 1$

Loop invariant (which must hold true)

$max \leftarrow \text{Max}\{0 \leq k < i | \text{values}[k]\}$

True when we enter the loop for  $i = 1$

$max \leftarrow \text{Max}\{0 \leq k < i + 1 | \text{values}[k]\}$



# Total Correctness of Algorithm

Example

```
int Maximum(int[] values)
{
    int max = values[0];

    for (int i = 1;
        i < values.Length;
        i++)
    {
        if (values[i] > max)
        {
            max = values[i];
        }
    }

    return max;
}
```

$max \leftarrow \text{Max}\{k = 0 | \text{values}[k]\}$

$N = \text{values.Length} > 1$

Loop invariant (which must hold true):

$max \leftarrow \text{Max}\{0 \leq k < i | \text{values}[k]\}$

True when we enter the loop for  $i = 1$

$max \leftarrow \text{Max}\{0 \leq k < i + 1 | \text{values}[k]\}$

$N = \text{values.Length}$

$max \leftarrow \text{Max}\{0 \leq k < N | \text{values}[k]\}$

# Thank You!!!

Have a good day

