

Measure Performance of VRS Model using Simulation Approach by Comparing COCOMO Intermediate Model in Software Engineering

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Abstract— Estimation of a software project failure is based on project analogies in the area of software cost estimation. In all software cost estimation approaches important decisions must be made regarding certain approaches in order to obtain reliable estimate. A reliable and accurate estimate can be proved by simulation approach. A simulation modeling approach is proposed for the prediction of software effort productivity indices, such as cost and time-to-market, and the sensitivity analysis of such indices to changes in the organization parameters and user requirements. Simulation provides a controlled environment for evaluating performance. Changes can be made easily to the model and results evaluated quickly. In this paper we measure performance of VRS model through simulation result by comparing standard COCOMO Intermediate-Organic Model. This approach uses a model specification. Results demonstrate the model representativeness, and its usefulness in verifying process conformance to expectations, and in performing continuous process improvement and optimization.

Keywords- Business Process Outsourcing (BPO), Kilo Line of code (KLOC), Constructive Cost Model (COCOMO), Interactive Voice Response (IVR), Voice Response Software (VRS), Effort Adjustment Factor (EAF), Person (P), Month (M).

I. INTRODUCTION

Simulation is a powerful tool for solving many problems. A Software MAT LAB used as simulator tool which will help in taking a decision on accuracy of VRS Model [1] to estimate the effort and time for development of IVR software. The cost and time estimates are useful for the initial validation and the monitoring of the project's progress during development [2]. These estimates may be useful for project productivity assessment after completion. Reducing the cost of large scale software projects and shortening cycle time, or time to market, is a major goal of most software development organizations. To pursue such a goal, organizations can set productivity goals for each project, and put in place statistical productivity controls to enable developers and management to take corrective actions when there are deviations from the goal, and to distinguish a random deviation from meaningful deviations. Simulation is one of the methods for performing such control. It can be used at various points in the software life cycle to perform risk analysis, in terms of time to product, and cost, to verify conformance to expectations, and to perform continuous process improvement and optimization. Simulation is a formal and robust technique [3]. It does not rely heavily on mathematical abstraction therefore it is suitable for modeling even complicated environments. Simulation is basically a numerical technique therefore it can be used to generate quantitative output data on various parameters that influence a system performance. Output data analysis, experimental design can be employed to ensure a significant degree of mathematical robustness at every stage of a simulation project. This requires that organizations use metrics and models to evaluate and predict effort and time duration as a function of the user requirement size (in LOC), and the organization parameters [4]. The model is parameterized on the basis of measurements and analyses of data coming from IVR software [1]. To meet the modeling and simulation requirements of a particular environment, appropriate simulation tool needs to be selected with the considerations of simulation performance such as efficiency, accuracy, and speed, etc [6]. Popular existing packages are MATLAB which is a PC based system simulation tool with medium to high system complexity, and spreadsheets software. A large number of simulations can be performed situation and gradually refine the model as our understanding of the process improves, which enable us to achieve a good accuracy approximation. Simulation is based on traditional queuing models, using statistical and experimental methods to generate an internal picture of the system from which the statistical data is gathered for performance analysis. The simulation package statically store the simulation configurations, simulation results, results analysis, and dynamically keeps tracks of the parameter modifications of the simulation data set. The initial parameters of a given problem are entered manually. The simulation models and configurations are stored in the input data set file, and from the file, the program extract the whole set of information as the input of the simulation program

[5]. The outputs of the simulation program are stored in the output file in order to conduct analysis, query or visualization.

The remainder of this paper is structured as follows: The second section of the paper describes the existing model of proposed work. The third section describes about research methodology. The fourth describes the simulation result process activities and the fifth section synthetically illustrates the conclusion of simulation experiments.

II. EXISTING WORK

This section provides some information of COCOMO - software effort estimation models which is used in this research work. The COCOMO model is supposed to be standard model to calculate the effort and development time of software projects.

A. COCOMO Basic Model

COCOMO Basic model is proposed by B.W.Boehm. COCOMO model have three sub-models i.e. basic, intermediate and detailed model [11], [12] .

Development Mode	Basic Effort Equation	Time Duration (D)
Organic	Effort = 2.4 KLOC ^{1.05} PM	D = 2.5 * (Effort) ^{0.38} M
Semi Detached	Effort = 3.0 KLOC ^{1.12} PM	D = 2.5 * (Effort) ^{0.35} M
Embedded	Effort = 3.6 KLOC ^{1.20} PM	D = 2.5 * (Effort) ^{0.32} M

B. COCOMO Intermediate Model

The Intermediate COCOMO is an extension of the basic COCOMO model. Here we use the same basic equation for the model. But coefficients are slightly different for the effort equation [11], [12].

Development Mode	Intermediate Effort Equation	Time Duration (D)
Organic:	Effort = EAF * 3.2 * (KLOC) ^{1.05} PM	D = 2.5 * (Effort) ^{0.38} M
Semi Detached	Effort = EAF * 3.0 * (KLOC) ^{1.12} PM	D = 2.5 * (Effort) ^{0.35} M
Embedded	Effort = EAF * 2.8 * (KLOC) ^{1.20} PM	D = 2.5 * (Effort) ^{0.32} M

C. COCOMO II model

This model is an extension of COCOMO Intermediate model and specific one among COCOMO-II series .It has 17 cost drivers [11], [12].

$$\text{EFFORT} = \text{EAF} * 2.9 \text{ KLOC}^{1.10}$$

$$\text{Duration} = 3.67 * (\text{Effort})^{0.31}$$

D. SEL – Model

The Software Engineering Laboratory (SEL) of the University of Maryland has established a model i.e. SEL Mode[11][12] for estimation. Estimation of effort according to SEL model is defined as follows.

$$\text{EFFORT} = 1.4 * (\text{Size})^{0.93}$$

$$\text{Duration D} = 4.6 \text{ KLOC}^{0.26}$$

Effort (Person-Months) and lines of code (size in thousands of lines of code i.e. KLOC) are used as predictors.

E. Walston-Felix Model

Walston and Felix had developed their model in 1977 to estimate effort from a various aspects of the software development environment such as sixty projects collected in IBM's Federal Systems division. It provides a relationship between delivered lines of source code. This model constitutes participation, customer-oriented changes, memory constraints etc. According to Walston and Felix model, effort is computed by [11][12] :-

$$\text{EFFORT} = 5.2 \text{ KLOC}^{0.91}$$

$$\text{Duration D} = 4.1 \text{ KLOC}^{0.36}$$

F. Bailey-Basil Model

This model developed by Bailey-Basil between delivered lines of source code and formulates a relation [11][12]

$$\text{EFFORT} = 5.5 \text{ KLOC}^{1.16}$$

E. Halstead Model

This model developed by Halstead between delivered lines of source code and formulates a relation [11][12]

$$\text{EFFORT} = 0.7 \text{ KLOC}^{1.50}$$

G. Doty (for KLOC > 9)

This model developed by Doty between delivered lines of source code and formulates a relation [11][12]

$$\text{EFFORT} = 5.288 \text{ KLOC}^{1.047}$$

III. PROPOSED WORK

Simulation is defined as the process of creating a model of an existing or proposed system e.g. a project in order to identify and understand those factors which control the system and/or to predict the future behavior of the system. The model is simulated and many experiments are carried out. We begin simulations with a simple approximation of the situation and gradually refine the model as our understanding of the process improves, which enable us to achieve a good accuracy approximation. The client / customer demands cost and time duration before development the software. In such way VRS model will be useful to predict the estimated effort and time duration of IVR software. We took seventeen assumed dataset in terms of line of code metric as shown in table 1 to observe the simulation results using MATLAB version 2.0. The COCOMO Intermediate organic model is supposed to be standard model to estimate the reliable effort and schedule. The author found that the parameter values of VRS model are close to COCOMO Intermediate -organic Model therefore the VRS-model is being compared with COCOMO intermediate organic model to represent the simulation result in graphical way.

VRS Model:

$$\begin{aligned} \text{Initial Effort } E &= 3.4 * (\text{Project Size})^{1.15} , \\ \text{Time Duration } D &= 2.2 * (\text{Effort})^{0.31} \end{aligned}$$

COCOMO Intermediate -organic Model

$$\begin{aligned} \text{Initial Effort} &= 3.2 * (\text{Project size})^{1.05} , \\ \text{Time Duration} &= 2.50 * (\text{Effort})^{0.38} \end{aligned}$$

Life Cycle of IVR Software

A Systems development life cycle (SDLC) is a process that includes Software Analysis, Requirements, Software Design, Software Coding, implementation, testing and installation, call testing, and operation phase.

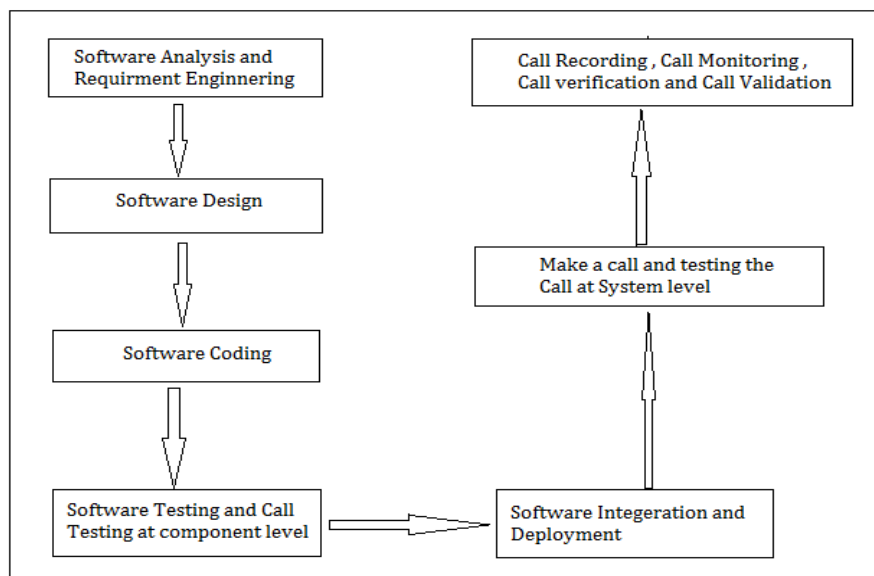


Figure 1: IVR software life cycle

Comparison of parameter value

Model	a	b	c	d
VRS Model	3.4	1.15	2.2	0.31
CoCoMo Intermediate Model	3.2	1.05	2.5	0.38

Here it is observed that parameter value of VRS Model is slightly vary than CoCoMo intermediate Model due to phases of IVR life cycle as shown in the Figure 1 , an additional phase i.e. call testing is added than tradition software life cycle.

Table-1: Calculated data used to get simulated result ('c: \data_simulation.xls)

Sr. No	Project Size(KLOC)	Effort(PM) (VRS Model)	Duration (M) (VRS Model)	Effort (PM) (CoCoMo)	Duration (M) (CoCoMo)
1	2.2	8.419087	4.258482	7.32308	5.327502
2	2.4	9.305117	4.392648	8.023646	5.515708
3	2.5	9.752364	4.457042	8.375041	5.606284
4	2.6	10.202304	4.51979	8.727141	5.865608
5	2.7	11.10991	4.640801	9.433349	5.865608
6	2.8	11.10991	4.640801	9.433349	5.865608
7	3.0	12.02730	4.756361	10.142086	6.029320
8	3.1	12.48949	4.812287	10.497351	6.108721
9	3.3	13.42053	4.920749	11.209587	6.263023
10	3.5	14.36009	5.025062	11.92399	6.411803
11	3.6	14.83292	5.075782	12.28196	6.484279
12	3.8	15.78447	5.174567	12.99938	6.625683
13	4.0	16.74356	5.270059	13.71870	6.762681
14	4.2	17.70987	5.362527	14.43981	6.895620
15	4.3	18.19565	5.407700	14.80102	6.960667
16	4.6	19.66303	5.539293	15.88714	7.150516
17	4.7	20.1554	5.581925	16.24998	7.212138
18	4.8	20.64935	5.623978	16.6132	7.272978
19	5.2	22.64033	5.786772	18.06981	7.509004
20	5.4	23.64459	5.865155	18.80024	7.622933
21	5.5	24.14883	5.903648	19.16597	7.678948
22	5.7	25.16141	5.979303	19.89842	7.789168
23	5.8	25.66972	6.01649	20.26513	7.843407
24	5.9	26.17934	6.053268	20.63215	7.897088
25	6.0	26.69026	6.089646	20.99949	7.950223
26	6.1	27.202462	6.125636	21.36713	8.002830
27	6.2	27.71592	6.161249	21.73508	8.054921
28	6.3	28.23063	6.196494	22.10332	8.106509
29	6.4	28.74656	6.231381	22.47186	8.157608
30	6.5	29.2637	6.265919	22.84068	8.208228
31	6.6	29.78204	6.300116	23.20978	8.258383
32	6.7	30.30156	6.333982	23.57917	8.308083
33	6.8	30.82224	6.367524	23.94883	8.357343
34	7.1	32.39112	6.466284	25.05943	8.502548
35	7.2	32.91632	6.498606	25.43016	8.550129
36	7.4	33.96999	6.562393	26.17238	8.644113
37	7.5	34.49843	6.593871	26.54387	8.690533
38	7.6	35.02793	6.625080	26.915611	8.736583
39	7.7	35.55848	6.656027	27.287593	8.782270
40	7.8	36.09007	6.686716	27.65982	8.827602
41	8.0	37.15629	6.839351	28.40497	8.917228
42	8.1	37.69091	6.777289	28.77790	8.961537
43	8.2	38.22652	6.807000	29.151071	9.005518
44	8.3	38.76312	6.836479	29.52446	9.049179
45	8.4	39.30068	6.86573	29.89808	9.092524

46	8.5	39.83921	6.894757	30.27191	9.13556
47	8.6	40.37868	6.923566	30.64597	9.178292
48	8.7	40.91913	6.95216	31.02024	9.220727
49	8.9	42.00272	7.008719	31.76943	9.304726
50	9.3	44.18087	7.119432	33.27033	9.469383
51	9.5	45.27527	7.173641	34.02199	9.550118
52	9.6	45.82377	7.20047	34.39813	9.590102
53	9.7	46.37313	7.22712	34.77445	9.629837
54	9.8	46.92334	7.253594	35.15097	9.669326
55	10.7	51.9123	7.484391	38.5481	10.01431
56	11.0	53.5896	7.558535	39.68371	10.12541
57	11.6	56.96467	7.703009	41.95955	10.34227
58	11.8	58.0956	7.750096	42.71949	10.41305
59	12.1	59.79737	7.819773	43.8606	10.51788
60	12.2	60.36604	7.842751	44.24129	10.55248
61	12.5	62.07624	7.910967	45.38428	10.65526
62	12.7	63.21981	7.955861	46.14704	10.72296
63	13.0	64.94022	8.022356	47.29231	10.82332
64	14.4	73.04589	8.320269	52.65391	11.27415
65	14.7	74.79867	8.381655	53.80631	11.36728
66	14.8	75.38412	8.401938	54.1907	11.39807
67	15.8	81.27081	8.600078	58.04167	11.69934
68	16.0	82.45498	8.63873	58.81336	11.7582

PROPOSED PROGRAM WRITTEN IN MATLAB 2.0

```

clc;
clear all;
N= xlsread('c:\data_simulation.xls',1);
% VRS method
for i=1; size (N)
    effort1(i)=3.4 * (N(i))^1.15;
    duration1(i) = 2.2 * (effort1(i))^0.31;
%COCOMO
    effort2(i)=3.2 * (N(i))^1.05;
    duration2(i)=2.50 * (effort2(i))^0.38;
end
figure(1);
plot(effort1, effort2);
figure(2);
plot(duration1, duration2);

```

IV RESULT THROUGH SIMULATION APPROACH

Obtained results are expressed by curves that give the effort and time duration values over VRS model and COCOMO Intermediate -organic model as shown in Figure 2 and Figure3 respectively.

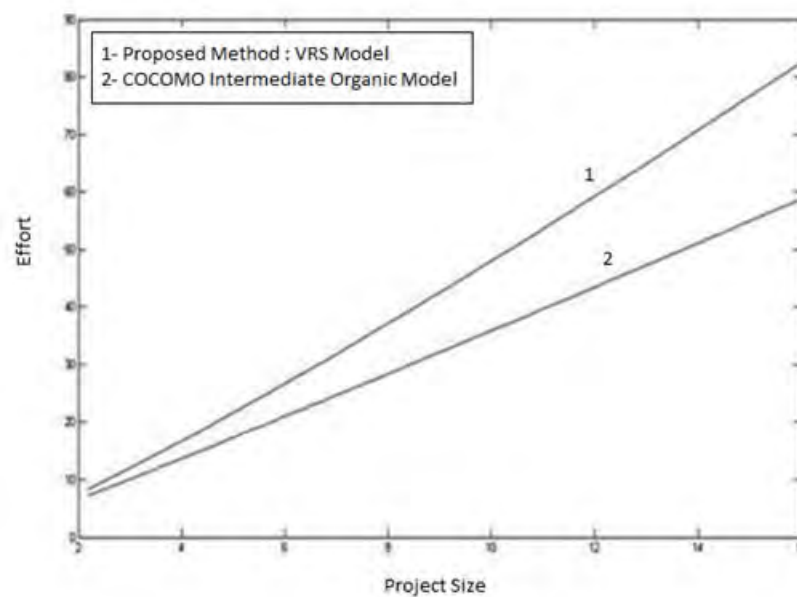


Figure 2: Graph shows effort effects.

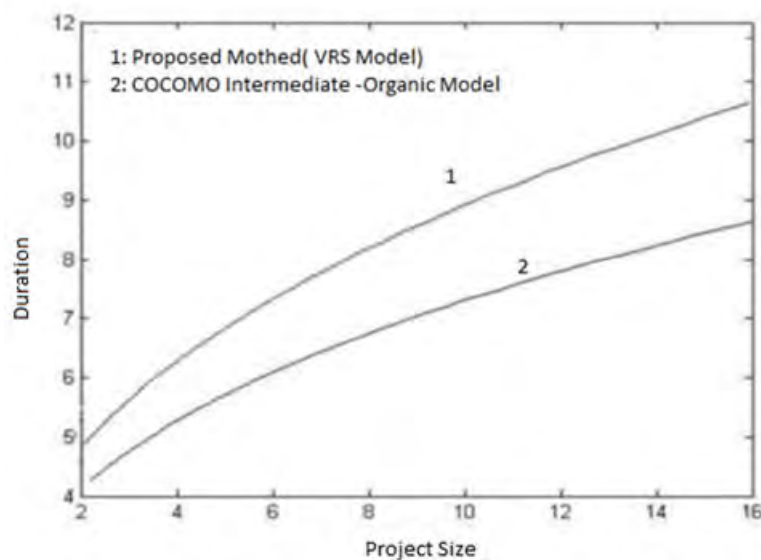


Figure 3: Graph shows duration effects.

V CONCLUSION

Simulation is one of the productivity control methods that enable software developers to take corrective actions and perform risk analysis in terms of effort cost and time to perform continuous process improvement and optimization. This paper has introduced a software process simulation modeling approach for the prediction of IVR software production cost and delivery times. In both the figure the curve of both models are closed to each other with respect of effort and duration. Therefore VRS model is accurate and may be used to estimate the effort and time duration for only IVR software.

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