

Code Review at Cisco Systems

The largest case study ever done on lightweight code review process; data and lessons.

In May of 2006 Smart Bear Software wrapped up a 10-month case study of peer code review in the Cisco MeetingPlace¹ product group at Cisco Systems, Inc. With 2500 reviews of 3.2 million lines of code written by 50 developers, this is the largest case study ever done on what's known as a "lightweight" code review process.

The subject of almost all published literature on code review is of formal, heavyweight meeting-based inspections. But in recent

¹ At the time of this writing (June 2006) Cisco is running television ads in America touting the advantages of their teleconferencing solution. This is the story of that development group.

years many development organizations have shrugged off the yoke of meeting schedules, paper-based code readings, and tedious metrics-gathering in favor of new lightweight review processes. Certain lightweight processes appear to have the same proven benefits and measurability found in heavyweight processes while drastically reducing total time spent engaged in procedures.

The studies in the previous chapter have already suggested that formal meetings add hours to the process without uncovering additional defects. Furthermore we have found that most developers prefer a faster, more lightweight approach, and managers like the idea of a process nimble enough to be applied to all code changes across the board, not just those dangerous enough to warrant the time investment of a formal inspection.

But you cannot sacrifice code quality. You cannot just throw away 30 years of evidence that heavyweight process works. Where are the numbers to support the effectiveness of a lightweight process, and what guidelines should be followed to ensure an effective review?

The Smart Bear / Cisco study sought to answer exactly those questions. We used real developers working on commercially-available software at an established software company; no students, no contrived code snippets, no sterile laboratory conditions.

Cisco has a long history of using various types of code review as part of their legendary quality control. The MeetingPlace group was no exception. In July 2005, 50 developers in the MeetingPlace group started using a software tool for lightweight review in the hopes that it would increase defect detection while speeding up overall review time and removing some of the drudgery normally associated with inspections.

We'll analyze the results of those reviews and determine the general characteristics of effective, efficient reviews under this system. In the process we will demonstrate that this particular

brand of lightweight review is able to uncover as many defects with as many process metrics in much less time than heavyweight formal inspections.

How reviews were conducted

The reviews were conducted using Smart Bear Software's Code Collaborator system for tool-assisted peer review. Code Collaborator is described in detail and with screenshots in another essay in this collection; here we'll only summarize the process.

Cisco wanted a review before every code change was checked into the version control server, which in their case was Perforce®. They used a Perforce server trigger (included with Code Collaborator) that prevented any code check-in unless a review existed in the Code Collaborator server, and that review was "complete" with all found defects fixed and verified.

Software developers were provided with several Code Collaborator tools allowing them to upload local changes from the command-line, a Windows GUI, or from a plug-in to the Perforce GUI clients P4Win and P4V.

Reviews were performed using Code Collaborator's web-based user interface. Authors determined who was "invited" to be a reviewer or observer; about half the reviews had a single reviewer, the rest two or more. Invitations were sent by Code Collaborator via e-mail.

During the inspection, Code Collaborator presented before/after difference views to all participants. Everyone could comment on any line of code by clicking on the line and typing. Comments are kept threaded and are always visible next to the code in question (see Figure 11).

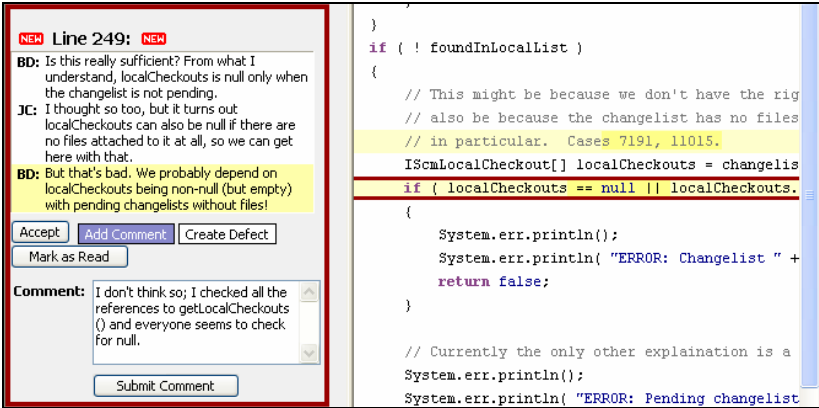


Figure 11: Code Collaborator screenshot showing threaded comments next to Java code under inspection. The author is defending a design decision.

Defects are logged like comments, also threaded by file and line number. When an author believed a defect had been fixed, the new files were uploaded to the same review. The web interface then presents these new changes against the original so reviews can verify that defects were fixed and no new defects opened. This back-and-forth process happens as many times as is necessary for all defects to be fixed.

Once all reviewers agree the review is complete and no defects are still open, the review is complete and the author is then allowed to check the changes into Perforce.

Code Collaborator automatically gathers key review metrics such as man-hours spent in review and lines of code under inspection. It is these metrics, combined with defect logs, that we analyze below.

Thinning the herd

Some reviews in the sample set reflect special cases that we don't wish to analyze in general. There are two specific cases we want to throw out of the analysis:

1. Reviews of enormous amounts of code. If many thousands of lines of code were under review, we can be sure this is not a true code review.
2. Trivial reviews. These are reviews in which clearly the reviewer never looked at the code, or at least not long enough for any real effect. For example, if the entire review took two seconds, clearly no review actually took place.

We can visualize these cases by considering a plot of “lines of code under inspection” against “inspection rate in lines per hour.” From the log-log chart in Figure 12 it is apparent that there are aberrant data points for both enormous LOC and enormous inspection rates.

There are some clear cut-off points for rejecting samples given the data in Figure 12. For example, a 10,000 line-per-hour inspection rate implies the reviewer can read and understand source code at a rate of three lines per second. As another example, a single review of 10,000 lines of code isn't possible. It is also apparent that the majority of reviews appear in much more reasonable ranges.

There are several explanations for these outliers. Because review was required before version control check-in, large unreviewed changes will still pass through the system. This explains for example the reviews of many tens of thousands of lines which are reviewed too quickly to be careful inspections.

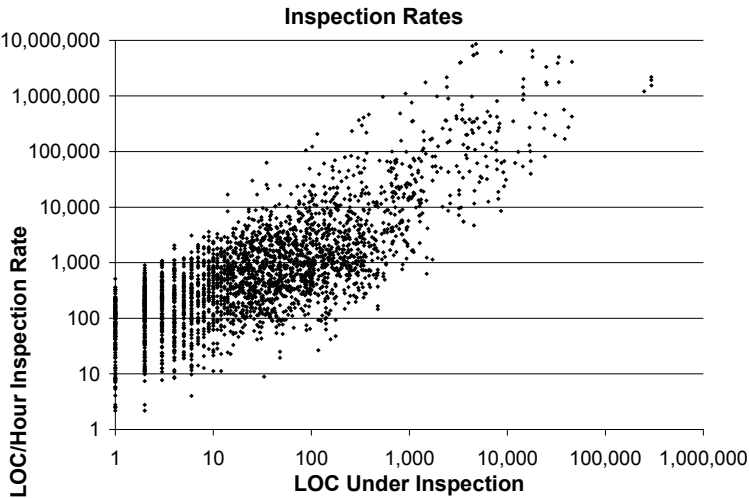


Figure 12: Log-log plot of lines of code (LOC) under inspection versus the speed of inspection (in LOC per hour).

The columnar grouping pattern for small LOC is a result of the LOC variable being an integer and the logarithmic scale expanding that region.

There are also cases of reasonable inspection sizes reviewed faster than is humanly possible. One explanation is the pass-through review – the reviewer simply OK’s the changes without looking at them. Another explanation is that the reviewer and developer communicated about this review outside the system, so by the time the official review came around the reviewer didn’t need to look at the code. In either case we are not interested in data from these lightning-fast reviews.

We therefore make the following rules about throwing out reviews from the study:

1. Throw out reviews whose total duration is shorter than 30 seconds.
2. Throw out reviews where the inspection rate is greater than 1500 LOC/hour.
3. Throw out reviews where the number of lines under review is greater than 2000.

This attempt at isolating “interesting” review cases cuts out 21% of the reviews. The distribution in Figure 13 shows that the most reviews are smaller than 200 lines of code and are inspected slower than 500 LOC/hour.

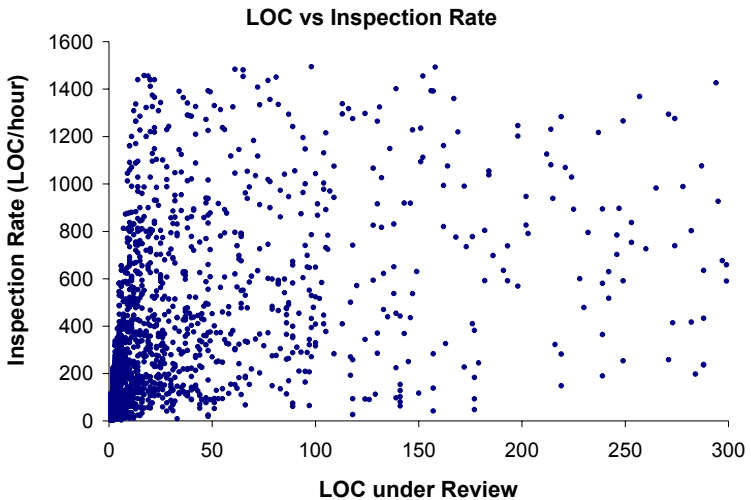


Figure 13: Distribution of reviews after discarding those that cannot represent proper reviews. Most reviews are under 150 lines of code and reviewed slower than 500 LOC/hour.

Inspection Rate Analysis

How fast should code be reviewed? If you go too fast you're liable to miss defects. Industry experts say inspection rates should not exceed 200 lines per hour if you want an effective review. Does review rate differ by reviewer or author or the type of code under inspection?

We might expect a relatively constant inspection rate. That is, it should take twice as long to review 200 lines of code than it does to review 100 lines of code. In general, if we plot code size versus time-to-review, we expect the values to cluster around a line that represents the average review rate. However, Figure 14 shows this is not the case. No clustering around a common rate, not even when we zoom in on the “cluster” of data with reviews under one hour and under 200 lines.

Although this result is unexpected, it's great news for our analysis. It means that in this experiment review inspection rates and sizes vary over a wide range of values, which means we have a good sampling of data to use when answering questions like “Does inspection rate or inspection size affect the number of defects found?” or “What inspection rate makes the reviewer most efficient at finding defects?”

Indeed, the next logical question is: “What are the factors that determine the inspection rate?” Do anal-retentive reviewers agonize over every line? Does the guy with empty Red Bull cans all over his cubicle race through code? Do certain files or modules take longer to review?

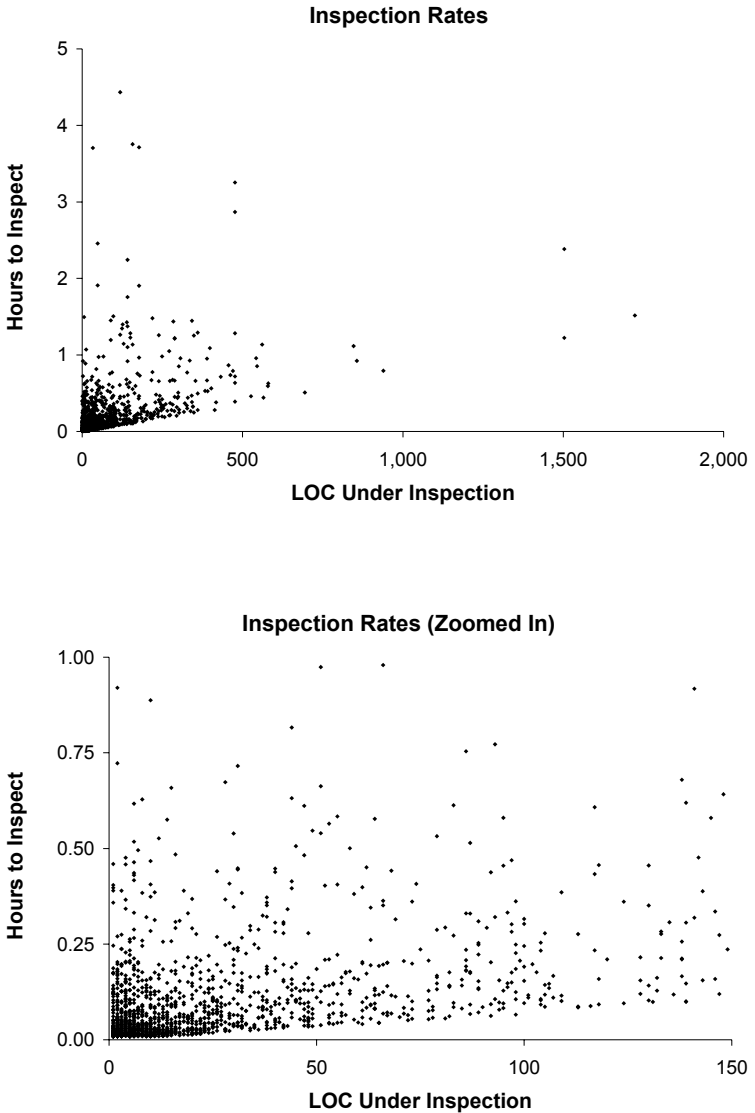


Figure 14: Plotting inspection size versus time, in total and zoomed into the cluster near the origin. There is no apparent systematic "inspection rate." The absence of data points below the invisible line with slope 1/1500 is due to our throwing out reviews with high inspection rates.

Does the inspection rate vary by reviewer?

Do some reviewers zoom through code while others linger? Do your star developers take longer because they are given the hardest code to review? Does the identity of the reviewer make the inspection rate predictable?

Unfortunately the assumptions of ANOVA are not met for these data, so we investigated individual reviewer rates by hand. A typical example is shown in Figure 15 for Reviewer #3. Clearly this reviewer has no one rate².

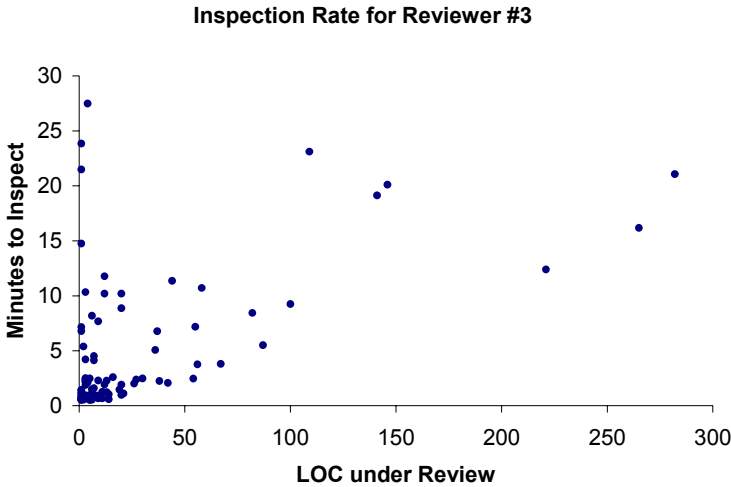


Figure 15: An analysis of inspection rate for Reviewer #3 shows there is no single rate and identifies some interesting special cases along the y-axis.

² The best-fit rate is only $R^2=0.29$.

We did notice something odd. There are four reviews of 1 or 2 lines of code that each took over 15 minutes to complete. The other reviews that took that long had over 100 lines of code! These might be aberrant, and removing aberrant data points might give us a statistically significant inspection rate. So we took a closer look.

Each of these outlier cases was explainable. In one case, a separate review contained the real changes; the reviewer had simply referred back to the first frequently while looking at the second. In all other cases there was a lot of dialog between the reviewer and the author or other reviewers. These code modifications, though small in physical size, all seemed to have significant ramifications for the rest of the system according to the comments.

So after close inspection it was clear that these data points did belong in our data set. And this in turn means that there still is not a clear inspection rate.

Another feature of the single-reviewer graphs (e.g. Figure 15) is the cluster of small-change, fast reviews near the origin, just as we saw with the global inspection rate graphs. And once again, when we zoomed into that area alone it was clear that no particular rule governs inspection rate, even for a single reviewer (see Figure 16).

But occasionally we found a reviewer who seemed to have a more regular inspection rate. Figure 17 shows one example with a decent inspection rate correlation. However these were rare and usually associated with reviewers who hadn't participated in many reviews yet; presumably as they encountered more types of source code they too would start to show a larger spread.

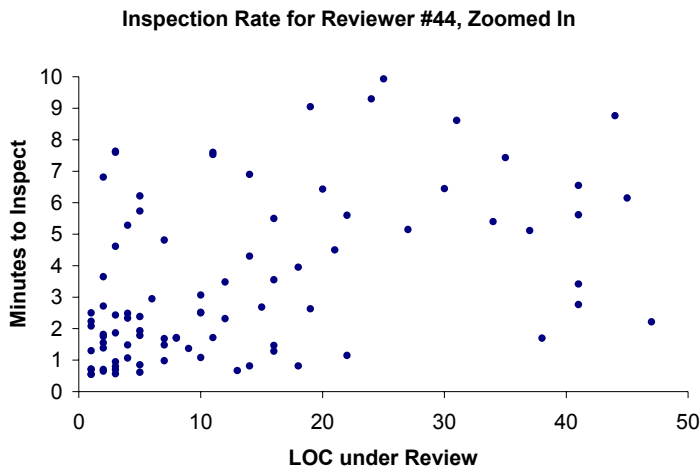


Figure 16: Another example showing no pattern in inspection rate even when zoomed into the mass of data points near the origin.

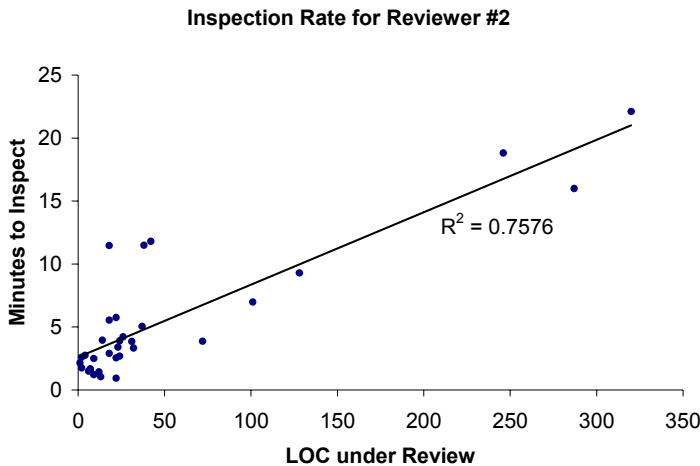


Figure 17: Example of a reviewer who appears to have a consistent inspection rate.

Does the inspection rate vary by author?

So the reviewer doesn't determine the rate, but perhaps the author does. Different authors work on different modules and types of code. Some authors might write code that takes longer to read.

Again, we find the same results (Figure 18): No linear relationship, clustering around the origin.

The column of data points at LOC=141 in needs to be explained. This is review #1174 which happened to have six different (and simultaneous) reviewers. Each participant took a different amount of time to examine the code and talk about it with the others.

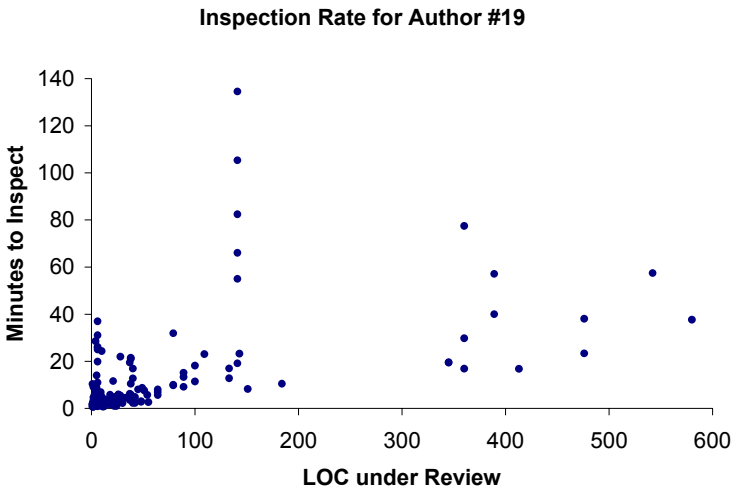


Figure 18: No pattern in per-author inspection rates. The column of points at LOC=141 is explained in the text.

In fact, review #1174 constitutes additional evidence that inspection rate doesn't depend on the reviewer. All six reviewers were examining and chatting about a single review, yet the amount of time spent during the review varied widely.

Conclusion for inspection rate

We found no metric that correlated significantly with inspection rate. It is clear that many factors combine to determine the speed at which a reviewer will scan a set of code changes.

But none of this means all these reviews were equally effective or efficient at finding defects. The literature suggests that slow inspections uncover more defects. But before we can explore review effectiveness we first need to decide what constitutes a "defect."

Counting Defects

What is a "defect?" Before we get into defect rate and density analysis we need to define exactly what a "defect" means and how we will identify defects in our sample data here.

Although the word "defect" has an inherent negative connotation, in code review it is defined this way:

When a reviewer or consensus of reviewers determines that code must be changed before it is acceptable, it is a "defect." If the algorithm is wrong, it's a defect. If the code is right but unintelligible due to poor documentation, it's a defect. If the code is right but there's a better way to do it, it's a defect. A simple conversation is not a defect nor is a conversation where a reviewer believed he found a defect but later agreed that it wasn't one. In any event a defect is an improvement to the code that would not have occurred without review.

Counting defects in Code Collaborator should be easy in theory because the software includes a built-in defect logging system

that not only logs defects against files and line numbers but also allows for a selection of severity and type. Unfortunately this theory does not apply with this data.

In particular, reviewers and authors are free to communicate the existence of a defect without creating a proper defect record in the database. Furthermore, with earlier versions of the software the workflow surrounding defects was confusing, so the path of least resistance was to talk about defects but not necessarily to open them.

Therefore we cannot just use the defect data from the database as a true measure of defects. Instead we took a random sample of 300 reviews and studied the conversations in each one to measure the number of true defects as defined above.

Defect Density Analysis

Almost all code review process analysts want to measure “defect density,” meaning the number of defects found per 1000 lines of code. This number is often associated with review “effectiveness” in that a more effective review will uncover more defects per line of code compared with a cursory review. In a predictive capacity, the density number allows us to answer questions like “How many defects will we expect code review to uncover in 10,000 lines of code?”

Our reviews had an average 32 defects per 1000 lines of code. 61% of the reviews uncovered no defects; of the others the defect density ranged evenly between 10 and 130 defects per kLOC.

Defect density and review size

The relationship between defect density and the amount of code under review is made clear by Figure 19.

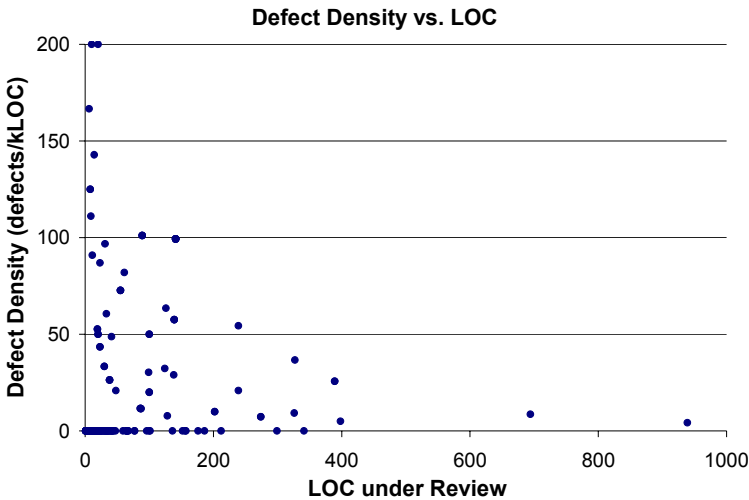


Figure 19: As the amount of code under review increases reviewers become less effective at finding defects assuming a constant true number of defects per kLOC.

Reviewers are most effective at reviewing small amount of code. Anything below 200 lines produces a relatively high rate of defects, often several times the average. After that the results trail off considerably; no review larger than 250 lines produced more than 37 defects per 1000 lines of code³.

³ The critical reader will notice we're tacitly assuming that true defect density is constant over both large and small code changes. That is, we assume a 400-line change necessarily contains four times the number of defects in a 100-line change, and thus if defect densities in code review fall short of this the review must be "less effective." Current literature generally supports this assumption although there are clearly cases where we would naturally expect large code changes to have fewer defects per line, e.g. a new class interface with detailed documentation and no executable code.

These results are not surprising. If the reviewer is overwhelmed with a large quantity of code he won't give the same attention to every line as he might with a small change. He won't be able to explore all the ramifications of the change in a single sitting.

Another explanation comes from the well-established fact that after 60 minutes reviewers "wear out" and stop finding additional defects⁴. Given this, a reviewer will probably not be able to review more than 300-400 lines of code before his performance drops.

But this hypothesis is more directly measurable by considering the inspection rate.

Defect density and inspection rate

It makes sense that reviewers hurried through a review won't find as many defects. A fast inspection rate might mean the reviewer didn't take enough time, or it could mean the reviewer couldn't give enough time for the large quantity of code under review.

The "slower is better" hypothesis is confirmed in Figure 20. Reviewers slower than 400 lines per hour were above average in their ability to uncover defects. But when faster than 450 lines/hour the defect density is below average in 87% of the cases.

⁴ A compelling example of this is given in the survey of case studies essay elsewhere in this collection.

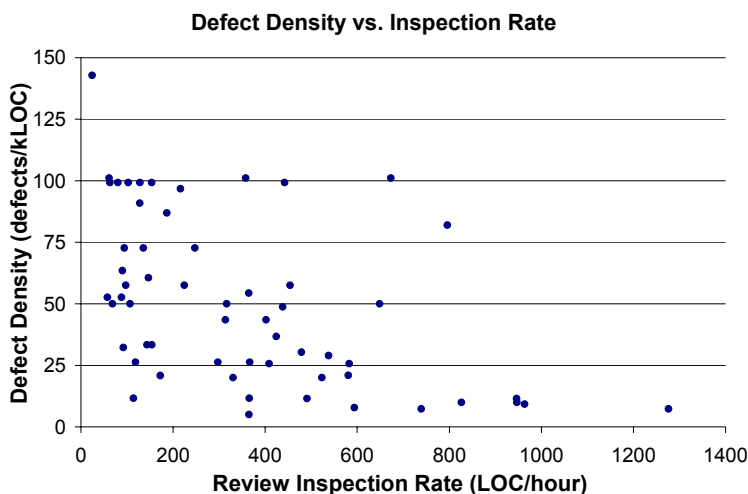


Figure 20: Reviewers become less effective at finding defects as their pace through the review accelerates.

Defect density and author preparation

Could authors eliminate most defects before the review even begins? If we required developers to double-check their work, maybe reviews could be completed faster without compromising code quality. We were able to test this technique at Cisco.

The idea of “author preparation” is that authors should annotate their source code before the review begins. Annotations guide the reviewer through the changes, showing which files to look at first and defending the reason and methods behind each code modification. The theory is that because the author has to re-think all the changes during the annotation process, the author will himself uncover most of the defects before the review even begins, thus making the review itself more efficient. Reviewers will

uncover problems the author truly would not have thought of otherwise.

If author preparation has a real effect it will be to reduce the number of defects found during the inspection. This means a *lower defect density* because in theory the author has already removed most of the defects.

So we tested the hypothesis: “Reviews with author preparation have small defect densities compared to reviews without.” It is easy to detect “author preparation” in our data because we record every comment, threaded by file and line of code. Without author preparation, conversations are typically started by a reviewer or observer and often answered by the author. Author preparation is signified by the author kicking off the conversation. In our manual scan of reviews we found almost no cases where the author started the conversation and yet wasn’t prepping the reviewer.

The relationship between author preparation and defect density is shown in Figure 21. The data supports our hypothesis in two specific ways. First, for all reviews with at least one author preparation comment, defects density is never over 30; in fact the most common case is for there to be no defects at all! Second, reviews without author preparation comments are all over the map whereas author-prepared reviews do not share that variability.

Clearly author preparation is correlated with low defect densities. But there are at least two ways to explain this correlation, each leading to opposite conclusions about whether author preparation should be mandatory.

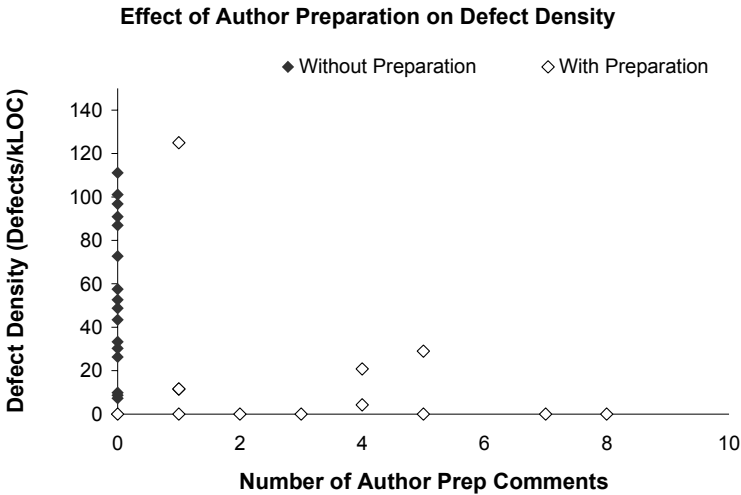


Figure 21: When the author takes the time to prime the review with explanatory comments, the number of defects found is very low and often zero. Without preparatory comments the defect density varies widely.

One conclusion is that the very act of deeply preparing for a review causes the author to identify and correct most defects on his own. The analogous adage is “I read I forget; I see I remember; I teach I understand.” We all have personal experience to back this up; when you’re forced to explain your work to someone else, anticipating their questions and teaching them your techniques, you uncover things you hadn’t thought about before.

The other conclusion is that prepping disables the reviewer’s capacity for criticism. Author comments prime the reviewer for what to expect. As long as the code matches the prose, the reviewer is satisfied. Because the reviewer is guided he doesn’t think outside the box, doesn’t approach the problem fresh, and

doesn't bring new insight to the problem. The reason defect density is low for an author-prepared review is not because the author pre-fixed defects, but rather because the reviewers aren't looking hard enough.

We believe the first conclusion is more tenable. A survey of the reviews in question show the author is being conscientious, careful, and helpful, and not misleading the reviewer. Often the reviewer will respond or ask a question or open a conversation on another line of code, demonstrating that he was not dulled by the author's annotations.

Indeed, we believe these preparation comments belie a fundamental personal development philosophy of attention to detail, consideration of consequences, and general experience. That is, we believe the developers who are naturally meticulous will exhibit this in the form of preparation – it's just another way of expressing their cautious approach. Even with developers who are not naturally this way, we believe that requiring preparation will cause anyone to be more careful, rethink their logic, and write better code overall.

Defect Rate Analysis

Where defect density measures a review's effectiveness, defect rate – defects per hour – measures a review's efficiency. It answers the question "How fast do we uncover defects?"

The overall defect rate was 13 defects per hour with 85% of the reviews slower than 25 defects per hour.

With defect density we determined that large reviews resulted in ineffective reviews. Will a large review also have a detrimental effect on defect rate?

From Figure 22 it is clear that review size does not affect the defect rate. Although the smaller reviews afforded a few especially

high rates, 94% of all reviews had a defect rate under 20 defects per hour regardless of review size.

So reviewers are able to uncover problems at a relatively fixed rate regardless of the size of the task put in front of them. In fact, the take-home point from Figure 22 is that defect rate is constant across all the reviews regardless of external factors.

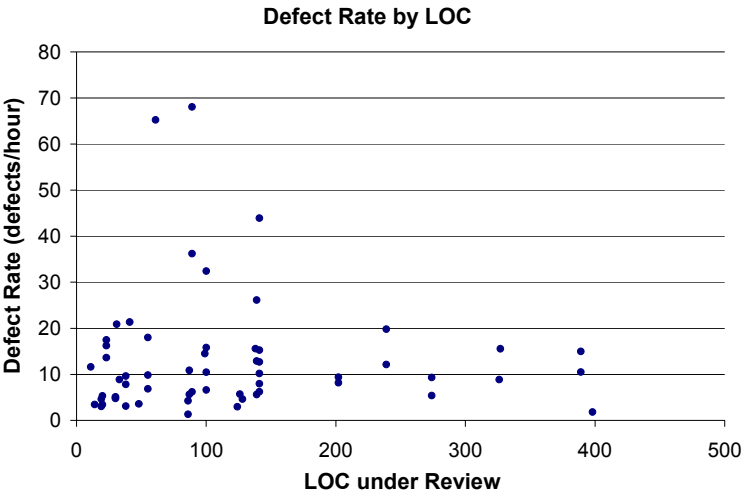


Figure 22: Defect rate is not influenced by the size of the review.

Conclusions

We believe our results allow us to conclude the following:

- LOC under review should be under 200, not to exceed 400. Anything larger overwhelms reviewers and defects are not uncovered.
- Inspection rates less than 300 LOC/hour result in best defect detection. Rates under 500 are still good; expect to miss significant percentage of defects if faster than that.
- Authors who prepare the review with annotations and explanations have far fewer defects than those that do not. We presume the cause to be that authors are forced to self-review the code.
- Total review time should be less than 60 minutes, not exceed 90. Defect detection rates plummet after that time.
- Expect defect rates around 15 per hour. Can be higher only with less than 175 LOC under review.
- Left to their own devices, reviewers' inspection rate will vary widely, even with similar authors, reviewers, files, and size of the review.

Given these factors, the single best piece of advice we can give is to review between 100 and 300 lines of code at a time and spend 30-60 minutes to review it.

Smaller changes can take less time, but always spend at least 5 minutes, even on a single line of code⁵.

⁵ We saw many reviews where a change to a single line of code had ramifications throughout the system.

Lightweight vs. Heavyweight

How do our results compare with those from heavyweight formal inspections? Were our lightweight inspections less effective at uncovering defects? Did they really take less time?

Some of our results exactly match those from established literature. It is well-established that total review time should be under 90 minutes and that slower inspections yield more defects.

Other results are quite different. Across four of the studies of heavyweight inspections given in the previous chapter the average defect detection rate was 2.6 defects per hour⁶; our reviews were *seven times* faster. This is to be expected since our reviews didn't include two-hour inspection meetings with 3-5 participants.

However the critical reader will point out that faster is only better if the same number of defects were uncovered than would have been under a formal inspection process. Unfortunately because this was a study *in situ* and not in a laboratory, we don't know how each of these reviews would have fared with a different process. We can point to the work of Votta and others in the previous chapter for evidence that removing the inspection meetings should not significantly decrease the number of reported defects, but we would have preferred to compare trials of the same code reviewed in both ways⁷.

In light of these other studies, we conclude that lightweight review using Code Collaborator is probably just as effective and definitely more time-efficient than heavyweight formal inspections.

⁶ 0.69 from Blakely 1991, 5.45 from Dunsmore 2000, 1.31 from Conradi 2003, and 3.06 from Kelly 2003.

⁷ We cannot give a single number for “expected defect density” for formal inspection because studies differ widely on this point. For example, Blakely 1991 found 105 defects per kLOC where Laitenberger 1999 found 7 and Kelly 2003 only 0.27!

This is not to say that formal inspections don't have a place in the software development process. Many of our other Code Collaborator customers perform formal inspections instead of or on top of lightweight reviews. Heavyweight process simply takes too much time to be practical with most code changes; here the lightweight process provides measurable, respectable results fast enough to be realistically applied during almost every part of the application development lifecycle.

Future Study

We would like to compare heavyweight and lightweight reviews on the same set of code. We would like to experiment with specific rules of review to see how we might improve defect density or defect rate numbers. Would an enforced minimum inspection-time rule increase defect densities? Would enforcing author preparation comments result in more defects detected in less time? Would reviewer-training result in better defect detection? Would a per-file-type or per-author checklist improve defect detection?

We are currently looking for development groups who would like to participate in future studies where some of these conclusions can be tested directly.

Check with the Smart Bear website for new studies, and please let us know if you would like to participate in one yourself.

