

Availability Simulation Modeling and Performance Optimization of the Screening Unit in a Paper Plant

Rajiv Khanduja*, Tewari P C** and Chauhan R S***

This paper deals with the availability simulation modeling and performance optimization of the screening unit in a paper plant. The screening unit of a paper industry has four main subsystems, arranged in series and parallel configurations. Considering exponential distribution for the probable failures and repairs, the mathematical formulation of the problem is done using probabilistic approach and differential equations are developed on the basis of Markov birth-death process. These equations are then solved using normalizing conditions so as to determine the steady state availability of the screening unit. The performance of each subsystem of the screening unit in a paper plant has also been optimized using genetic algorithm. So, the findings of the present paper will be highly useful to the plant management for the timely execution of proper maintenance decisions and hence to enhance the system performance.

Keywords: Availability Simulation Modeling, Screening Unit, Performance Optimization, Genetic Algorithm

1.0 INTRODUCTION

The paper plant comprises of large complex engineering systems arranged in series, parallel or a combination of both the configurations. Some of these systems are chipping, cooking, washing, bleaching, screening, stock preparation and paper production etc. The important process of a paper plant, upon which the quality of paper depends, is the screening process. In the process of paper formation, the chips from storage are fed in to a digester to form the pulp which is processed through various subsystems called knotter, decker, opener and washers. These systems have been discussed in detail in [5-7]. The washed pulp is kept in a chamber where chlorine, at a controlled rate, is pressed through the pulp for a few hours. The pulp is passed

over filter and washer in four stages to get chlorine free white pulp. The white bleached pulp so obtained, is first passed through a screen to separate out oversize and odd shape particles. It is then processed through a cleaner and finally sent to paper making machine [8,9].

2.0 THE SCREENING UNIT

It consists of four subsystems in hybrid configuration with the following description:

- **Subsystem E_1** : It consists of a medium consistency (m.c.) pump unit used to flow the pulp from washer with consistency 4-5% with fresh water. Its failure can cause a sudden and complete failure of the system.

*Assistant Professor, Department of Mechanical Engineering, SKIET, Kurukshetra-136118 (Haryana), INDIA.
email: rajiv_khanduja@rediffmail.com

**Assistant Professor, Department of Mechanical Engineering, National Institute of Technology, Kurukshetra-136119 (Haryana), INDIA.
email: pctewari1@rediffmail.com

***Assistant Professor, Department of Electronic and Communication Engineering, JMIT, Radaur, Yamuna Nagar-135133 (Haryana), INDIA.
email: Chauhan443@yahoo.co.in

- **Subsystem E₂** : It consists of a screen unit used for removing the knots and other undesirable foreign materials from the pulp. Its failure can cause a sudden and complete failure of the system.
- **Subsystem E₃** : It consists of three cleaner units connected in parallel to mix the water with the pulp by centrifugal action. Failure of anyone cleaner results in poor quality of paper. Complete failure of this system reduces the efficiency of the plant but the system remains operative.
- **Subsystem E₄** : It consists of a decker unit used to reduce the blackness of the pulp if any by controlling the water contents. Its failure causes complete failure of the system.

e_1, e_2, e_3, e_4 : Represent failed states of respective m.c pump, screen, cleaner and decker.

$\lambda_{19}, \lambda_{20}, \lambda_{21}, \lambda_{22}$: Respective mean constant failure rates of E₁, E₂, E₃, E₄.

$\mu_{19}, \mu_{20}, \mu_{21}, \mu_{22}$: Respective mean constant repair rates of e_1, e_2, e_3, e_4 .

$P_0(t)$: State probability that the system is working at full capacity at time t.

$P_i(t)$: State probability that the system is in the ith state at time t.

$P_i^1(t)$: First-order derivative of the probabilities.

3.0 ASSUMPTIONS

The assumptions used in the probabilistic model are:

- Failure / repair rates are constant over time and statistically independent.
- A repaired unit is as good as new, performance wise for a specified duration.
- Sufficient repair facilities are provided, i.e. no waiting time to start the repairs.
- Standby units (if any) are of same nature and capacity as active units.
- System failure/repair follow exponential distribution.
- System may work at a reduced capacity.

4.0 NOTATIONS

The following notations are associated with the screening unit:

E₁, E₂, E₃, E₄: Represent good working states of respective m.c.pump, screen, cleaner and decker.

5.0 AVAILABILITY SIMULATION MODELING

The availability simulation modeling is carried out using simple probabilistic considerations and differential equations associated with the transition diagram (Fig. 1) are developed on the basis of Markov birth-death process [2, 4, 10, 11]. These equations are further solved for determining the steady state availability of the screening unit. Various probability considerations give the following differential equations associated with the screening unit:

State 0 shows the full capacity working with no standby.

States 1 to 2 show the system in reduced capacity working.

States 3 to 11 represent the system in failed state.

$$P_0^1(t) + \sum_{i=19}^{22} \lambda_i P_0(t) = \sum_{i=19}^{20} \mu_i P_{i-16}(t) + \mu_{21} P_1(t) + \mu_{22} P_5(t) + \mu_{21} P_2(t) \dots \dots \dots (1)$$

$$P_1^1(t) + \sum_{i=19}^{22} \lambda_i P_1(t) + \mu_{21} P_1(t) = \sum_{i=19}^{20} \mu_i P_{i-13}(t) + \mu_{22} P_8(t) + \lambda_{21} P_0(t) \dots \dots \dots (2)$$

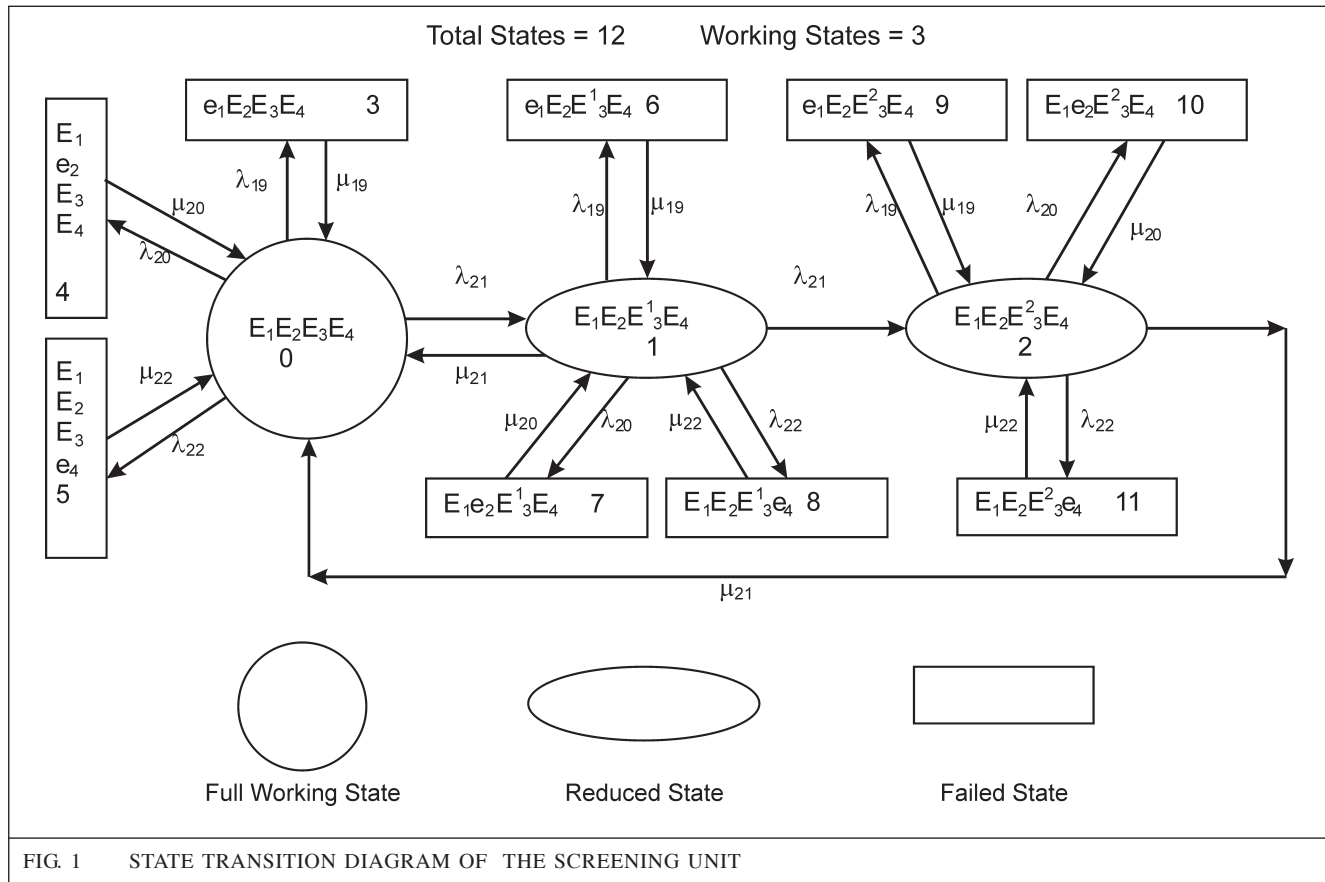


FIG. 1 STATE TRANSITION DIAGRAM OF THE SCREENING UNIT

$$P_2^1(t) + \sum_{i=19}^{20} \lambda_i P_2(t) + \lambda_{22} P_2(t) + \mu_{21} P_2(t) = \sum_{i=19}^{20} \mu_i P_{i-10}(t) + \mu_{22} P_{11}(t) + \lambda_{21} P_1(t) \dots \dots \dots (3)$$

$$P_i^1(t) + \mu_{19} P_i(t) = \lambda_{19} P_j(t)$$

where $i=3, 6, 9$ $j=0, 1, 2 \dots \dots \dots (4)$

$$P_i^1(t) + \mu_{20} P_i(t) = \lambda_{20} P_j(t).$$

where $i=4, 7, 10$ $j=0, 1, 2 \dots \dots \dots (5)$

$$P_i^1(t) + \mu_{22} P_i(t) = \lambda_{22} P_j(t)$$

where $i=5, 8, 11$ $j=0, 1, 2 \dots \dots \dots (6)$

With initial conditions at time $t = 0$

$$P_i(t) = 1 \text{ for } i = 0$$

$$= 0 \text{ for } i \neq 0$$

Since the paper plant is a process industry, its every unit should be available for long period. Therefore, steady state behaviour of the system is analyzed by substituting $d/dt \rightarrow 0$ and as $t \rightarrow \infty$ for equations (1)-(6) and solving them recursively.

$$P_1 = B P_0 \quad P_2 = B_3 B P_0 \quad P_3 = B_1 P_0$$

$$P_4 = B_2 P_0 \quad P_5 = B_4 P_0 \quad P_6 = B_1 B P_0$$

$$P_7 = B_2 B P_0 \quad P_8 = B_4 B P_0 \quad P_9 = B_1 B_3 B P_0$$

$$P_{10} = B_2 B_3 B P_0 \quad P_{11} = B_4 B_3 B P_0$$

Where $B_i = \lambda_i / \mu_i$ $i=19, 20, 21, 22$ and
 $B = \lambda_{21} / (\lambda_{21} + \mu_{21})$

Using normalizing condition i.e. sum of all the state probabilities is equal to one $[\sum_{i=0}^{11} P_i = 1]$, we get:

$$P_0 + B P_0 + B_3 B P_0 + B_1 P_0 + B_2 P_0 + B_4 P_0 + B_1 B P_0 + B_2 B P_0 + B_4 B P_0 + B_1 B_3 B P_0 + B_2 B_3 B P_0 + B_4 B_3 B P_0 = 1$$

$$P_0 [1 + B + B_3 B + B_1 + B_2 + B_4 + B_1 B + B_2 B + B_4 B + B_1 B_3 B + B_2 B_3 B + B_4 B_3 B] = 1$$

$$P_0 = 1 / [1 + B + B_3 B + B_1 + B_2 + B_4 + B_1 B + B_2 B + B_4 B + B_1 B_3 B + B_2 B_3 B + B_4 B_3 B]$$

$$P_0 = 1 / [(1 + B_1 + B_2 + B_4)(1 + B + B_3 B)]$$

The steady state availability (A_v) of this screening unit is given by summation of all the

full working and reduced capacity states.

$$Av. = P_0 + P_1 + P_2$$

$$Av. = [1 + B + B_3 B] / [(1 + B_1 + B_2 + B_4)(1 + B + B_3 B)]$$

$$\text{Availability (Av.)} = 1 / [1 + B_1 + B_2 + B_4]$$

Here, the unit performance has been evaluated in terms of availability [12-14].

6.0 GENETIC ALGORITHM

Genetic algorithms are computerized search and optimization algorithms based on the mechanics of natural genetics and natural selection [1,3]. Genetic algorithms have become important because they are found to be potential search and optimization techniques for complex engineering optimization problems.

The action of genetic algorithm for parameter optimization in the present problem can be stated as follows:

1. Initialize the parameters of the genetic algorithm.
2. Randomly generate the initial population and prepare the coded strings.
3. Compute the fitness of each individual in the old population.
4. Form the mating pool from the old population.
5. Select two parents from the mating pool randomly.
6. Perform the crossover of the parents to produce two offsprings.
7. Mutate if required.
8. Place the child strings to new population.
9. Compute the fitness of each individual in new population.

10. Create best-fit population from the previous and new population.

11. Repeat the steps 4 to 10 until the best individuals in new population represent the optimum value of the performance function (Unit Availability).

7.0 PERFORMANCE OPTIMIZATION USING GENETIC ALGORITHM

The performance behaviour of the screening unit is highly influenced by the failure and repair parameters of each subsystem. These parameters ensure high performance of the screening unit. Genetic algorithm is hereby proposed to coordinate the failure and repair parameters of each subsystem for stable unit performance i.e. high availability. Here, number of parameters is six (three failure parameters and three repair parameters). The design procedure is described as follows:

To use genetic algorithm for solving the given problem, the chromosomes are to be coded in real structures. Unlike, unsigned fixed point integer coding parameters are mapped to a specified interval $[X_{\min}, X_{\max}]$, where X_{\min} and X_{\max} are the minimum and maximum values of unit parameters. The maximum value of the availability function corresponds to optimum values of unit parameters. These parameters are optimized according to the performance index i.e. desired availability level. To test the proposed method, failure and repair rates are determined simultaneously for optimal value of unit availability. Effect of population size and crossover probability on the availability of the screening unit is shown in Tables 1-2 and Figs. 2-3. To specify the computed simulation more precisely, trial sets are also chosen for genetic algorithm and system parameters. The performance [availability] of the screening unit is evaluated by using the designed values of the unit parameters [15-17].

Failure and repair rate parameter constraints are: $(\lambda_{19}, \mu_{19}, \lambda_{20}, \mu_{20}, \lambda_{22}, \mu_{22})$

$$\lambda_{19} \in [0.05, 0.10] \quad \lambda_{20} \in [0.01, 0.09]$$

$$\lambda_{22} \in [0.02, 0.08]$$

$$\mu_{19} \in [0.10, 0.50] \quad \mu_{20} \in [0.05, 0.45]$$

$$\mu_{22} \in [0.10, 0.30]$$

The simulation is done to maximum number of population size which is varying from 20 to 100.

Number of generations - 40

Crossover probability - 0.85

Mutation probability - 0.015

Here, real-coded structures are used.

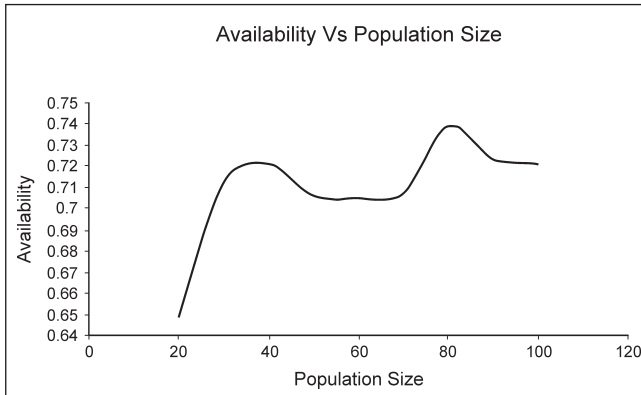


FIG. 2 EFFECT OF POPULATION SIZE ON THE SCREENING UNIT AVAILABILITY USING GENETIC ALGORITHM

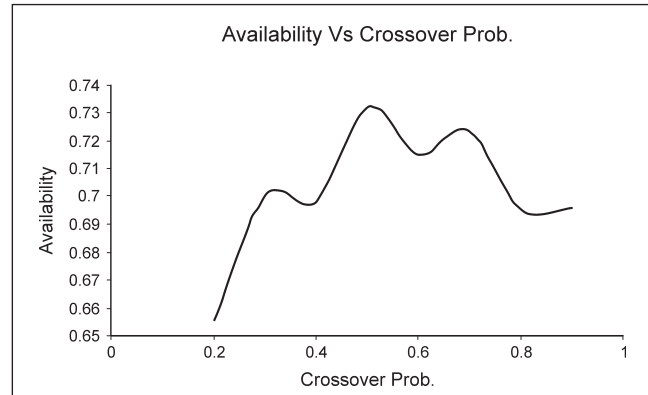


FIG. 3 EFFECT OF CROSSOVER PROBABILITY ON THE SCREENING UNIT AVAILABILITY USING GENETIC ALGORITHM

TABLE 1							
EFFECT OF POPULATION SIZE ON AVAILABILITY OF THE SCREENING UNIT USING GENETIC ALGORITHM							
Pop.Size	Av.	λ_{19}	μ_{19}	λ_{20}	μ_{20}	λ_{22}	μ_{22}
20	0.6488	0.0505	0.50	0.0111	0.3962	0.02	0.2592
30	0.7127	0.0527	0.4947	0.01	0.3808	0.02	0.30
40	0.7211	0.05	0.50	0.01	0.4450	0.02	0.2992
50	0.7068	0.05	0.4832	0.01	0.45	0.02	0.30
60	0.7055	0.0535	0.50	0.01	0.45	0.02	0.30
70	0.7076	0.05	0.50	0.0107	0.45	0.02	0.30
80	0.7390	0.05	0.50	0.01	0.4403	0.02	0.2938
90	0.7230	0.05	0.50	0.01	0.4401	0.02	0.30
100	0.7210	0.05	0.50	0.01	0.4416	0.02	0.2718

TABLE 2							
EFFECT OF CROSSOVER PROBABILITY ON AVAILABILITY OF THE SCREENING UNIT USING GENETIC ALGORITHM							
Crossover Probability	Av.	λ_{19}	μ_{19}	λ_{20}	μ_{20}	λ_{22}	μ_{22}
0.20	0.6550	0.0525	0.4137	0.0117	0.4276	0.0224	0.2970
0.30	0.7007	0.0583	0.4871	0.0110	0.4403	0.0306	0.2563
0.40	0.6982	0.0504	0.3848	0.0104	0.3422	0.0249	0.2693
0.50	0.7321	0.05	0.5	0.01	0.4473	0.02	0.3
0.60	0.7154	0.0513	0.5	0.01	0.45	0.0201	0.3
0.70	0.7238	0.05	0.5	0.01	0.45	0.02	0.2764
0.80	0.6953	0.05	0.5	0.01	0.45	0.0258	0.2781
0.90	0.6960	0.05	0.4997	0.01	0.4485	0.0201	0.2991

The effect of population size on availability of the screening unit is shown in Fig. 2. It is well indicated that at population size 80, the performance level of the Screening Unit is 73.90%. It is the optimum availability of the unit. The corresponding values of failure and repair parameters are $\lambda_{19} = 0.05$, $\mu_{19} = 0.50$, $\lambda_{20} = 0.01$, $\mu_{20} = 0.4403$, $\lambda_{22} = 0.02$, $\mu_{22} = 0.2938$ as given in Table 1. This is the best possible combination of failure and repair rates with respect to optimum availability of the unit.

Now the simulation is done for maximum number of crossover probability which is varying from 0.20 to 0.90.

Population size - 40
 Number of generations - 40
 Mutation probability - 0.015

The effect of crossover probability on availability of the screening unit is shown in

Fig. 3. The optimum value of unit's performance is 73.21%, for which the best possible combination of failure and repair rates are $\lambda_{19} = 0.05$, $\mu_{19} = 0.50$, $\lambda_{20} = 0.01$, $\mu_{20} = 0.4473$, $\lambda_{22} = 0.02$, $\mu_{22} = 0.3$ at crossover probability 0.50 as clearly shown in Table 2.

8.0 CONCLUSION

The performance optimization of the screening unit of a paper plant has been carried out in this paper. Genetic Algorithm Technique is hereby proposed to select the various feasible values of the unit failure and repair parameters along with unit availability levels. Finally, Genetic Algorithm Technique is successfully applied to coordinate simultaneously these parameters for determining an optimum level of unit availability. Besides, the effect of Genetic Algorithm parameters such as population size and crossover probability on the unit performance i.e. availability has also been analyzed. By varying the above mentioned

parameters of Genetic Algorithm the optimum unit availability achieved is about 73.90% with best possible combinations of the failure and repair rates of all the subsystems of the screening unit. Then, the findings of this paper are discussed with the concerned paper plant management. Such results are found highly beneficial for the purpose of performance optimization of a screening unit in the paper plant concerned.

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