

# Predicting Upgrade Project Defects based on Enhancement Requirements: An Empirical Study

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# Outlines

- Introduction
- Methodology
- Experiment
- Conclusion

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# Introduction

## ● Background

- Upgrade Projects: delivering updated release or version of the software product
- Enhancement Requirements (ER): requirement additions and modifications – **Inject defects**

## ● Motivation

- ER impact analysis: risk management and cost reduction in early software lifecycle
- Decision-making support: predict properties of defects may introduced (quantity, workload, type, priority...)

# Introduction-cont. 1

- **Prior work**

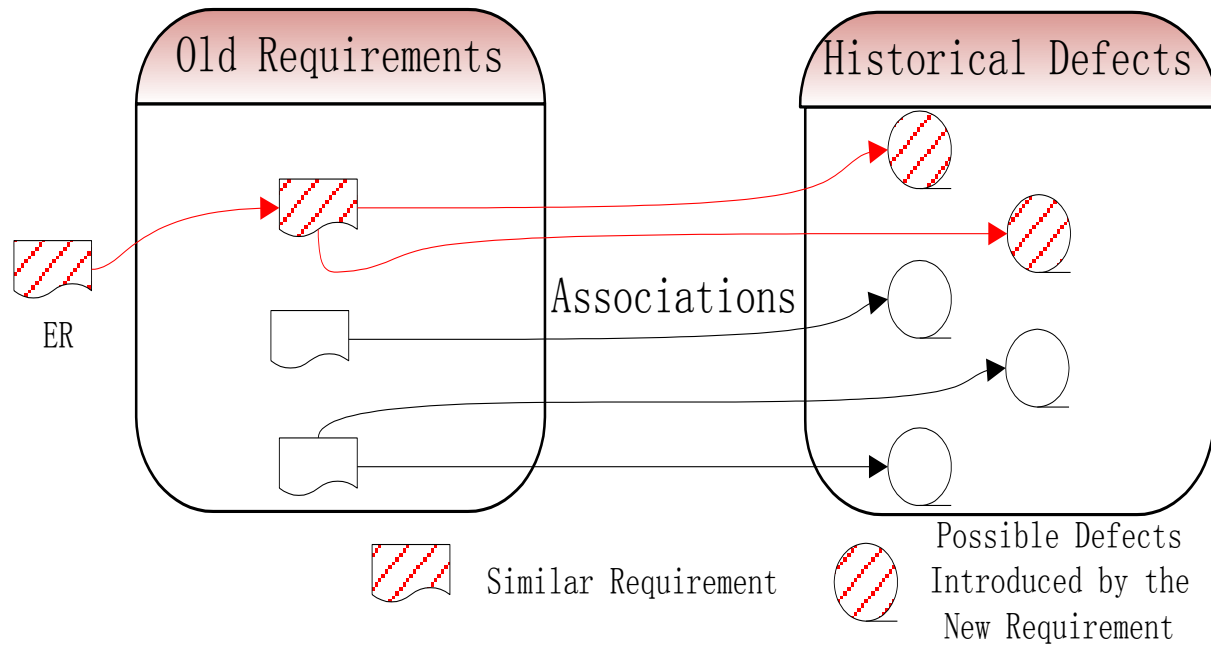
- Static defect prediction approaches applying Support Vector Machines(SVM)
- Requirement measurement technologies (consistent, complete, correct... )

- **Problem & Solving**

- Gap between requirement and defect:  
requirement → design → coding → testing(**defects**)
- Solving: An empirical methodology applying Information Retrieval and Machine Learning technology (skip details)

# Introduction-cont. 2

- Basic Idea

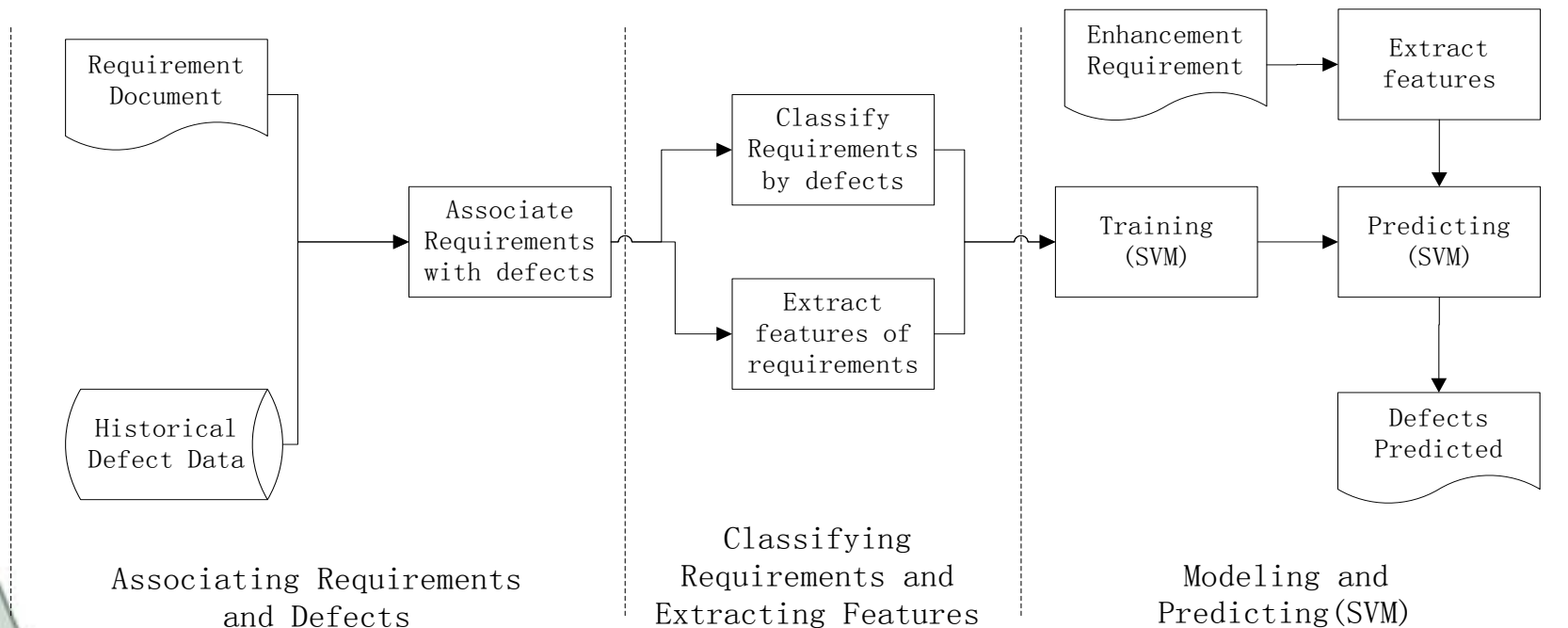


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# Methodology-overview

- Methodology overview: Coverting prediction problem to a classification problem



# Methodology

- Associating historical requirements and defects
- Classifying historical Rqmt and extracting Rqmt features
- Modeling and predicting using SVM

# Methodology

- *Associating historical requirements and defects*
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# Step 1: Associating

- Historical requirements(docs): use cases

- Name: *Submit Work Plan;*
- Description: *To save and submit work plan in the work space;*
- Pre-conditions: *User logged in;*
- Basic Event Flow: *1. Enter the work space, open the work plan table. 2. Input work plan. 3. Click 'submit' Button; Additional Work Flow: None;*
- Post-conditions: *A new plan added in project window. ....*

- Historical defects(database records)

Name	Description
Title	Title of the defect.
Description	Brief description of the defect.
Workload	Workload to fix the defect (man-hour).
Priority	Priority of the defect: , Serious and Critical.
Type	Type of the defect: UI, Internal or Integrated.

# Step 1: Associating – cont. 1

- **Associating (IR technique)**

- Defect title +description as query
- Requirement items (use case) as target indexing documents
- Automated get matching scores using formula:

$$\sum_{t \text{ in } d} TF(t \text{ in } r) * IDF(t) * Boost(t.field \text{ in } r) * LengthNorm(t.field \text{ in } r)$$

- Couples which [matching score > threshold(0.6 as default)] are considered as associated defects and requirements
- Manual filtering

# Methodology

- Associating historical requirements and defects
- *Classifying historical Rqmt and extracting Rqmt features*
- Modeling and predicting using SVM

## Step 2: Classifying historical Rqmt and extracting Rqmt features

- **Classifying historical requirements**
  - According to the statistic properties of defects related, historical requirements can be divided to several classes.
  - Different classification standards are for predicting different defect properties. E.g.: classify requirements according to defect quantity related is to predict quantity of defects may introduced.
  - Discover more defects properties from DB records to predict more detailed information from different aspects.

## Step 2: Classifying historical Rqmt and extracting Rqmt features –cont. 1

- **Extracting features from requirements**
  - Common Features: CHI metrics. Basic context metrics of requirement, counting classification weight of each term in the documents
  - Special Features: metrics perform differently in models predicting different properties
    - Module of the requirement (MoR)
    - Text length of the requirement item (LoR)
    - Number of items in event flow (NEF)
    - ...

# Step 2: Classifying historical Rqmt and extracting Rqmt features -cont.2

- Generating SVM input: each requirement item are converted to one vector

R:      3      1 3 2 5 0 0 2 4... 4 2 1 0 0 1 5...

class lable

Common Features

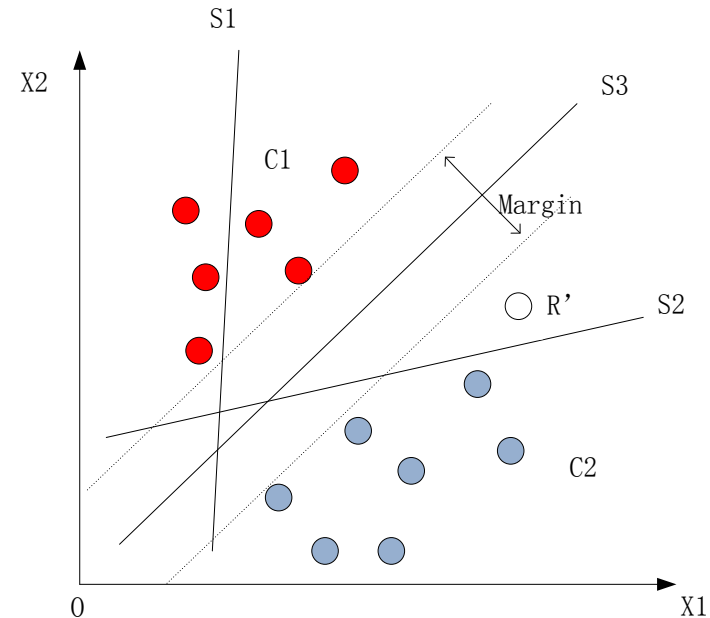
Special Features

# Methodology

- Associating historical requirements and defects
- Classifying historical Rqmt and extracting Rqmt features
- *Modeling and predicting using SVM*

# Step3: Modeling and Predicting

- SVM Models
- Advantages
  - High dimensional
  - Global optimum solution
  - Non linear
  - Prevent over-fitting



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# Experiment

- **Data Set:**

- Defects and requirement data of 9 versions of SoftPM
  - SoftPM project is a software quality management platform, targeting to facilitate the process improvement initiatives in many small and medium software organizations(Since 2002).
- 4893 defects and 581 requirement items (use case).
- Establish four models of common defect properties
  - Quantity, workload, type, priority

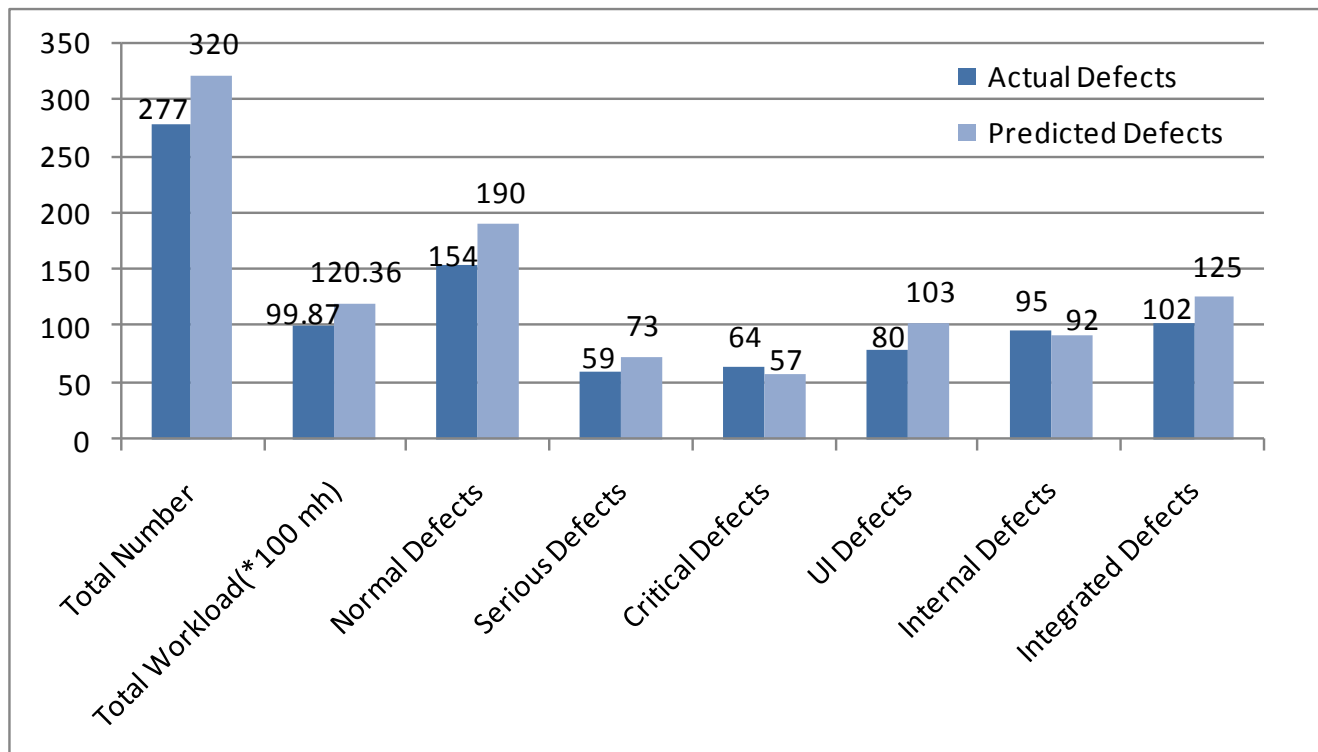
# Experiment-cont. 1

- Step 1 and 2 : historical data association and classification

Classification Standard	Classes	Boundary (Range)	Requirement Features
By <b>Quantity</b> of Related Defects	Extra Few	0~5	CHI (1000)
	Few	6~10	MoR (50)
	Medium	11~29	LoR (100)
	High	30~49	NEF (100)
	Extra High	More than 50	PCR (50)
By <b>Average Workload</b> of Related Defects	Extra Low	0~24 (man-hour)	CHI (1000)
	Low	25~48	MoR (100 )
	Medium	49~144	LoR (50)
	High	145~288	NEF (150)
	Extra High	More than 289	
By <b>Priority</b> of related Defects	Normal	Regarding to defect properties recorded in the database	CHI (1000)
	Serious		MoR (200)
	Critical		NEF (100)
By <b>Type</b> of related Defects	UI	Regarding to defect properties recorded in the database	CHI (1000)
	Internal		MoR (100)
	Integrated		NEF (100)
			DUI (50)
			RDP (50)

# Experiment- cont.2

- Step 3: modeling and predicting
  - 10-fold cross validation accuracy is around 65%
  - Quantified predicted results: a test set of 48 ERs



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# Conclusion

- **Advantage**

- Novel and promising impact analysis for requirement evolution and software maintenance
- Defect prediction in early software lifecycle

- **Disadvantage**

- Empirical model, Differ from projects.
- Requirement deletion?

- **Future Work**

- Improve Accuracy: optimize metric selection
- More data sets

# Thank You!