

# Mining Software Repositories for Intelligent Software Maintenance

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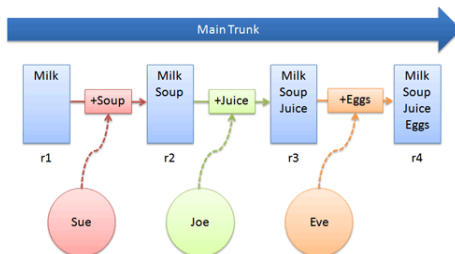
# Executive Summary

- Change management with version control systems
- Improving software maintenance through software repository mining
- Framework for preventive maintenance
- Novelty: Metrics for localization
- Study of Open Source projects

# Outline

- 1 Version Control**
- 2 Mining Software Repositories**
  - Frequent Item Set Mining
  - Maintenance Challenges
- 3 Framework**
- 4 Novelty**
- 5 Experiments**
  - Linux 2.6
  - Wine
  - Insights

# Version Control



- Management of changes to computer files in a repository
- Changes identified by a number or letter code ("revision")
- Each revision associated with timestamp and person making the change
- Version control systems: CVS, Subversion, Git, ...

# Working Copy, Commits and Change Sets

- Working copy: Local copy of files from a repository
- Commit: Writing changes to the working copy into the repository
- Change set: Set of changes made in a single commit

```
commit 3d2d827f5ca5e32816194119d5c980c7e04474a6
Author: Michael S. Tsirkin <mst@redhat.com>
Date:   Mon Sep 21 17:03:51 2009 -0700
```

```
mm: move use_mm/unuse_mm from aio.c to mm/
```

```
M      fs/aio.c
A      include/linux/mmu_context.h
M      mm/Makefile
A      mm/mmu_context.c
```

# Outline

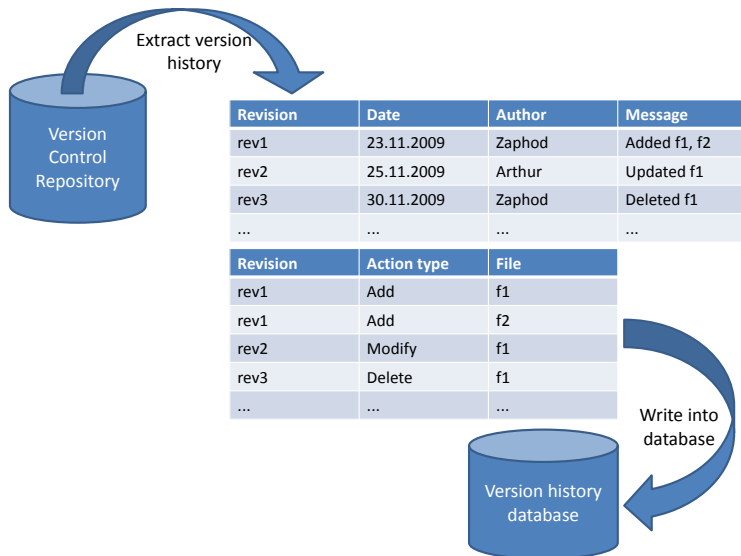
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# Software Repository Mining

- Version control systems contain large amounts of historical information: “*Who changed what, why and when.*”
- Learn from the past to shape the future
- Automated extraction, collection, and abstraction of information from software development data



# Populating Version History Database





# Frequent Item Set Mining

- Popular method for market basket analysis
- Identify sets of products frequently bought together: Beer and diapers
- Framework applies frequent item set mining to the version history of software repositories
- Identify which code files have been frequently changed together



Source: <http://research.nii.ac.jp/~uno>

# Frequent Item Set Mining: Transactions

- Transactions: Change sets
- Example:

```
{ fs/aio.c, include/linux/mmu_context.h,  
  mm/Makefile, mm/mmu_context.c }
```

```
commit 3d2d827f5ca5e32816194119d5c980c7e04474a6
```

```
M      fs/aio.c  
A      include/linux/mmu_context.h  
M      mm/Makefile  
A      mm/mmu_context.c
```

# Frequent Item Set Mining: Definitions

- **Transaction database** contains all change sets
- Members of transactions are **items**
- **Item set** is a subset of possible items
- **Support** of an item set  $i$ :  $\text{sup}(i) :=$  number of transactions  $t$  that contain  $i$

# Frequent Item Set Mining: Association Rules

## Customers Who Bought This Item Also Bought

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The screenshot shows four book recommendations from Amazon. Each recommendation includes a book cover, the title, author, star rating, and price. The books are:

- So Long, and Thanks for All the Fish** by Douglas Adams: 4.5 stars (80 reviews), \$7.99.
- The Restaurant at the End of the Universe** by Douglas Adams: 4.5 stars (122 reviews), \$11.20.
- Mostly Harmless** by Douglas Adams: 4.5 stars (166 reviews), \$7.99.
- The Hitchhiker's Guide to the Galaxy, 25th Anniversary Edition** by Douglas Adams: 4.5 stars (873 reviews), \$10.20.

Source: amazon.com

- If a customer buys **bread** and **wine**, **then** she will probably also buy **cheese**
- Problem decomposed into two subproblems:
  - Finding frequent item sets with minimum support
  - Generate association rules with minimal confidence
- Confidence for association rule  $R : X \rightarrow Y$ :  

$$\text{conf}(R) = \text{conf}(X \rightarrow Y) = \frac{\text{sup}(X \cup Y)}{\text{sup}(X)}$$

# Frequent Item Set Mining: Example

Transaction IDs	Transactions (Files)
1	{1, 2, 3, 4}
2	{2, 3, 4}
3	{2, 3}
4	{1, 2, 4}
5	{1, 2, 3, 4}
6	{2, 4}

# Maintenance Challenges

- Predicting changes
  - Incomplete changes
- Traceability links
  - “Cross-language” changes
- Predicting faults
- Understanding software evolution
  - Measure change localization



# Predicting Changes



A. Ying, G. Murphy et al., *Predicting Source Code Changes by Mining Change History*

- Determines change patterns from change history of the code base
- Uses association rule mining for identifying implicit dependencies
- Change patterns can be used to recommend potentially relevant source code to a developer performing a modification task

# Predicting Changes: Incomplete Change

- Comments in modification task report of Mozilla:
  - 2002-06-12 14:14: Patch to gtk/nsFontMetricsGTK.cpp, limiting the size of fonts to twice the display height.
  - 2002-06-12 14:37: Patch misses the Xlib version.
  - A patch was later submitted with the correct changes in the X-windows font handling code in the file xlib/nsFontMetricsXlib.cpp
- gtk/nsFontMetricsGTK.cpp does not reference xlib/nsFontMetricsXlib.cpp → used in different configurations
- Files were changed 41 times together → change pattern
- Changing the gtk/nsFontMetricsGTK.cpp could trigger a recommendation for xlib/nsFontMetricsXlib.cpp



# Traceability Links



H. Kadgi et al., *Mining Software Repositories for Traceability Link*

- If files of difference types are co-changed with a high frequency over multiple versions → potential traceability link
- Traceability links derived from the actual changes to files by mining software repositories
- Uses sequential-pattern mining to identify and analyze sets of files that are committed together
- Sequential-pattern mining produces ordered lists of co-changing files
- Ordering information can be used to infer directionality

# Traceability Links: Example

- Mining change sets in the Wine repository
- Changes are tested:
  - `./dlls/gdiplus/graphicspath.c -> ./dlls/gdiplus/tests/graphicspath.c`
  - `./dlls/inetmib1/main.c -> ./dlls/inetmib1/tests/main.c`
- Cross language changes:
  - `./dlls/rpcrt4/tests/server.c -> ./dlls/rpcrt4/tests/server.idl`
  - `./dlls/dxgi/dxgi_private.h -> ./include/wine/winedxgi.idl`

# Predicting Faults



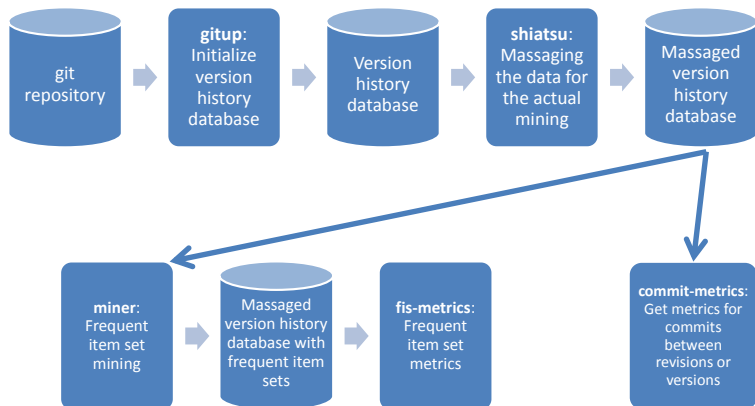
S. Kim, T. Zimmermann et al., *Predicting Faults from Cached History*

- Assumption: faults do not occur in isolation, but rather in bursts of several related faults
- Identifying bug fixes by mining commit messages: Searching for keywords such as “Fixed” or “Bug” and references to bug reports like “42”
- Cache locations that are likely to have faults
- By consulting the cache at the moment a fault is fixed, a developer can detect likely fault-prone locations

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# Architecture



# Git

- Free distributed version control system
- Initially designed and developed for Linux kernel development
- Every working directory is a full-fledged repository:
  - Complete history
  - Full revision tracking capabilities
  - Not dependent on network access or a central server
- Easily convert repositories of other version control systems like Subversion into Git repositories
- Only need to write mining and analysis tools for one format rather than many

# Populating the Version History Database

- Gitup generates logfile and initializes versions history database
- Shiatsu massages the data to be used by the metrics applications
  - Set modularization according to specified directory depth
  - Remove deleted files
  - Set number of modifications
  - Heuristics for file moves
- Massaged version history database is used to generate frequent item sets and calculate metrics



# Preventing Maintenance

Our framework can help with solving all mentioned maintenance challenges:

- Predicting changes
  - Incomplete changes
- Traceability links
  - “Cross-language” changes
- Predicting faults
- Understanding software evolution
  - Measure change localization



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# Change Localization

- A change is well localized, if it touches only one or very few modules
- A change is not well localized, if it touches many modules
- Apply change localization for frequent item sets

## Hypothesis

Changes in frequent item sets in well modularized software systems are localized

# Change Localization: Example

- Well localized: Touches only one module

`dlls/ntdll/signal_i386.c`

`dlls/ntdll/thread.c`

- Badly localized: Touches four modules out of five possible

`if1632/thunk.c`

`include/process.h`

`loader/task.c`

`scheduler/process.c`

`scheduler/thread.c`

- Not localized at all: Touches all possible modules

`files/dos_fs.c`

`scheduler/syslevel.c`

`tools/winapi-check`

# Change Localization Metrics

- Value between 0 and 1
- 0: Not localized at all
- 1: Fully localized

$$\frac{\sum_{i=FIS_1}^{FIS_n} 1 - \left( \text{if} \left( i.modules\_touched = 1, 0, \frac{i.modules\_touched}{i.files\_touched} \right) \right)}{n}$$

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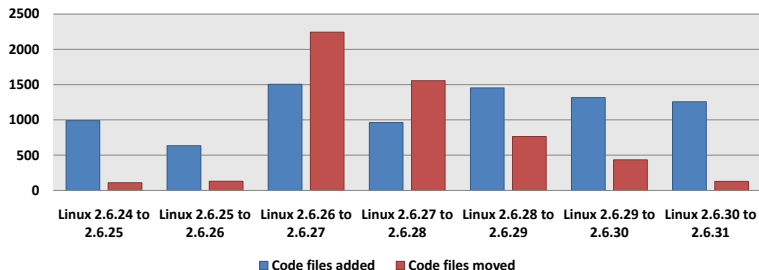
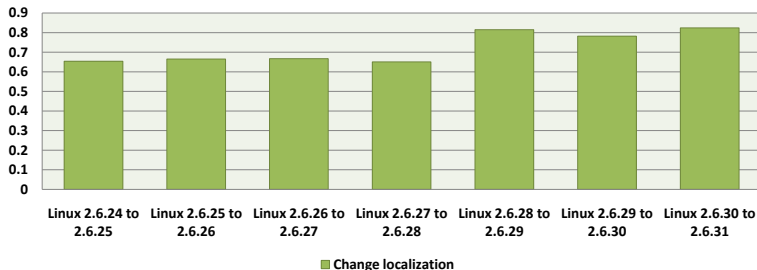
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# Linux 2.6

- Unix-like operating system kernel
- Repository checked out on November 19, 2009
- 25,277 code files
- 168,800 commits
- Frequent item set mining:
  - Minimum number of commits (modifications) for code files: 4
  - Minimum support: 4
  - Maximum size of commits (number of code files): 50



# Linux 2.6: Frequent Item Set Metrics



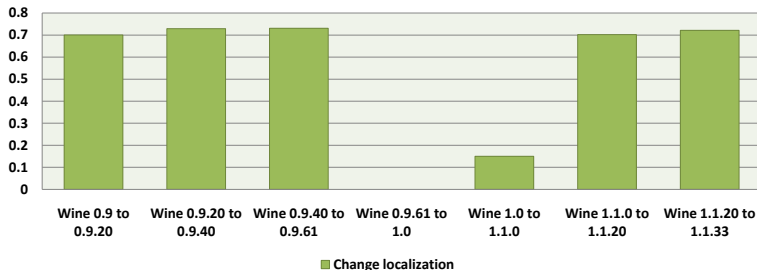
# Wine

- Allows execution of Microsoft Windows programs on Unix-like operating systems
- Repository checked out on November 20, 2009
- 3,479 code files
- 63,864 commits
- Frequent item set mining:
  - Minimum number of commits (modifications) for code files: 4
  - Minimum support: 4
  - Maximum size of commits (number of code files): 50





# Wine: Frequent Item Set Metrics



# Insights

- Code file moves increase localization
- Adding of many code files decrease localization
- Adding of code files can clear effect of moves on localization
- Stable versions contain mostly bug fixes
  - ⇒ low localization, only few moves and adds
- Unstable versions contain mostly new features
  - ⇒ high localization, many moves and adds

# Future Work

- Use framework to mine software repositories of commercial systems
- Compare localization metrics of Open Source and Closed Source systems
- Use the frequent item sets extracted to come up with a better modularization
- Publish research in the form of a paper

# Summary

- Mining software repositories for intelligent software maintenance
- Applications of frequent item set mining in improving software maintainability
- Framework for preventing software maintenance
- New metrics for change localization
- Localization of frequent item sets of Open Source projects

