

Predicting Fault Incidence Using Software Change History

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Code Decay

- ⌘ Software structure degrades over time
 - ⌘ Why?
- ⌘ Changes can become:
 - ⌘ Costly
 - ⌘ Time-consuming
 - ⌘ **Fault-producing**
 - ⌘ When one fix leads to one fault on average, what's the use? We should just go home...

Fault Analysis

- ⌘ Usually looks at:
 - ⌘ Number of faults remaining
 - ⌘ Explaining the number of faults found
- ⌘ *This paper assumes that new faults are added as the system is changed.*

Definitions

- ⌘ *Module*
 - ⌘ Collection of related files
- ⌘ *Delta*
 - ⌘ Change to a module
- ⌘ *Age*
 - ⌘ Weighted average of dates of deltas weighted with sizes of the deltas

Predictors of the Number of Faults

- ⌘ Product Measures
 - ⌘ Computed from syntactic data
- ⌘ Process Measures
 - ⌘ Computed from change and defect histories

Product Measures

- ⌘ Lines of Code
- ⌘ Other Complexity Measures (McCabe, etc.)
 - ⌘ Highly correlated with lines of code
- ⌘ Not very good predictors of faults

Process Measures

- ⌘ Number of past faults
 - ⌘ "Stable model"
- ⌘ Number of historical deltas to a module
- ⌘ Average age of the code
- ⌘ Development organization

Process Measures Continued

- ⌘ Number of developers making changes
- ⌘ Module's connection to other modules
 - ⌘ In terms of the modules being changed together
- ⌘ "Weighted time damp model"
 - ⌘ More recent changes contribute more to fault potential

The Experiment

- ⌘ 1.5 million LOC legacy from a telephone switching system
- ⌘ Looked at data from a two-year period
- ⌘ Modules have different versions (domestic, international, and common)

IMRs

- ⌘ "Initial Modification Request"
 - ⌘ Read: "Change Request"
 - ⌘ Official record of a problem to be solved
 - ⌘ Two types, set by originator
 - ⌘ "Bug" – bug fix or request for missing feature
 - ⌘ "New" – new feature
 - ⌘ Typically results in several deltas

Data Sources

- ⌘ Data sources:
 - ⌘ IMR database
 - ⌘ Only examine those classified as bug fixes
 - ⌘ Delta database
 - ⌘ Read "Change Management"
 - ⌘ Deltas associated with IMRs
 - ⌘ Source code
 - ⌘ Comments included in LOC counts

Models

- ⌘ Hypothesized formulas for fault prediction
- ⌘ Composed of one or more variables (such as deltas, age, or lines of code)
- ⌘ Different models are postulated and their fault-predicting powers are statistically examined

Statistics Technique

- ⌘ Generalized Linear Models (GLMs)
 - ⌘ Curve-fitting technique (i.e. attacks the same type of problem as linear regression / least-squares)
 - ⌘ Effective on Poisson distributions
 - ⌘ Made a logarithmic function of the mean to be linear in the variables
 - ⌘ Error measure chosen to minimize the effects of having radically different sizes and fault counts of modules
 - ⌘ Deviance function for the Poisson distribution

Simulations

- ⌘ Used to compute the significance of variables in models
- ⌘ Generated synthetic fault data and compared deviances between models

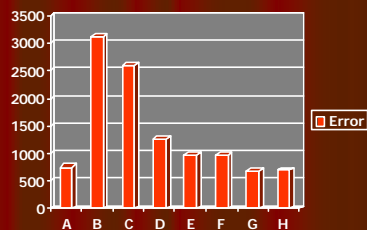
Basic Generalized Linear Models

- ⌘ Stable Model
 - ⌘ Assumes that fault generation dynamics for a module remain stable over time
 - ⌘ In other words, if you found 100 faults last year, you'll find 100 this year
 - ⌘ Insight-free
 - ⌘ Implicitly incorporates many of the other predicting variables
- ⌘ Null Model
 - ⌘ All modules have the same number of faults
- ⌘ Organization Only
 - ⌘ Prediction by module version (international, domestic, or common)

Results

	Model	Error
A	Stable	757.4
B	Null	3100.0
C	Organization Only	2587.7
D	$0.84 \log(\text{lines}/1000)$	1271.4
E	$0.14 \log(\text{lines}/100) + 1.19 \log(\text{deltas}/1000)$	980.0
F	$1.05 \log(\text{deltas}/1000)$	985.1
G	$0.07 \log(\text{lines}/1000) + 0.95 \log(\text{deltas}/1000) - 0.44 \text{ age}$	696.3
H	$1.02 \log(\text{deltas}/1000) - 0.44 \text{ age}$	697.4

Results Again



Observations

- ⌘ Predictors
 - ⌘ Deltas are a better measure of fault likelihood than lines
 - ⌘ Age idea is helpful to incorporate, too
- ⌘ Non-predictors
 - ⌘ Lines don't help much
 - ⌘ Complexity metrics were predictable from lines of code
 - ⌘ Number of developers working on the code
 - ⌘ Module's connectivity to other modules

Correlation of Complexity Metrics

	1	2	3	4	5	6	7	8	9	10	11	12
1 Lines Of Code	1	.97	.88	.86	.91	.89	.98	.92	.97	.88	.72	.35
2 McCabe V(G2)	.97	1	.88	.90	.98	.85	.85	.89	.93	.86	.76	.39
3 Functions	.88	.88	1	.82	.89	.85	.84	.91	.86	.78	.65	.39
4 Symbols	.86	.90	.82	1	.85	.89	.85	.85	.85	.78	.67	.27
5 Unique Operators	.82	.89	.89	.83	1	.89	.87	1.00	.94	.86	.67	.48
6 Total Operators	.89	.95	.88	.86	.89	1	1.00	.90	.98	.88	.72	.31
7 Program Volume	.86	.95	.84	.82	.87	1.00	1	.89	.97	.87	.74	.36
8 Expected Length	.82	.89	.89	.85	1.00	.90	.88	1	.94	.88	.83	.42
9 Variable Count	.87	.93	.84	.85	.94	.89	.97	.94	1	.77	.60	.28
10 MaxSpan	.85	.86	.78	.76	.82	.85	.87	.69	.77	1	.82	-0.39
11 MeanSpan	.72	.76	.65	.67	.67	.72	.74	.55	.60	.62	1	-0.15
12 Prag Level	.35	.29	.39	.27	.48	.31	.28	.42	.36	0.10	-0.25	1

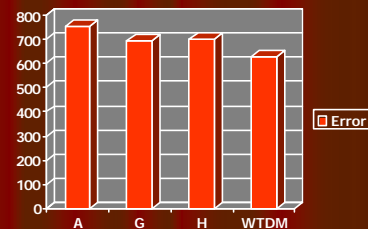
Weighted Time Damp Model

- ⌘ Considers the fault potential to be a weighted sum of all historical changes in a module
- ⌘ Contribution of a change goes down about 50% per year
- ⌘ Assumes that old changes have been fixed or proven to be fault-free
- ⌘ Treats changes individually

Results

	Model	Error
A	Stable	757.4
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D	$0.84 \log(\text{lines}/1000)$	1271.4
E	$-0.14 \log(\text{lines}/100) + 1.19 \log(\text{deltas}/1000)$	980.0
F	$1.05 \log(\text{deltas}/1000)$	985.1
G	$0.07 \log(\text{lines}/1000) + 0.95 \log(\text{deltas}/1000) - 0.44 \text{ age}$	696.3
H	$1.02 \log(\text{deltas}/1000) - 0.44 \text{ age}$	697.4
	<i>Weighted Time Damp</i>	<i>631.0</i>

Results Again



Weighted Time Damp Model (Cont.)

- ⌘ After picking some parameters, they were able to get an error of 631.0
- ⌘ ***This was their most successful model***

?

- ⌘ Exponential (damping) parameter in the time damp model
- ⌘ Rate at which the contribution of old changes disappears
- ⌘ Error is minimized with respect to this
- ⌘ ***Over different time periods, ? could differ by a factor of 2***

Any Questions?

Results

TABLE 1
Models to Fit Fault Data

Model	Intep	Common	Intl	US	Error
(A) Stable	-	-	-	-	757.4
(B) Null model	-	-	-	-	3108.8
(C) Organization only	3.46	0	-0.13	-1.39	2587.7
(D) $0.84 \log(\text{lines}/1000)$	0.92	0	0.17	-0.92	1271.4
(E) $-0.14 \log(\text{lines}/1000) + 1.19 \log(\text{deltas}/1000)$	3.31	0	0.46	-0.70	980.0
(F) $1.05 \log(\text{deltas}/1000)$	2.95	0	0.43	-0.72	985.1
(G) $0.07 \log(\text{lines}/1000) + 0.95 \log(\text{deltas}/1000) - 0.44 \text{Age}$	2.63	0	0.73	-0.65	696.3
(H) $1.02 \log(\text{deltas}/1000) - 0.44 \text{Age}$	2.87	0	0.74	-0.63	697.4

Weighted Time Damp Model

631.0