# BAYESIAN BELIEF NETWORKS FOR SOFTWARE DEFECT PREDICTION

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## DEPARTMENT OF COMPUTER SCIENCE & ENGINEERING **CERTIFICATE**

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## **DECLARATION**

We hereby declare that the results embodied in this dissertation entitled "Bayesian Belief Networks for Software Defect Prediction" has been carried out by us together during the academic year 2013-2014 as a partial fulfillment of the award of the B-tech degree in information technology from JNTUH. We have not submitted the same to any other university or organisation for award of any other degree.

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#### **ABSTRACT**

There are lots of different software metrics discovered and used for defect prediction in the literature. Instead of dealing with so many metrics, it would be practical and easy if we could determine the set of metrics that are most important and focus on them more to predict defectiveness. We use Bayesian networks to de-termine the probabilistic influential relationships among software metrics and defect proneness. In addition to the metrics used in Promise data repository, we define two more metrics, i.e. NOD for the number of developers and LOCQ for the source code quality. We extract these metrics by inspecting the source code repositories of the selected Promise data repository data sets. At the end of our modeling, we learn the marginal defect proneness probability of the whole software system, the set of most effective metrics, and the influential relationships among metrics and defectiveness. Our experiments on nine open source Promise data repository data sets show that response for class (RFC), lines of code (LOC), and lack of coding quality (LOCQ) are the most effective metrics whereas coupling between objects (CBO), weighted method per class (WMC), and lack of cohesion of methods (LCOM) are less effective metrics on defect proneness. Furthermore, number of children (NOC) and depth of inheritance tree (DIT) have very limited effect and are untrustworthy. On the other hand, based on the experiments on Poi, Tomcat, and Xalan data sets, we observe that there is a positive correlation between the number of developers (NOD) and the level of defectiveness. However, further investigation involving a greater number of projects is needed to confirm our findings.

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