Enhancements to an OO Metric: CB Measure

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Abstract: Due to the wide usage of the object-oriented paradigm as a development paradigm many researches have proposed metrics to measure the complexity of object-oriented programs. The proposed object-oriented metrics can be divided into two categories based on the main aspect they have considered: metrics based on object-oriented aspects and metrics based on the cognitive aspects. Majority of the metrics which belong to the latter category have relied on a maximum of three complexity factors to derive the complexity of a program. CB measure is one of the few metrics that has considered four or more complexity factors to measure the complexity associated with a software program. However, there exists some other factors that could be considered by the CB measure to make it a more practically applicable measure. Such factors were proposed by the authors in a previous study. This paper demonstrates how those factors can be incorporated to the CB measure. In addition, it validates the practical applicability of the modified CB measure.

Key words: Object-oriented paradigm, software complexity, complexity metrics, CB measure

1. Introduction

Towards the later part of the 1980s, the introduction of object-oriented (OO) paradigm brought about a different means of looking at programming with the use of objects. Objects are the instances of classes. They can consist of fields and methods. Fields are used to store data of objects and methods are used to perform the actions of them.

The wide usage of the object-oriented paradigm as a development paradigm led the way for researches to propose metrics to measure the complexity of object-oriented programs. The proposed object-oriented metrics can be mainly divided into two categories based on the aspects they have considered: metrics based on object-oriented aspects and metrics based on the cognitive aspects. Chidamber and Kemerers' suite of six metrics [1], Chen and Luis' suite of eight metrics [2], MOOD metrics [3], and Li's suite of six metrics [4] are some of the OO metrics which were proposed based on the OO aspect. Class complexity measure [5], weighted class complexity measure [6], cognitive code complexity measure [7], CB measure [8] are some of the OO metrics which were proposed based on the cognitive aspect.

Majority of the cognitive based complexity metrics have relied on a maximum of three complexity factors to compute the complexity of an entire program. CB measure is one of the few metrics that has considered four or more complexity factors to measure the complexity associated with a software program. However, there exists some other factors that could be considered by the CB measure to make it a more practically applicable measure. A previous study proposed some of such factors that could be incorporated to the CB

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measure [9]. This study demonstrates how those factors can be incorporated to the CB measure. Additionally, it validates the practical applicability of the modified CB measure.

The next section discusses about the CB measure. Section 3 discusses how the previously proposed factors are accommodated in the modified CB measure. Section 4 discusses about the practical applicability of the modified CB measure. Section 5 concludes the paper.

2. CB Measure

Chhillar and Bhasin believed that software complexity is a multidimensional attribute of software and thereby it cannot be measured by considering a single factor. With this in mind, they proposed the CB measure in 2011. It was based on four significant factors that contribute to the complexity of software:

Inheritance level of classes (Wi): The degree of understanding a statement increases with the level of inheritance of classes. Taking this into account Chhillar and Bhasin assigned a weight of zero for executable statements in the bass class, a weight of one for the executable statements which are at the first derived class, 2 for the statements at the next derived class. Similarly, the weight allocated for the statements increased by one for each derived class.

Type of control structures in classes (Wc): Chhillar and Bhasin believed that the complexity added to a class/program by a control structure differs depending on its type. Thus, a weight of zero was assigned to sequential statements, one for conditional control structures such as if-else conditions, two for iterative control structures such as for, while, and do-while loops, and n for switch-case statements with n cases.

Nesting level of control Structures (Wn): The understandability of a program decreases with the number of nesting levels of control structures. As a result, the complexity of that program will increase. Taking this into consideration Chhillar and Bhasin assigned a weight of zero for sequential statements, one for statements which are at the outer most level of nesting, two for statements which are at the next inner level of nesting. Similarly, the weight allocated for the statements was increased by one for each level of nesting.

Size of a class in terms of token count: With the belief that the complexity of a class or program increases along with the size of it, Chhillar and Bhasin considered the size of a class or program to be the final factor of their measure. The size of an executable statement was calculated in terms of the operators, operands, methods/functions, and strings in that statement.

Considering the above-mentioned factors, a weighted complexity measure for an object-oriented program P was suggested as:

$$C_w(P) = \sum_{j=1}^{n} (S_j) * (W_t)_j$$

where

 $C_w(P)$ = Weighted complexity measure of program P

 S_i = Size of j^{th} executable statement in terms of tokens count

n = Total number of executable statements in program P

i = Index variable

 $W_t = W_n + W_i + W_c$

3. Accommodating the New Factors

In a previous work [9], the authors presented additional factors that could be considered by the CB measure to make its complexity calculation more accurate. The present paper demonstrates how the CB measure can be modified to incorporate those factors. Thus, in addition to considering the complexities that arises due to inheritance level, type and nesting level of control structures and size, modified CB measure

(MCB) considers the complexity that arises due to dynamic memory access, recursive methods, threads, pointers and references, exceptions, compound conditional statements.

As there is a minor difference between dynamic memory access and pointers and references, with regards to regular data allocation and variable retrieval, they can be considered in an equivalent manner in the MCB formula by adding a constant value of two for the *, & and new operators. These can be considered by the size (S) attribute. In languages such as C++ when the memory is dynamically allocated using the "new" operator it should be deallocated using the "delete" operator. Thus, when a "delete" operator is present in a statement, like the other operators mentioned above, a value of two is added when computing values for the size (S) attribute. Fig. 1 shows a sample program which was written to demonstrate the effect that dynamic memory access and pointers and references have on the complexity of a program. Table 1 and Table 2 clearly demonstrates how complexity is calculated for the program given in Fig. 1, using the CB and MCB measures.

```
#include <iostream>
using namespace std;
void allocate()
int * p1 ;
int * p2;
void allocate() {
 int number = 88;
int *p1 = &number;
 p2 = new int(99);
int main() {
 allocate();
  cout <<"p1 value :"<< p1;
  cout << "p2 value : "<<*p2;
  delete p1;
 delete p2;
  return 0;
```

Fig. 1. Sample program to demonstrate the effect of dynamic memory access and pointers and references.

Executable Statement	S	Wn	Wi	Wc	W	S*W
void allocate()	2	0	1	0	1	2
int number = 88	4	0	1	0	1	4
int *p1 = &number	6	0	1	0	1	6
p2 = new int(99)	4	0	1	0	1	4
int main()	2	0	1	0	1	2
allocate()	1	0	1	0	1	1
cout<<"p1 value :"<< p1	5	0	1	0	1	5
cout<< "p2 value :"<<*p2	6	0	1	0	1	6
delete p1	2	0	1	0	1	2
delete p2	2	0	1	0	1	2
return 0	1	0	1	0	1	1
CB value						35

Table 1. Complexity Caculation of CB for Allocate Program

Table 2. Complexity Caculation of MCB for Allocate Program

Executable Statement	S	Wn	Wi	Wc	W	S*W
void allocate()	2	0	1	0	1	2
int number = 88	4	0	1	0	1	4
int *p1 = &number	8	0	1	0	1	8
p2 = new int(99)	5	0	1	0	1	5
int main()	2	0	1	0	1	2
allocate()	1	0	1	0	1	1
cout<<"p1 value :"<< p1	5	0	1	0	1	5
cout<< "p2 value :"<<*p2	7	0	1	0	1	7
delete p1	3	0	1	0	1	3
delete p2	3	0	1	0	1	3
return 0	1	0	1	0	1	1
MCB value						41

The complexity that arises due to the usage of threads is considered in the MCB measure by adding a constant value of two to the total size (S) value, for statements with a thread invocation. Fig. 2 shows a sample program which was written to demonstrate the effect that threads has on the complexity of a program. Table 3 and Table 4 clearly demonstrates how complexity is calculated for the program given in Fig. 2, using the CB and MCB measures.

```
public class MyThread implements Runnable{
  public void run() {
    System.out.println("Thread running");
  }
  public static void main(String args[]) {
    MyThread mythread = new MyThread();
    Thread thread = new Thread(mythread);
    thread.start();
  }
}
```

Fig. 2. Sample program to demonstrate the effect of threads.

Table 3. Complexity Caculation of CB for MyThread Program

Executable Statement	S	Wn	Wi	Wc	W	S*W
public void run()	2	0	1	0	1	2
System.out.println("Thread running")	6	0	1	0	1	6
public static void main (String args[])	4	0	1	0	1	4
MyThread mythread = new MyThread()	5	0	1	0	1	5
Thread thread = new Thread(mythread)	5	0	1	0	1	5
thread.start()	3	0	1	0	1	3
CB Value						25

Table 4. Complexity Caculation of MCB for MyThread Program

Executable Statement		Wn	Wi	Wc	W	S*W
public void run()	2	0	1	0	1	2
System.out.println("Thread running")	6	0	1	0	1	6
public static void main (String args[])	4	0	1	0	1	4
MyThread mythread = new MyThread()	6	0	1	0	1	6
Thread thread = new Thread(mythread)	8	0	1	0	1	8
thread.start()	5	0	1	0	1	5
MCB Value						31

During exception handling the complexity of handling all the predicates of a try-catch() statement and throws have to be considered. Since throwing of exceptions without try-catch statements is often in line with class declarations, it will be considered with the use of the size (S) attribute by adding a value of one for the "throw" keyword. However, when try-catch() statements are involved, the catch blocks will act as the predicates for that type of exception. Therefore, each catch block can be considered as a branch in a conditional statement. Thus, this can be considered with the use of the weight due to type of control structure (Wc) attribute by assigning a weight of one for each catch block. Any exception that is thrown by a function can only be thrown once by a program for a given instance. Therefore, if the same exception is caught in several statements, only the first statement which catches that exception is considered. Fig. 3 shows a sample program which was written to demonstrate the effect that exceptions has on the complexity of a program. Table 5 and Table 6 clearly demonstrates how complexity is calculated for the program given in Fig. 3, using the CB and MCB measures.

```
public class Test{
  int a, b, c;

public void output1(){
    | System.out.println("Output1");
}

public void output2() {
    | System.out.println("Output2");
}

public static void main(String args[]) {
    | Test t1 = new Test();
    | Test t2 = new Test();

    try{
        | t1.output1();
    }

    catch (Exception e) {
        | System.out.println("Error");
    }
    t1.output2();
}
```

Fig. 3. Sample program to demonstrate the effect of exceptions.

Table 5. Complexity	Caculation of CB	for Test Program
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Executable Statement	S	Wn	Wi	Wc	W	S*W
public void output1()	2	0	1	0	1	2
System.out.println("Output1")	6	0	1	0	1	6
public void output2()	2	0	1	0	1	2
System.out.println("Output2")	6	0	1	0	1	6
public static void main(String args[])	4	0	1	0	1	4
Test t1 = new Test()	5	0	1	0	1	5
t1.output1()	3	0	1	0	1	3
catch (Exception e)	1	0	1	0	1	1
System.out.println("Error")	6	0	1	0	1	6
t1.output2()	3	0	1	0	1	3
CB Value						38

Table 6. Complexity Caculation of MCB for Test Program

Executable Statement	S	Wn	Wi	Wc	W	S*W
public void output1()	2	0	1	0	1	2
System.out.println("Output1")	6	0	1	0	1	6

public void output2()	2	0	1	0	1	2
System.out.println("Output2")	6	0	1	0	1	6
public static void main(String args[])	4	0	1	0	1	4
Test t1 = new Test()	6	0	1	0	1	6
t1.output1()	3	0	1	0	1	3
catch (Exception e)	1	0	1	1	2	2
System.out.println("Error")	6	0	1	0	1	6
t1.output2()	3	0	1	0	1	3
MCB Value						40

To accommodate the impact of recursive methods in the MCB formula, initially the total CB value for the entire program is calculated in the usual manner. Next, the addition of S*W value of the statements belonging to the recursive function is added to the computed total CB value. A calculation approach such as this was suggested to differentiate the complexity added by recursive functions, based on their content. Fig. 4 shows a sample program which was written to demonstrate the effect that recursive methods has on the complexity of a program. Table 7 and Table 8 clearly demonstrates how complexity is calculated for the program given in Fig. 4, using the CB and MCB measures.

```
public class Factorial{
  int a,b,c;

public int fact(int n) {
    if (n == 0)
        return 1;
    else
        return (n * fact(n-1));
    }

public static void main(String args[]) {
        Factorial t1 = new Factorial();
        int f = 0;
        f = t1.fact(5);
        System.out.println(f);
    }
}
```

Fig. 4. Sample program to demonstrate the effect of recursion.

Table 7. Com	plexity	Caculation	of CB f	or Factorial	Program

Executable Statement	S	Wn	Wi	Wc	W	S*W
public int fact(int n)	2	0	1	0	1	2
if (n==0)	4	1	1	1	3	12
return 1	1	1	1	0	2	2
return (n * fact(n-1))	3	1	1	0	2	6
public static void main()	4	0	1	0	1	4
Factorial t1 = new Factorial()	5	0	1	0	1	5
int f = 0	4	0	1	0	1	4
f = t1.fact(10)	5	0	1	0	1	5
System.out.println(f)	5	0	1	0	1	5
CB Value						45

Table 8. Complexity Caculation of MCB for Factorial Program

Executable Statement	S	Wn	Wi	Wc	W	S*W
public int fact(int n)	2	0	1	0	1	2
if (n==0)	4	1	1	1	3	12

return 1	1	1	1	0	2	2
return (n * fact(n-1))	3	1	1	0	2	6
public static void main()	4	0	1	0	1	4
Factorial t1 = new Factorial()	5	0	1	0	1	5
int $f = 0$	4	0	1	0	1	4
f = t1.fact(10)	5	0	1	0	1	5
System.out.println(f)	5	0	1	0	1	5
MCB Value						65

To differentiate the complexity that arises due to the usage of simple and compound conditions, a weight of one was assigned to the weight due to type of control structure attribute (Wc), whenever the "&&" or "||" operator was used in a decisional statement to concatenate two or more conditions. Fig. 5 shows a sample program which was written to demonstrate the effect that compound conditions has on the complexity of a program. Table 9 and Table 10 clearly demonstrates how complexity is calculated for the program given in Fig. 5, using the CB and MCB measures.

Fig. 5. Sample program to demonstrate the effect of compound conditions.

Table 9. Complexity	Caculation o	t CR t	or Re	sult P	rogra	ım

Executable Statement	S	Wn	Wi	Wc	W	S*W
public void res (int marks)	2	0	1	0	1	2
if (marks >0 && marks <50)	8	1	1	1	3	24
System.out.println("Fail")	6	1	1	0	2	12
System.out.println("Pass")	6	1	1	0	2	12
public static void main (String args[])	4	0	1	0	1	4
Result r = new Result()	5	0	1	0	1	5
r.res(50)	3	0	1	0	1	3
CB Value						62

Tabel 10. Complexity Caculation of MCB for Result Program

Executable Statement	S	Wn	Wi	Wc	W	S*W
public void res(int marks)	2	0	1	0	1	2
if (marks >0 && marks <50)	8	1	1	2	4	32
System.out.println("Fail")	6	1	1	0	2	12
System.out.println("Pass")	6	1	1	0	2	12
public static void main(String args[])	4	0	1	0	1	4
Result r = new Result()	5	0	1	0	1	5

MCB Value						70
r.res(50)	3	0	1	0	1	3

4. Experiments and Results

To find the practical applicability of the CB measure and the MCB measure, as explained in [10], an empirical study was conducted using thirty software engineers with two or more years of working experience and five open source java programs. Initially, the software engineers were asked to rank five programs that were given to them on scale of 1-5 by giving one to the program with the least complexity and five to the program with the highest complexity. From the obtained ranks, an average value was calculated for each program. Table 13 shows the ranking of the software engineers for the five programs.

Next, complexities of the five programs were calculated using the CB and MCB measures. Table 11 and Table 12 shows the CB and MCB values derived for the five programs. Subsequently, the five programs were ranked based on the complexity values derived for the two measures. Table 13 shows the ranking of the five programs for the two measures.

Finally, Spearman's rank correlation coefficient between the two measures and the ranking of the software engineers was used to find out the most practically applicable measure.

Table. 11. CB Values Of The Five Java Programs

Program	CB Value
Class2	142
FactMain	84
FibonacciMain	96
ListStack	95
BubbleSort	257

Table. 12. MCB Values of the Five Java Programs

Program	MCB Value
Class2	142
FactMain	108
FibonacciMain	148
ListStack	154
BubbleSort	259

Table. 13. Comparison of Ranks Obtained

Programs	Ranks of Software	Ranks of CB	Ranks of MCB
Class2	1	4	2
FactMain	2	1	1
FibonacciMain	3	3	3
ListStack	4	2	4
BubbleSort	5	5	5

As demonstrated in Table 14, it can be observed that, from the two measures, the MCB measure has the highest correlation coefficient with the rankings of the software engineers. CB measure has a correlation coefficient of 0.300 and MCB measure has a correlation coefficient of 0.900, with the rankings of the software engineers. Thus, it can be concluded that MCB measure is practically more applicable than the CB measure.

Table. 14. Spearman's Correlation Coefficient between Complexity Measures and Expert Rankings

			Ranking of Software Engineers
	Ranking of Software Engineers	Correlation Coefficient	1.000
		Sig. (2-tailed)	
		N	5
Spearman's rho CB Rankings	Correlation Coefficient	.300	
	nan's rho CB Rankings	Sig. (2-tailed)	.624
1		N	5
		Correlation Coefficient	.900*
	MCB Rankings	Sig. (2-tailed)	.037
		N	5
*. Correlation is si	gnificant at the 0.05 level (2-tailed).	•

5. Conclusion

Majority of the OO complexity metrics that has been proposed based on the cognitive aspect have relied on a maximum of three complexity factors to derive the complexity of a program. CB measure is one of the few metrics that has considered four or more complexity factors to measure the complexity associated with a software program. It considers the complexity that arises due to inheritance level of statements, type and nesting level of control structures and size of the program. In a previous study [9] authors propose some other factors that could be considered by the CB measure to make its complexity calculation more accurate. This paper demonstrates how those factors can be accommodated to the CB measure. It also validates the practical applicability of the MCB measure.

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