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“Scientific Computation and the Scientific Method: a tentative road map for convergence”

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Overview



- Popperian deniability
- Some early thoughts
- A tentative model for defect
- Conclusions

Popperian deniability



- Truth cannot be verified by scientific testing, it can only be falsified.
- Falsification requires quantification of experimental error.
- This has been at the heart of scientific progress.
- This process is NOT generally followed in scientific (or indeed any other kind of) computation.

The problem with defects



- We seek quantification. This means we would like to know how big the errors in our numerical experiments are.
- Unfortunately, most of what we know concerns how many defects are present and **not** how big a problem they cause.
- More than a whiff of chaos
 - `{int a; b = (a=0) + a; ...}` b can be almost anything.
 - 14 out of 14 compilers got volatile wrong in a 2008 study
 - Undetected array bound violations still with us in 2011 !
- Any engineering technology which relies on somebody getting it ‘right’ is fundamentally flawed.

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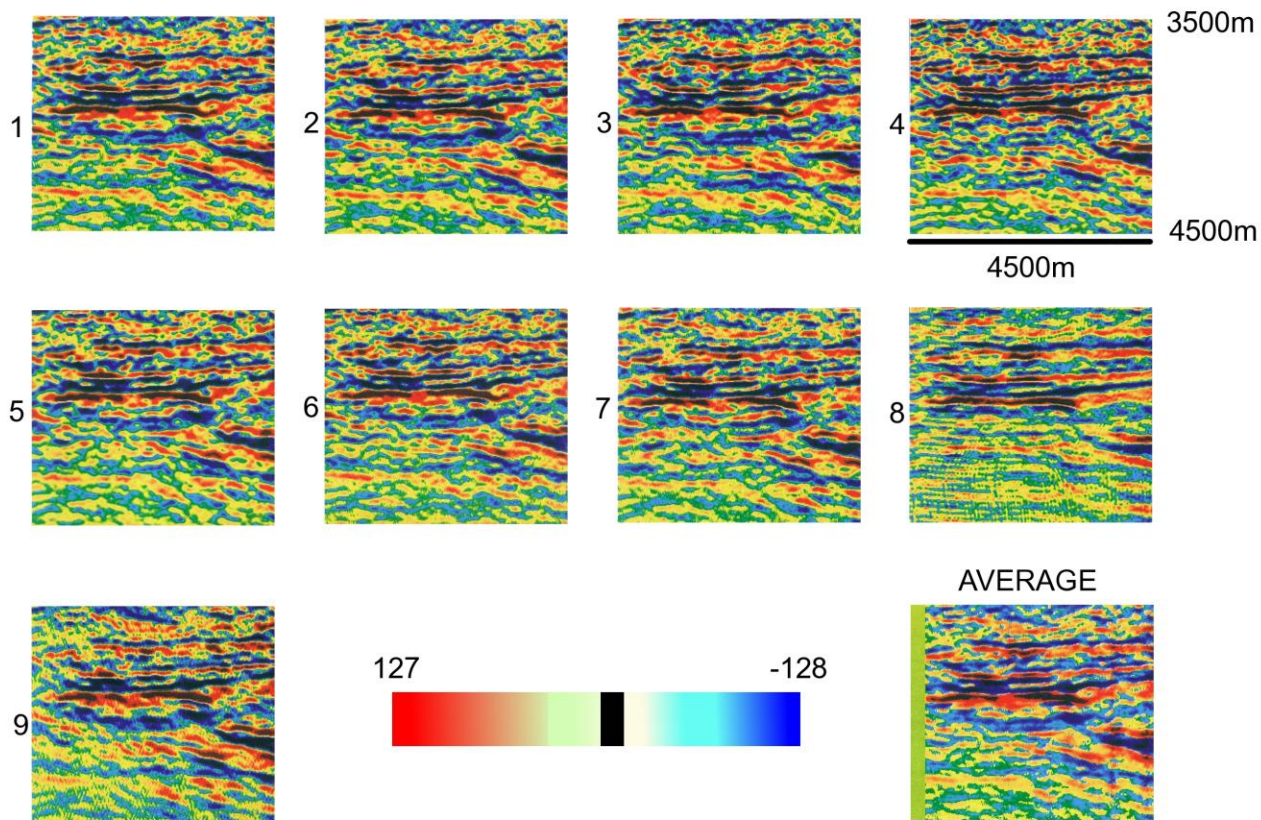
Some early thoughts



By 2010 I was reasonably convinced that:

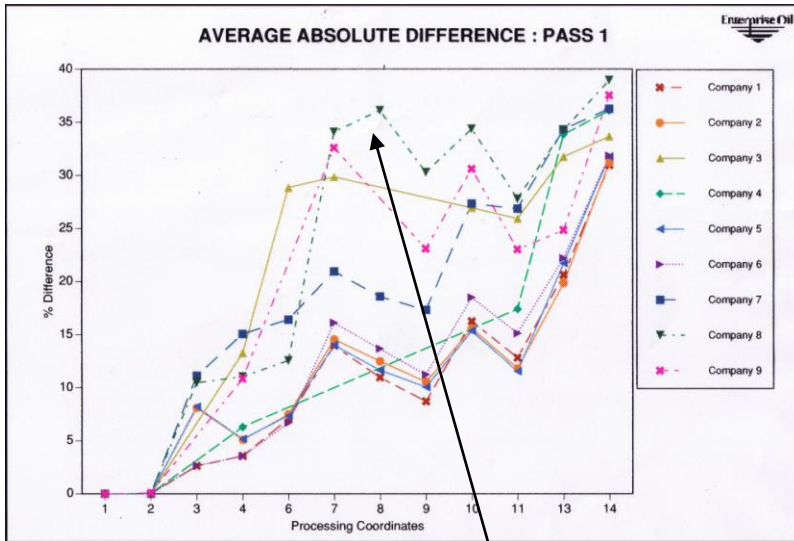
- N-version experiments are exceedingly valuable at highlighting differences, (for whatever reason), and effective at reducing those differences. (1994)
- Scientific software is littered with statically detectable faults which fail with a certain frequency (1997)
- The language does not seem to make much difference. (1999-)
- Defects appear to be fundamentally statistical rather than predictive, (2005-8)
- Software systems exhibit implementation INdependent behaviour (2007-10).

Quantification of differences by N-version (1994)

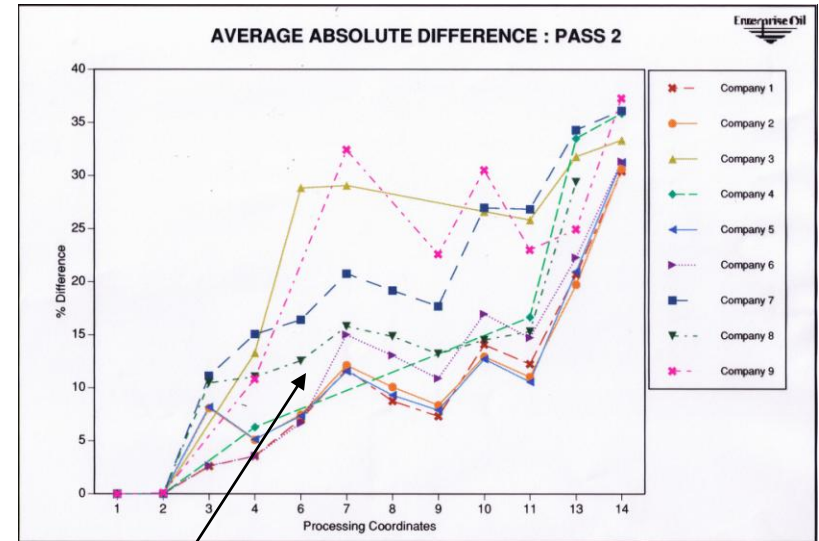


Convergence using N-version

– but to what ?



Before

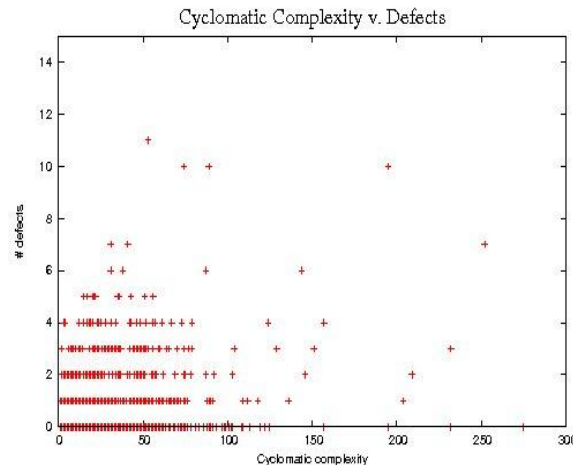


After

Are defects related to static complexity ?

- There is little evidence that complexity measures such as the cyclomatic complexity $v(G)$ are of any use at all in predicting defects

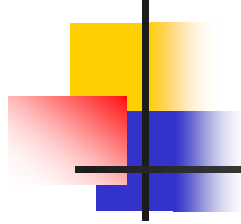
Defects



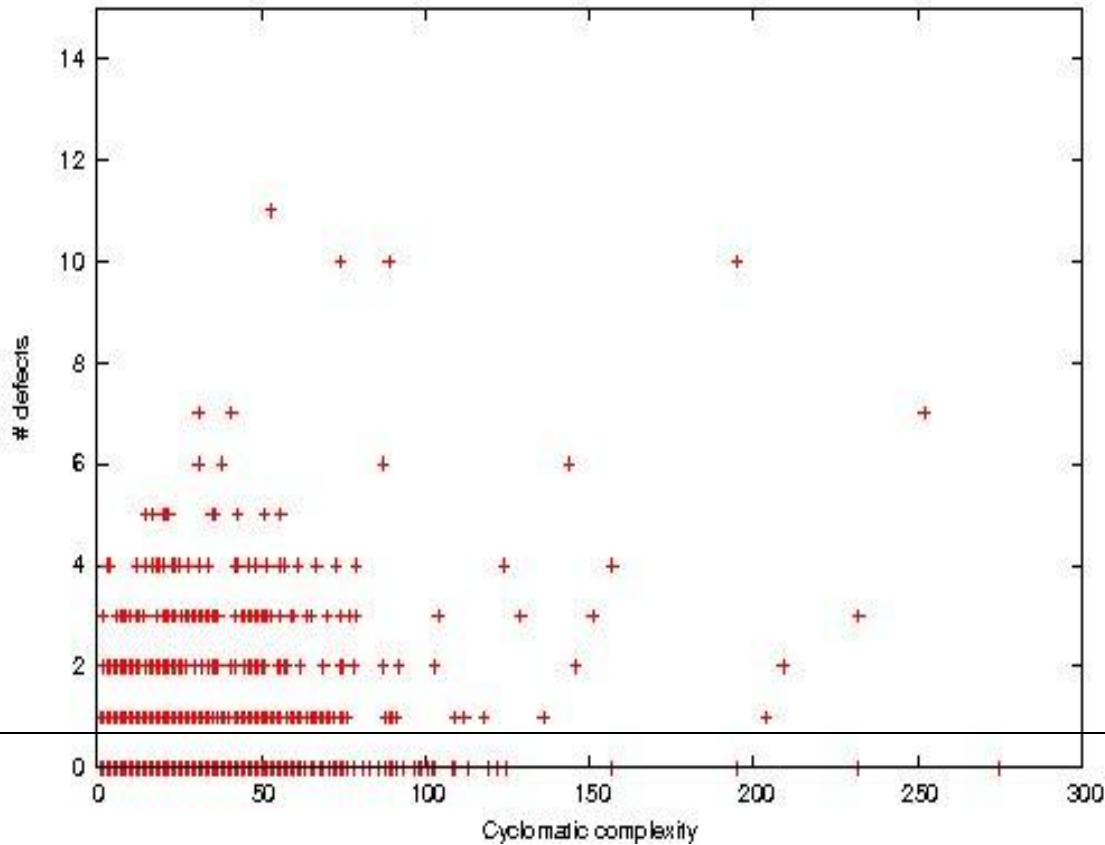
Cyclomatic number $v(G)$

NAG Fortran library over 25 years
(Hopkins and Hatton (2008))

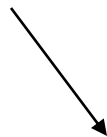
Is there anything unusual about ‘zero’ defect ?



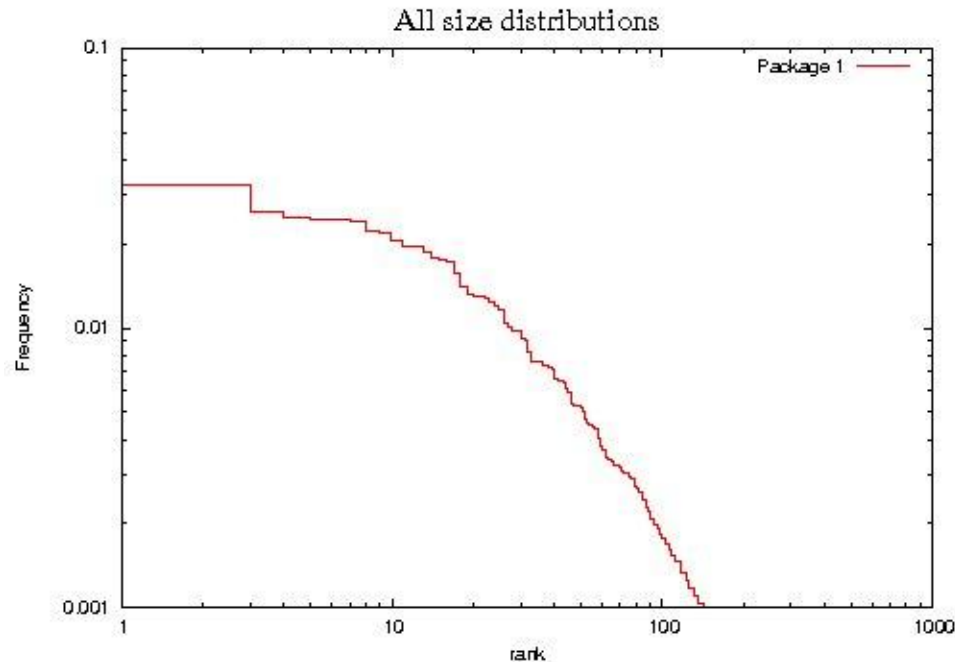
Cyclomatic Complexity v. Defects



PCA and endless rummaging suggest not. This may undermine *root-cause analysis*.



Software size distributions appear power-law in LOC



Smoothed (cdf) data for 21 systems, C, Tcl/Tk and Fortran, combining 603,559 lines of code distributed across 6,803 components, (Hatton 2009, IEEE TSE)

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A tentative model



We are looking for:-

- Language independent behaviour
- Application independent behaviour
- Predicts power-law behaviour in component sizes
- Predicts simple and apparently *power-law* behaviour in defect, (observed frequently)
- Makes other testable predictions.

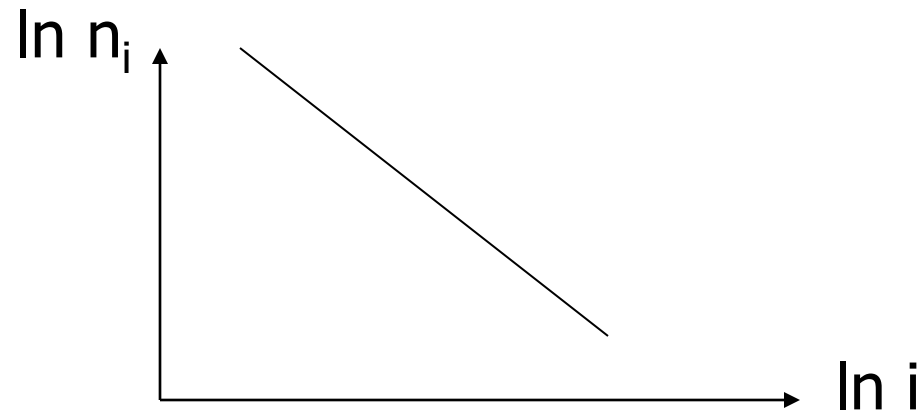
What is power-law behaviour ?

Frequency of occurrence n_i given by $n_i = \frac{nc}{i^p}$

This is usually shown as

$$\ln n_i = \ln(nc) - p \ln i$$

which looks like

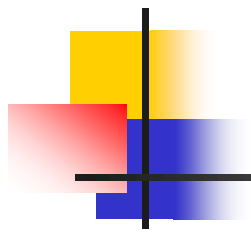


Is power-law behaviour persistent ?

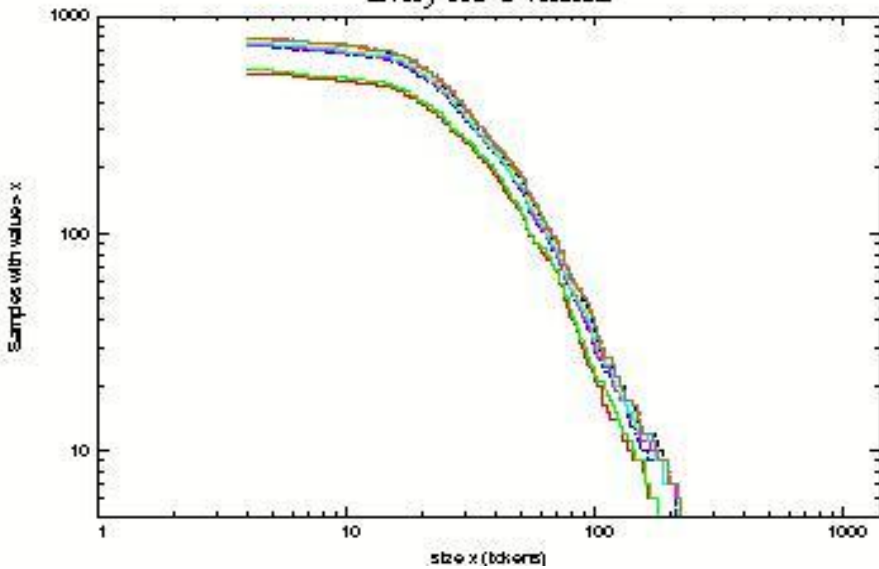


- Question: Does power-law behaviour in component size establish itself over time as a software system matures or is it present at the beginning ?

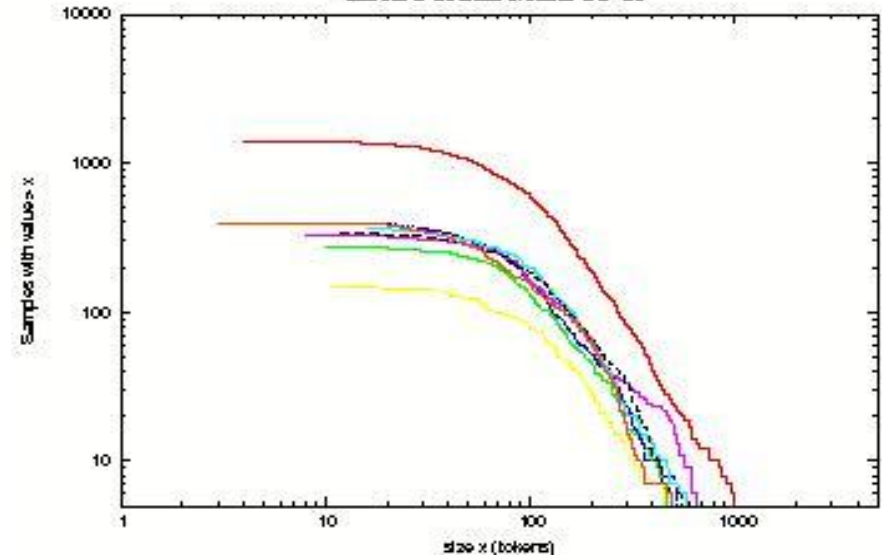
Is power-law behaviour persistent ?



Every 3rd C version



Each Fortran Mark 12-19



Is power-law behaviour persistent ?



- Answer: *Power-law behaviour in component size appears to be present at the beginning of the software life-cycle.*
- Given that this appears independent of programming language and application area, can we explain why ?

Building systems



- When we build a system we are making choices
 - Choices on functionality
 - Choices on architecture
 - Choices on programming language(s)

- There is a general theory of choice – Shannon information theory.

Building systems



- Software component size - approximate
 - *Number of lines of code.* This is quite dependent on the programming language, (consider the influence of the pre-processor in C and C++ for example).
- Software component size - better
 - Based on *tokens of a programming language.*

Building systems from tiny pieces

■ Tokens of language

- *Fixed tokens.* You have no choice in these. There are 49 operators and 32 keywords in ISO C90. Examples include the following in C, (but also in C++, PHP, Java, Perl ...):

{ } [] () if while * + *= == // / , ; :

- *Variable tokens.* You can choose these. Examples include:-
identifier names, constants, strings

- Every computer program is made up of combinations of these, (note also the Boehm-Jacopini theorem (1966)).

A model for emergent power-law size behaviour using Shannon entropy

Suppose component i in a software system has t_i tokens in all constructed from an alphabet of a_i unique tokens.

First we note that

$$a_i = a_f + a_v(i)$$

Fixed tokens of a language, {
} [] ; **while** ...

Variable tokens, (id names
and constants)

A model for emergent power-law size behaviour using Shannon entropy

An example from C:

Fixed
(18)

```
void int ( ) [ ] { , ;
for = >= -- <=
++ if > -
```

+

Variable
(8)

```
bubble a N i j t 1 2
```

```
void bubble( int a[], int N)
{
  int i, j, t;
  for( i = N; i >= 1; i--)
  {
    for( j = 2; j <= i; j++)
    {
      if ( a[j-1] > a[j] )
      {
        t = a[j-1]; a[j-1] = a[j]; a[j] = t;
      }
    }
  }
}
```

Total
(94)

A model for emergent power-law size behaviour using Shannon entropy

For an alphabet a_i the Hartley-Shannon information content density I'_i per token of component i is defined by

$$t_i I'_i \equiv I_i = \log(a_i a_i \dots a_i) = \log(a_i^{t_i}) = t_i \log(a_i)$$

We think of I'_i as fixed by the nature of the algorithm we are implementing.

Consider now building a system as follows

Consider a general software system of T tokens divided into M pieces each with t_i tokens, each piece having an *externally imposed information content density* property I'_i associated with it. *Note: no nesting.*

1	2	3			
			t_i, I'_i			
				...	M	

$$T = \sum_{i=1}^M t_i$$

$$I = \sum_{i=1}^M t_i I'_i$$

General mathematical treatment

The most likely distribution of the I'_i ($= I_i/t_i$) subject to the constraints of T and I held constant

$$T = \sum_{i=1}^M t_i \quad \text{and} \quad I = \sum_{i=1}^M t_i I'_i$$

is

$$p_i \equiv \frac{t_i}{T} = \frac{e^{-\beta I'_i}}{\sum_{i=1}^M e^{-\beta I'_i}}$$

where p_i can be considered *the probability of piece i occurring with a share I_i of I* . β is a constant.

General mathematical treatment

However

$$I'_i = \left(\frac{I_i}{t_i} \right) = \left(\frac{t_i \log(a_i)}{t_i} \right) = \log(a_i)$$

Giving the
general theorem

$$P_i \sim a_i^{-\beta}$$

This states that in any software system, conservation of size and information (i.e. choice) is overwhelmingly likely to produce a power-law alphabet distribution. (Think ergodic here).

One last little bit of maths

- Note that for small components, the fixed token overhead is a much bigger proportion of all tokens, $a_f \gg a_v(i)$, so

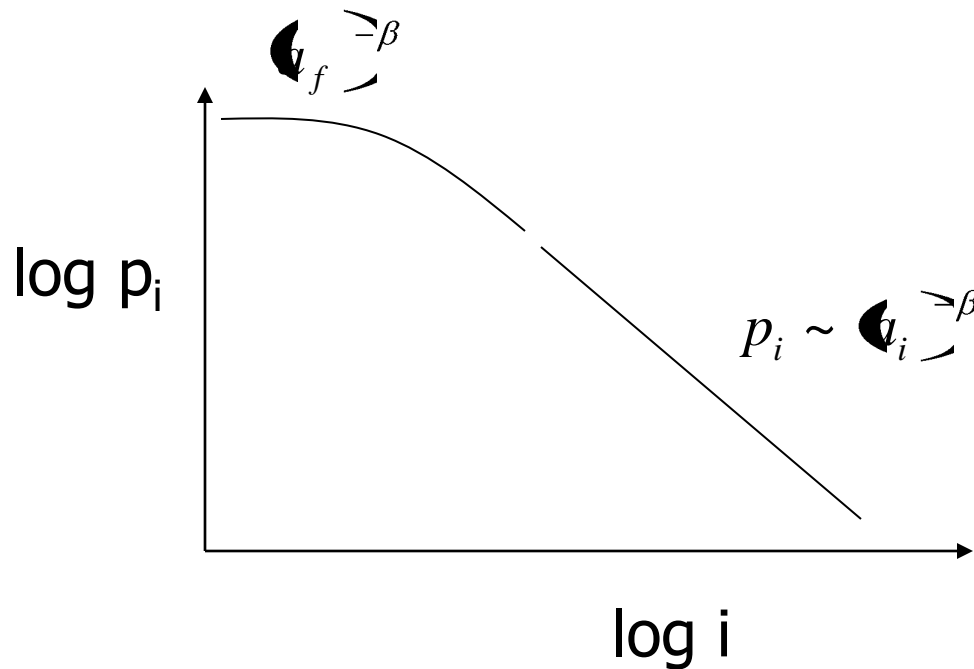
$$p_i = \frac{1}{Q(\beta)} (a_f + a_v(i))^{-\beta} \approx (a_f)^{-\beta} \left(1 + \frac{a_v(i)}{a_f}\right)^{-\beta} \approx (a_f)^{-\beta} \text{ Constant}$$

- For large components, the general rule takes over

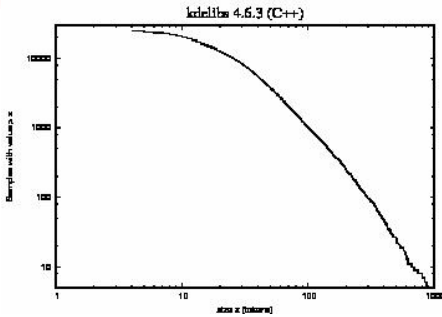
$$p_i \sim a_i^{-\beta}$$

Application to software systems

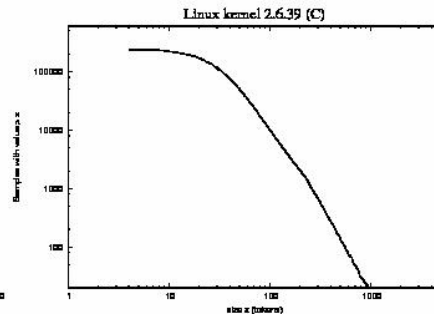
So we are looking for the following signature



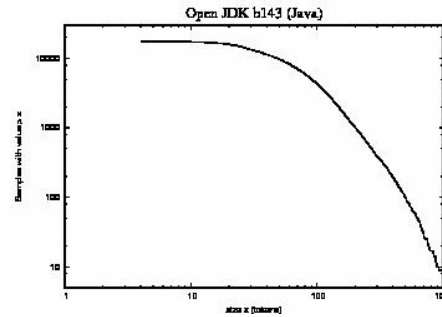
Some results



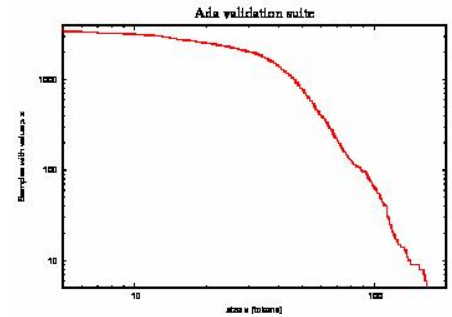
C++



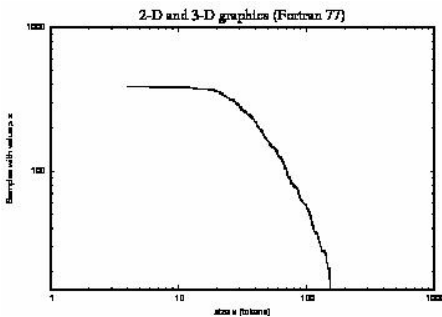
C



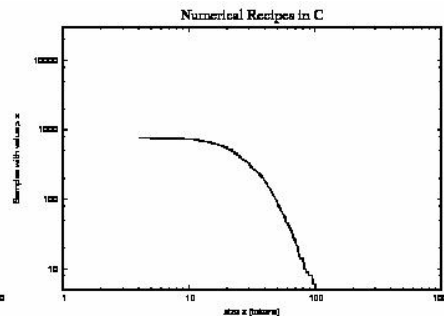
Java



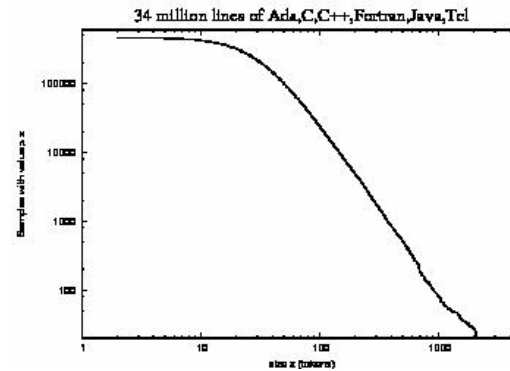
Ada



Fortran



C Numerical



34 million lines of Ada, C, C++, Fortran, Java, Tcl in 75 systems.

Some model predictions

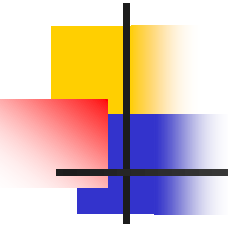
- Suppose there is a constant probability P of making a mistake on any token. The total number of defects is then given by $d_i = P \cdot t_i$. Then

$$p_i = \frac{1}{Q(\beta)} t_i^{-\beta} \approx d_i^{-\beta} \approx t_i^{-\beta}$$

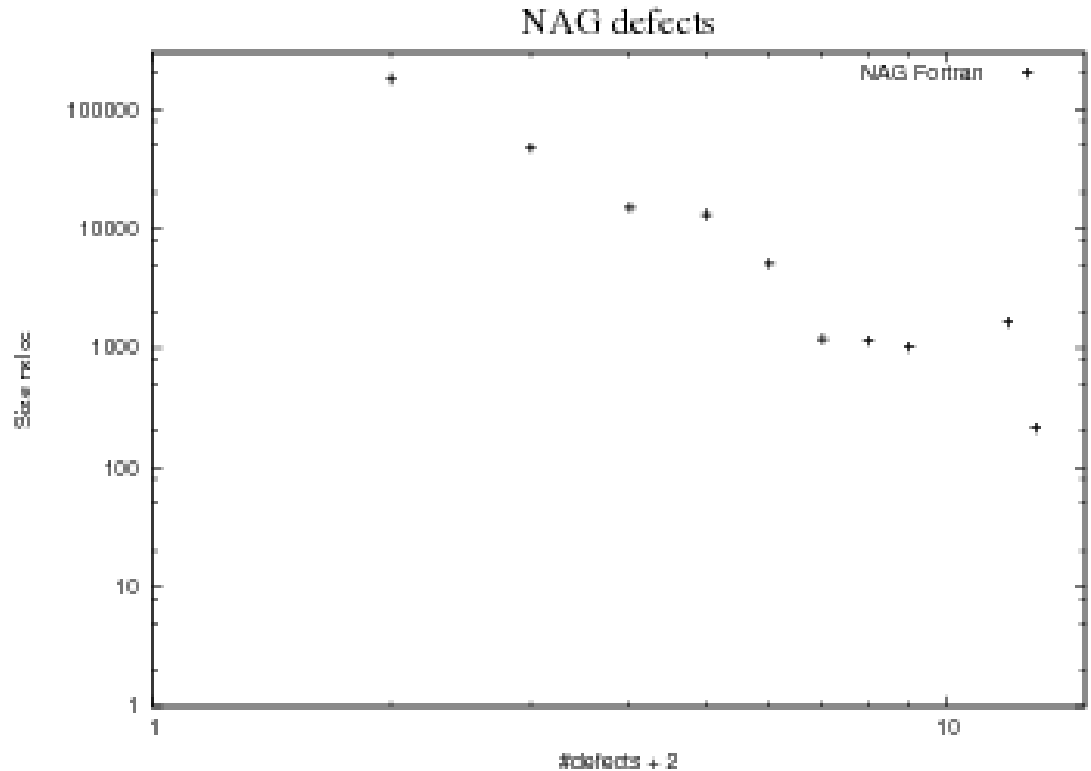
← This step uses Zipf's law, Hatton (2009)

- So defects will also be distributed according to a power-law – *i.e they will cluster.*

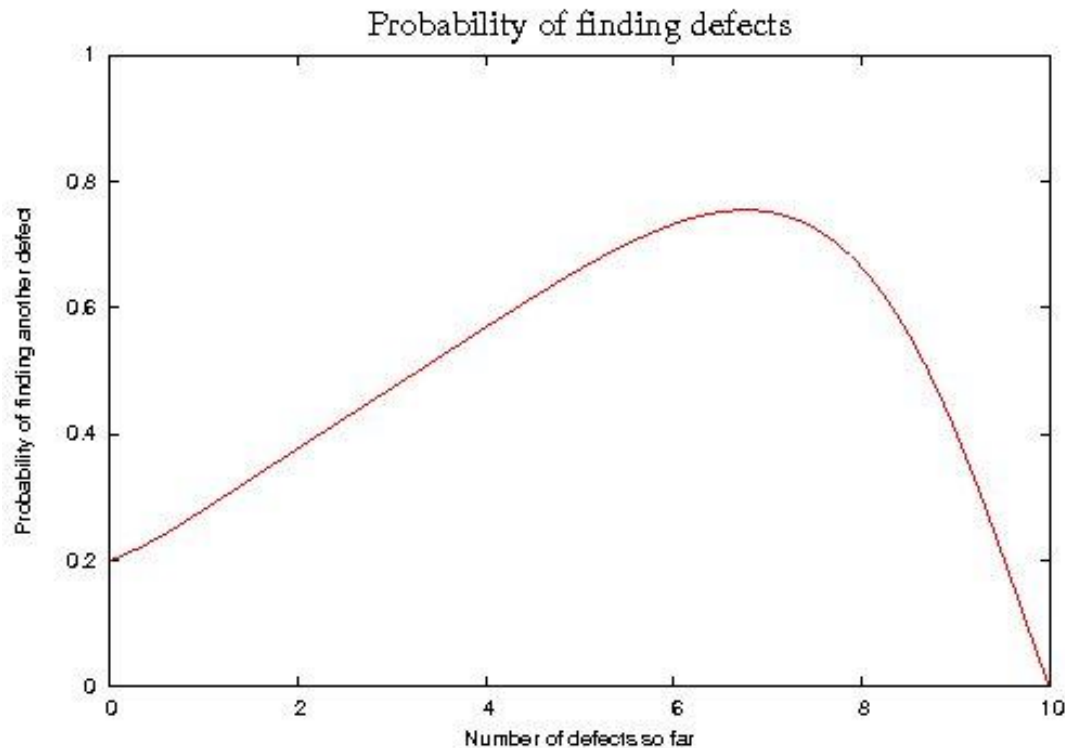
Defect clustering in the NAG Fortran library (over 25 years)



Defects	components	XLOC
0	2865	179947
1	530	47669
2	129	14963
3	82	13220
4	31	5084
5	10	1195
6	4	1153
7	3	1025
> 7	5	1867



Clustering can be exploited: Conditional probability of finding defects*



* See, Hopkins and Hatton (2008), http://www.leshatton.org/NAG01_01-08.html

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Conclusions

- Bounding defects is inherently difficult but N versions (or open source) both seem to offer ways of improving software agreement but by an unknown amount.
- Static structural relationships with defect appear to be a blind alley, (cyclomatic complexity ...).
- Defects cluster and this can be exploited.
- Software systems appear to exhibit macroscopic behaviour independent of implementation or language

$$P_i \sim a_i^{\beta}$$

References



My writing site:-

<http://www.leshatton.org/>

Specifically,

http://www.leshatton.org/variations_2010.html

Thanks for your attention.