Optimizing Your Programs

Chapter 25

Since program optimization is generally one of the last steps in software development, it is only fitting to discuss program optimization in the last chapter of this text. Scanning through other texts that cover this subject, you will find a wide variety of opinions on this subject. Some texts and articles ignore instruction sets altogether and concentrate on finding a better algorithm. Other documents assume you've already found the best algorithm and discuss ways to select the "best" sequence of instructions to accomplish the job. Others consider the CPU architecture and describe how to "count cycles" and pair instructions (especially on superscalar processors or processes with pipelines) to produce faster running code. Others, still, consider the system architecture, not just the CPU architecture, when attempting to decide how to optimize your program. Some authors spend a lot of time explaining that their method is the "one true way" to faster programs. Others still get off on a software engineering tangent and start talking about how time spent optimizing a program isn't worthwhile for a variety of reasons. Well, this chapter is not going to present the "one true way," nor is it going to spend a lot of time bickering about certain optimization techniques. It will simply present you with some examples, options, and suggestions. Since you're on your own after this chapter, it's time for you to start making some of your own decisions. Hopefully, this chapter can provide suitable information so you can make correct decisions.

25.0 Chapter Overview

25.1 When to Optimize, When Not to Optimize

The optimization process is not cheap. If you develop a program and then determine that it is too slow, you may have to redesign and rewrite major portions of that program to get acceptable performance. Based on this point alone, the world often divides itself into two camps – those who optimize early and those who optimize late. Both groups have good arguements; both groups have some bad arguements. Let's take a look at both sides of this arguement.

The "optimize late" (OL) crowd uses the 90/10 arguement: 90% of a program's execution time is spent in 10% of the code¹. If you try to optimize every piece of code you write (that is, optimize the code before you know that it needs to be optimized), 90% of your effort will go to waste. On the other hand, if you write the code in a normal fashion first and then go in an optimize, you can improve your program's performance with less work. After all, if you *completely removed* the 90% portion of your program, your code would only run about 10% faster. On the other hand, if you completely remove that 10% portion, your program will run about 10 times faster. The math is obviously in favor of attacking the 10%. The OL crowd claims that you should write your code with only the normal attention to performance (i.e., given a choice between an $O(n^2)$ and an $O(n \lg n)$ algorithm, you should choose the latter). Once the program is working correctly you can go back and concentrate your efforts on that 10% of the code that takes all the time.

The OL arguements are persuasive. Optimization is a laborious and difficult process. More often that not there is no clear-cut way to speed up a section of code. The only way to determine which of several different options is better is to actually code them all up and compare them. Attempting to do this on the entire program is impractical. However, if you can find that 10% of the code and optimize that, you've reduced your workload by 90%, very inviting indeed. Another good arguement the OL group uses is that few programmers are capable of anticipating where the time will be spent in a program. Therefore, the only real way to determine where a program spends its time is to *instrument it* and measure which functions consume the most time. Obviously, you must have a working program before you can do this. Once

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^{1.} Some people prefer to call this the 80/20 rule: 80% of the time is spent in 20% of the code, to be safer in their esitmates. The exact numbers don't matter. What is important is that most of a program's execution time is spent in a small amount of the code.

again, they argue that any time spent optimizing the code beforehand is bound to be wasted since you will probably wind up optimizing that 90% that doesn't need it.

There are, however, some very good counter arguments to the above. First, when most OL types start talking about the 90/10 rule, there is this implicit suggestion that this 10% of the code appears as one big chunk in the middle of the program. A good programmer, like a good surgeon, can locate this malignant mass, cut it out, and replace with with something much faster, thus boosting the speed of your program with only a little effort. Unfortunately, this is not often the case in the real world. In real programs, that 10% of the code that takes up 90% of the execution time is often spread all over your program. You'll get 1% here, 0.5% over there, a "gigantic" 2.5% in one function, and so on. Worse still, optimizing 1% of the code within one function often requires that you modify some of the other code as well. For example, rewriting a function (the 1%) to speed it up quite a bit may require changing the way you pass parameters to that function. This may require rewriting several sections of code outside that slow 10%. So often you wind up rewriting much more than 10% of the code in order to speed up that 10% that takes 90% of the time.

Another problem with the 90/10 rule is that it works on percentages, and the percentages change during optimization. For example, suppose you located a single function that was consuming 90% of the execution time. Let's suppose you're Mr. Super Programmer and you managed to speed this routine up by a factor of two. Your program will now take about 55% of the time to run before it was optimized². If you triple the speed of this routine, your program takes a total of 40% of the original time to execution. If you are really great and you manage to get that function running nine times faster, your program now runs in 20% of the original time, i.e., five times faster.

Suppose you could get that function running nine times faster. Notice that the 90/10 rule no longer applies to your program. 50% of the execution time is spent in 10% of your code, 50% is spent in the other 90% of your code. And if you've managed to speed up that one function by 900%, it is very unlikely you're going to squeeze much more out of it (unless it was *really* bad to begin with). Is it worthwhile messing around with that other 90% of your code? You bet it is. After all, you can improve the performance of your program by 25% if you double the speed of that other code. Note, however, that you only get a 25% performance boost *after* you optimized the 10% as best you could. Had you optimized the 90% of your program first, you would only have gotten a 5% performance improvement; hardly something you'd write home about. Nonetheless, you can see some situations where the 90/10 rule obviously doesn't apply and you can see some cases where optimizing that 90% can produce a good boost in performance. The OL group will smile and say "see, that's the benefit of optimizing late, you can optimize in stages and get just the right amount of optimization you need."

The optimize early (OE) group uses the flaw in percentage arithmetic to point out that you will probably wind up optimizing a large portion of your program anyway. So why not work all this into your design in the first place? A big problem with the OL strategy is that you often wind up designing and writing the program twice – once just to get it functional, the second time to make it practical. After all, if you're going to have to rewrite that 90% anyway, why not write it fast in the first place? The OE people also point out that although programmers are notoriously bad at determining where a program spends most of its time, there are some obvious places where they know there will be performance problems. Why wait to discover the obvious? Why not handle such problem areas early on so there is less time spent measuring and optimizing that code?

Like so many other arguements in Software Engineering, the two camps become quite polarized and swear by a totally pure approach in either direction (either all OE or all OL). Like so many other arguements in Computer Science, the truth actually lies somewhere between these two extremes. Any project where the programmer set out to design the perfect program without worry about performance until the end is doomed. Most programmers in this scenario write *terribly slow* code. Why? Because it's easier to do so and they can always "solve the performance problem during the optimization phase." As a result, the 90% portion of the program is often so slow that even if the time of the other 10% were reduced to zero,

^{2.} Figure the 90% of the code originally took one unit of time to execute and the 10% of the code originally took nine units of time to execute. If we cut the execution time of the 10% in half, we now have 1 unit plus 4.5 units = 5.5 units out of 10 or 55%.

the program would still be way too slow. On the other hand, the OE crowd gets so caught up in writing the best possible code that they miss deadlines and the product may never ship.

There is one undeniable fact that favors the OL arguement – optimized code is difficult to understand and maintain. Furthermore, it often contains bugs that are not present in the unoptimized code. Since incorrect code is unacceptable, even if it does run faster, one very good arguement against optimizing early is the fact that testing, debugging, and quality assurance represent a large portion of the program development cycle. Optimizing early may create so many additional program errors that you lose any time saved by not having to optimize the program later in the development cycle.

The correct time to optimize a program is, well, at the correct time. Unfortunately, the "correct time" varies with the program. However, the first step is to develop program performance requirements along with the other program specifications. The system analyst should develop target response times for all user interactions and computations. During development and testing, programmers have a target to shoot for, so they can't get lazy and wait for the optimization phase before writing code that performs reasonably well. On the other hand, they also have a target to shoot for and once the code is running fast enough, they don't have to waste time, or make their code less maintainable; they can go on and work on the rest of the program. Of course, the system analyst could misjudge performance requirements, but this won't happen often with a good system design.

Another consideration is when to perform *what*. There are several types of optimizations you can perform. For example, you can rearrange instructions to avoid hazards to double the speed of a piece of code. Or you could choose a different algorithm that could run twice as fast. One big problem with optimization is that it is not a single process and many types of optimizations are best done later rather than earlier, or vice versa. For example, choosing a good algorithm is something you should do early on. If you decide to use a better algorithm *after* implementing a poor one, most of the work on the code implementing the old algorithm is lost. Likewise, instruction scheduling is one of the last optimizations you should do. Any changes to the code after rearranging instructions for performance may force you to spend time rearranging them again later. Clearly, the lower level the optimization (i.e., relying upon CPU or system parameters), the later the optimization should be. Conversely, the higher level the optimization (e.g., choice of algorithm), the sooner should be the optimization. In all cases, though, you should have target performance values in mind while developing code.

25.2 How Do You Find the Slow Code in Your Programs?

Although there are problems with the 90/10 rule, the concept behind it is basically solid – programs tend to spend a large amount of their time executing only a small percentage of the code. Clearly, you should optimize the slowest portion of your code first. The only problem is how does one find the slowest code in a program?

There are four common techniques programmers use to find the "hot spots" (the places where programs spend most of their time). The first is by trial and error. The second is to optimize everything. The third is to analyze the program. The fourth is to use a *profiler* or other software monitoring tool to measure the performance of various parts of a program. After locating a hot spot, the programmer can attempt to analyze that section of the program.

The trial and error technique is, unfortunately, the most common strategy. A programmer will speed up various parts of the program by making educated guesses about where it is spending most of its time. If the programmer guesses right, the program will run much faster after optimization. Experienced programmers often use this technique successfully to quickly locate and optimize a program. When the programmer guesses correctly, this technique minimizes the amount of time spent looking for hot spots in a program. Unfortunately, most programmers make fairly poor guesses and wind up optimizing the wrong sections of code. Such effort often goes to waste since optimizing the *wrong* 10% will not improve performance significantly. One of the prime reasons this technique fails so often is that it is often the first choice of inexperienced programmers who cannot easily recognize slow code. Unfotunately, they are probably unaware of other techniques, so rather than try a structured approach, they start making (often) uneducated guesses.

Another way to locate and optimize the slow portion of a program is to optimize everything. Obviously, this technique does not work well for large programs, but for short sections of code it works reasonably well. Later, this text will provide a short example of an optimization problem and will use this technique to optimize the program. Of course, for large programs or routines this may not be a cost effective approach. However, where appropriate it can save you time while optimizing your program (or at least a portion of your program) since you will not need to carefully analyze and measure the performance of your code. By optimizing everything, you are sure to optimize the slow code.

The analysis method is the most difficult of the four. With this method, you study your code and determine where it will spend most of its time based on the data you expect it to process. In theory, this is the best technique. In practice, human beings generally demonstrate a distaste for such analysis work. As such, the analysis is often incorrect or takes too long to complete. Furthermore, few programmers have much experience studying their code to determine where it is spending most of its time, so they are often quite poor at locating hot spots by studying their listings when the need arises.

Despite the problems with program analysis, this is the first technique you should always use when attempting to optimize a program. Almost all programs spend most of their time executing the body of a loop or recursive function calls. Therefore, you should try to locate all recursive function calls and loop bodies (especially nested loops) in your program. Chances are very good that a program will be spending most of its time in one of these two areas of your program. Such spots are the first to consider when optimizing your programs.

Although the analytical method provides a good way to locate the slow code in a program, analyzing program is a slow, tedious, and boring process. It is very easy to completely miss the most time consuming portion of a program, especially in the presence of indirectly recursive function calls. Even locating time consuming nested loops is often difficult. For example, you might not realize, when looking at a loop within a procedure, that it is a nested loop by virtue of the fact that the calling code executes a loop when calling the procedure. In theory, the analytical method should always work. In practice, it is only marginally successful given that fallible humans are doing the analysis. Nevertheless, some hot spots are easy to find through program analysis, so your first step when optimizing a program should be analysis.

Since programmers are notoriously bad at analyzing programs to find their hot spots, it would make since to try an automate this process. This is precisely what a *profiler* can do for you. A profiler is a small program that measures how long your code spends in any one portion of the program. A profiler typically works by interrupting your code periodically and noting the return address. The profiler builds a histogram of interrupt return addresses (generally rounded to some user specified value). By studying this histogram, you can determine where the program spends most of its time. This tells you which sections of the code you need to optimize. Of course, to use this technique, you will need a profiler program. Borland, Microsoft, and several other vendors provide profilers and other optimization tools.

25.3 Is Optimization Necessary?

Except for fun and education, you should never approach a project with the attitude that you are going to get maximal performance out of your code. Years ago, this was an important attitude because that's what it took to get anything decent running on the slow machines of that era. Reducing the run time of a program from ten minutes to ten seconds made many programs commercially viable. On the other hand, speeding up a program that takes 0.1 seconds to the point where it runs in a millisecond is often pointless. You will waste a lot of effort improving the performance, yet few people will notice the difference.

This is not to say that speeding up programs from 0.1 seconds to 0.001 seconds is never worthwhile. If you are writing a data capture program that requires you to take a reading every millisecond, and it can only handle ten readings per second as currently written, you've got your work cut out for you. Furthermore, even if your program runs fast enough already, there are reasons why you would want to make it run twice as fast. For example, suppose someone can use your program in a multitasking environment. If you modify your program to run twice as fast, the user will be able to run another program along side yours and not notice the performance degradation.

However, the thing to always keep in mind is that you need to write software that is *fast enough*. Once a program produces results instantaneously (or so close to instantaneous that the user can't tell), there is little need to make it run any faster. Since optimization is an expensive and error prone process, you want to avoid it as much as possible. Writing programs that run faster than fast enough is a waste of time. However, as is obvious from the set of bloated application programs you'll find today, this really isn't a problem, most programming produce code that is way too slow, not way too fast.

A common reason stated for not producing optimal code is advancing hardware design. Many programmers and managers feel that the high-end machines they develop software on today will be the mid-range machines two years from now when they finally release their software. So if they design their software to run on today's very high-end machines, it will perform okay on midrange machines when they release their software.

There are two problems with the approach above. First, the operating system running on those machines two years from now will gobble a large part of the machine's resources (including CPU cycles). It is interesting to note that today's machines are hundreds of times faster than the original 8088 based PCs, yet many applications actually run *slower* than those that ran on the original PC. True, today's software provides many more features beyond what the original PC provided, but that's the whole point of this arguement – customers will demand features like multiple windows, GUI, pull-down menus, etc., that all consume CPU cycles. You cannot assume that newer machines will provide extra clock cycles so your slow code will run faster. The OS or user interface to your program will wind up eating those extra available clock cycles.

So the first step is to realistically determine the performance requirements of your software. Then write your software to meet that performance goal. If you fail to meet the performance requirements, then it is time to optimize your program. However, you shouldn't waste additional time optimizing your code once your program meets or exceed the performance specifications.

25.4 The Three Types of Optimization

There are three forms of optimization you can use when improving the performance of a program. They are choosing a better algorithm (high level optimization), implementing the algorithm better (a medium level optimization), and "counting cycles" (a low level optimization). Each technique has its place and, generally, you apply them at different points in the development process.

Choosing a better algorithm is the most highly touted optimization technique. Alas it is the technique used least often. It is easy for someone to announce that you should always find a better algorithm if you need more speed; but finding that algorithm is a little more difficult. First, let us define an algorithm change as using a fundamentally different technique to solve the problem. For example, switching from a "bubble sort" algorithm to a "quick sort" algorithm is a good example of an algorithm change. Generally, though certainly not always, changing algorithms means you use a program with a better Big-Oh function³ For example, when switching from the bubble sort to the quick sort, you are swapping an algorithm with an $O(n^2)$ running time for one with an $O(n \lg n)$ expected running time.

You must remember the restrictions on Big-Oh functions when comparing algorithms. The value for *n* must be sufficiently large to mask the effect of hidden constant. Furthermore, Big-Oh analysis is usually *worst-case* and may not apply to your program. For example, if you wish to sort an array that is "nearly" sorted to begin with, the bubble sort algorithm is usually much faster than the quicksort algorithm, regard-

^{3.} Big-Oh function are approximations of the running time of a program.

less of the value for *n*. For data that is almost sorted, the bubble sort runs in almost O(n) time whereas the quicksort algorithm runs in $O(n^2)$ time⁴.

The second thing to keep in mind is the constant itself. If two algorithms have the same Big-Oh function, you cannot determine any difference between the two based on the Big-Oh analysis. This does not mean that they will take the same amount of time to run. Don't forget, in Big-Oh analysis we throw out all the low order terms and multiplicative constants. The asymptotic notation is of little help in this case.

To get truly phenomenal performance improvements requires an algorithmic change to your program. However, discovering an O(n lg n) algorithm to replace your O(n^2) algorithm is often difficult if a published solution does not already exist. Presumably, a well-designed program is not going to contain many obvious algorithms you can dramatically improve (if they did, they wouldn't be well-designed, now, would they?). Therefore, attempting to find a better algorithm may not prove successful. Nevertheless, it is always the first step you should take because the following steps operate on the algorithm you have. If you perform the other steps on a bad algorithm and then discover a better algorithm later, you will have to repeat these time-consumings steps all over again on the new algorithm.

There are two steps to discovering a new algorithms: research and development. The first step is to see if you can find a better solution in the existing literature. Failing that, the second step is to see if you can develop a better algorithm on your own. The key thing is to budget an appropriate amount of time to these two activities. Research is an open-ended process. You can always read one more book or article. So you've got to decide how much time you're going to spend looking for an existing solution. This might be a few hours, days, weeks, or months. Whatever you feel is cost-effective. You then head to the library (or your bookshelf) and begin looking for a better solution. Once your time expires, it is time to abandon the research approach unless you are sure you are on the right track in the material you are studying. If so, budget a little more time and see how it goes. At some point, though, you've got to decide that you probably won't be able to find a better solution and it is time to try to develop a new one on your own.

While searching for a better solution, you should study the papers, texts, articles, etc., exactly as though you were studying for an important test. While it's true that much of what you study will not apply to the problem at hand, you are learning things that will be useful in future projects. Furthermore, while someone may not provide the solution you need, they may have done some work that is headed in the same direction that you are and could provide some good ideas, if not the basis, for your own solution. However, you must always remember that the job of an engineer is to provide a cost-effective solution to a problem. If you waste too much time searching for a solution that may not appear anywhere in the literature, you will cause a cost overrun on your project. So know when it's time to "hang it up" and get on with the rest of the project.

Developing a new algorithm on your own is also open-ended. You could literally spend the rest of your life trying to find an efficient solution to an intractible problem. So once again, you need to budget some time for this process accordingly. Spend the time wisely trying to develop a better solution to your problem, but once the time is exhausted, it's time to try a different approach rather than waste any more time chasing a "holy grail."

Be sure to use all resources at your disposal when trying to find a better algorithm. A local university's library can be a big help. Also, you should network yourself. Attend local computer club meetings, discuss your problems with other engineers, or talk to interested friends, maybe they're read about a solution that you've missed. If you have access to the Internet, BIX, Compuserve, or other technically oriented on-line services or computerized bulletin board systems, by all means post a message asking for help. With literally millions of users out there, if a better solution exists for your problem, someone has probabaly solved it for you already. A few posts may turn up a solution you were unable to find or develop yourself.

At some point or another, you may have to admit failure. Actually, you may have to admit success – you've already found as good an algorithm as you can. If this is still too slow for your requirements, it may be time to try some other technique to improve the speed of your program. The next step is to see if you

^{4.} Yes, O(n²). The O(n lg n) rating commonly given the quicksort algorithm is actually the *expected* (average case) analysis, not the worst case analysis.

can provide a better implementation for the algorithm you are using. This optimization step, although independent of language, is where most assembly language programmers produce dramatic performance improvements in their code. A better implementation generally involves steps like unrolling loops, using table lookups rather than computations, eliminating computations from a loop whose value does not change within a loop, taking advantage of machine idioms (such as using a shift or shift and add rather than a multiplication), trying to keep variables in registers as long as possible, and so on. It is surprising how much faster a program can run by using simple techniques like those whose descriptions appear thoughout this text.

As a last resort, you can resort to *cycle counting*. At this level you are trying to ensure that an instruction sequence uses as few clock cycles as possible. This is a difficult optimization to perform because you have to be aware of how many clock cycles each instruction consumes, and that depends on the instruction, the addressing mode in use, the instructions around the current instruction (i.e., pipelining and superscalar effects), the speed of the memory system (wait states and cache), and so on. Needless to say, such optimizations are very tedious and require a very careful analysis of the program and the system on which it will run.

The OL crowd always claims you should put off optimization as long as possible. These people are generally talking about this last form of optimization. The reason is simple: any changes you make to your program after such optimizations may change the interaction of the instructions and, therefore, their execution time. If you spend considerable time scheduling a sequence of 50 instructions and then discover you will need to rewrite that code for one reason or another, all the time you spent carefully scheduling those instructions to avoid hazards is lost. On the other hand, if you wait until the last possible moment to make such optimizations to you code, you will only optimize that code once.

Many HLL programmers will tell you that a good compiler can beat a human being at scheduling instructions and optimizing code. This isn't true. A good compiler will beat a mediocre assembly language program a good part of the time. However, a good compiler won't stand a chance against a good assembly language programmer. After all, the worst that could happen is that the good assembly language programmer will look at the output of the compiler and improve on that.

"Counting cycles" can improve the performance of your programs. On the average, you can speed up your programs by a factor of 50% to 200% by making simple changes (like rearranging instructions). That's the difference between an 80486 and a Pentium! So you shouldn't ignore the possibility of using such optimizations in your programs. Just keep in mind, you should do such optimizations last so you don't wind up redoing them as your code changes.

The rest of this chapter will concentrate on the techniques for improving the implementation of an algorithm, rather than designing a better algorithm or using cycle counting techniques. Designing better algorithms is beyond the scope of this manual (see a good text on algorithm design). Cycle counting is one of those processes that differs from processor to processor. That is, the optimization techniques that work well for the 80386 fail on a 486 or Pentium chip, and vice versa. Since Intel is constantly producing new chips, requring different optimization techniques, listing those techniques here would only make that much more material in this book outdated. Intel publishes such optimization hints in their processor programmer reference manuals. Articles on optimizing assembly language programs often appear in technical magazines like Dr. Dobb's Journal, you should read such articles and learn all the current optimization techniques.

25.5 Improving the Implementation of an Algorithm

One easy way to partially demonstrate how to optimize a piece of code is to provide an example of some program and the optimization steps you can apply to that program. This section will present a short program that *blurs* an eight-bit gray scale image. Then, this section will lead though through several optimization steps and show you how to get that program running over 16 times faster.

The following code assumes that you provide it with a file containing a 251x256 gray scale photographic image. The data structure for this file is as follows:

Image: array [0..250, 0..255] of byte;

Each byte contains a value in the range 0..255 with zero denoting black, 255 representing white, and the other values representing even shades of gray between these two extremes.

The blurring algorithm averages a pixel⁵ with its eight closest neighbors. A single blur operation applies this average to all interior pixels of an image (that is, it does not apply to the pixels on the boundary of the image because they do not have the same number of neighbors as the other pixels). The following Pascal program implements the blurring algorithm and lets the user specify the amount of blurring (by looping through the algorithm the number of times the user specifies)⁶:

program PhotoFilter(input,output);

```
(* Here is the raw file data type produced by the Photoshop program *)
type
image = array [0..250] of array [0..255] of byte;
(* The variables we will use. Note that the "datain" and "dataout" *)
(* variables are pointers because Turbo Pascal will not allow us to *)
(* allocate more than 64K data in the one global data segment it *)
(* supports. *)
var
h,i,j,k,l,sum,iterations:integer;
datain, dataout: ^image;
f,g:file of image;
begin
 (* Open the files and real the input data *)
 assign(f, `roller1.raw');
assign(g, `roller2.raw');
reset(f);
rewrite(q);
new(datain);
new(dataout);
read(f,datain^);
 (* Get the number of iterations from the user *)
write(`Enter number of iterations:');
readln(iterations);
 writeln('Computing result');
 (* Copy the data from the input array to the output array. *)
 (* This is a really lame way to copy the border from the *)
 (* input array to the output array. *)
 for i := 0 to 250 do
        for j := 0 to 255 do
              dataout^ [i][j] := datain^ [i][j];
                                                                  *)
 (* Okay, here's where all the work takes place. The outside
 (* loop repeats this blurring operation the number of
                                                                  *)
                                                                  *)
 (* iterations specified by the user.
 for h := 1 to iterations do begin
         (* For each row except the first and the last, compute
                                                                  * )
        (* a new value for each element.
                                                                  * )
        for i := 1 to 249 do
```

^{5.} Pixel stands for "picture element." A pixel is an element of the Image array defined above.

^{6.} A comparable C program appears on the diskette accompanying the lab manual.

```
(* For each column except the first and the last, com-
                                                                         * )
              (* pute a new value for each element.
                                                                         *)
             for j := 1 to 254 do begin
                         (* For each element in the array, compute a new
                           blurred value by adding up the eight cells
                            around an array element along with eight times
                            the current cell's value. Then divide this by
                           sixteen to compute a weighted average of the
                           nine cells forming a square around the current
                           cell. The current cell has a 50% weighting,
                            the other eight cells around the current cel
                           provide the other 50% weighting (6.25% each). *)
                         sum := 0;
                          for k := -1 to 1 do
                            for l := -1 to 1 do
                                       sum := sum + datain^ [i+k][i+l];
                         (* Sum currently contains the sum of the nine
                                                                            *)
                         (* cells, add in seven times the current cell so *)
                         (* we get a total of eight times the current cell. *)
                        dataout^ [i][j] := (sum + datain^ [i][j]*7) div 16;
             end;
              (* Copy the output cell values back to the input cells
                                                                         *)
                                                                         *)
              (* so we can perform the blurring on this new data on
             (* the next iteration.
                                                                         *)́
             for i := 0 to 250 do
              for j := 0 to 255 do
                        datain^ [i][j] := dataout^ [i][j];
end;
writeln(`Writing result');
write(g,dataout<sup>^</sup>);
close(f);
close(g);
```

end.

The Pascal program above, compiled with Turbo Pascal v7.0, takes 45 seconds to compute 100 iterations of the blurring algorithm. A comparable program written in C and compiled with Borland C++ v4.02 takes 29 seconds to run. The same source file compiled with Microsoft C++ v8.00 runs in 21 seconds. Obviously the C compilers produce better code than Turbo Pascal. It took about three hours to get the Pascal version running and tested. The C versions took about another hour to code and test. The following two images provide a "before" and "after" example of this program's function:

Before blurring:



After blurring (10 iterations):



The following is a crude translation from Pascal directly into assembly language of the above program. It requires 36 seconds to run. Yes, the C compilers did a better job, but once you see how bad this code is, you'll wonder what it is that Turbo Pascal is doing to run so slow. It took about an hour to translate the Pascal version into this assembly code and debug it to the point it produced the same output as the Pascal version.

```
; IMGPRCS.ASM
;
; An image processing program.
;
; This program blurs an eight-bit grayscale image by averaging a pixel
; in the image with the eight pixels around it. The average is computed
; by (CurCell*8 + other 8 cells)/16, weighting the current cell by 50%.
;
; Because of the size of the image (almost 64K), the input and output
; matrices are in different segments.
;
; Version #1: Straight-forward translation from Pascal to Assembly.
```

; Performance comparisons (66 MHz 80486 DX/2 system). ; ; 36 seconds. This code-; ; Borland Pascal v7.0-45 seconds. Borland C++ v4.02-29 seconds. ; Microsoft C++ v8.00-21 seconds. ; .xlist include stdlib.a includelib stdlib.lib .list .286 dsea segment para public 'data' ; Loop control variables and other variables: h ? word ? i word j word ? k word ? 1 word ? ? word Sum iterations ? word ; File names: "roller1.raw",0 InName byte byte OutName "roller2.raw",0 dseq ends ; Here is the input data that we operate on. para public `indata' InSeq segment DataIn byte 251 dup (256 dup (?)) InSeg ends ; Here is the output array that holds the result. OutSeg segment para public 'outdata' DataOut byte 251 dup (256 dup (?)) OutSeq ends segment para public 'code' cseq cs:cseg, ds:dseg assume Main proc ax, dseg mov mov ds, ax meminit mov ax, 3d00h ;Open input file for reading. lea dx, InName int 21h jnc GoodOpen print "Could not open input file.", cr, lf, 0 byte Quit jmp GoodOpen: ;File handle. bx, ax mov dx, InSeg mov ;Where to put the data. mov ds, dx lea dx, DataIn

	mov	cx, 256*251 ah, 3Fh	;Size of data file to read.
	int cmp je	21h ax, 256*251 GoodRead	;See if we read the data.
	print byte jmp	"Did not read the Quit	e file properly",cr,lf,0
GoodRead:	mov	ax, dseg ds, ax	
	print byte getsm atoi	"Enter number of	iterations: ",0
	free mov	iterations, ax	
	print byte	"Computing Result	" cr lf 0
· Comu the in	-		
; Copy the ir	iput data to	the output buffer	•
iloop0:	mov cmp	i, 0 i, 250	
-	ja	iDone0	
jloop0:	mov cmp	j, 0 j, 255	
	ja	jDone0	
	mov	bx, i	;Compute index into both
	shl add	bx, 8 bx, j	; arrays using the formula ; i*256+j (row major).
		-	
	mov mov	cx, InSeg es, cx	;Point at input segment.
	mov		;Get DataIn[i][j].
	mov	cx, OutSeg	;Point at output segment.
	mov mov	es, cx es:DataOut[bx], a	al ;Store into DataOut[i][j]
	inc jmp	j jloop0	;Next iteration of j loop.
jDone0:	inc	i	;Next iteration of i loop.
JUSIIO	jmp	iloop0	mext relation of 1 100p.
iDone0:			
; for h := 1	to iteratic	ns-	
-			
hloop:	mov mov	h, 1 ax, h	
	cmp	ax, iterations	
	ja	hloopDone	
; for i := 1	to 249 -		
	mov	i, 1	
iloop:	cmp ja	i, 249 iloopDone	
	-		
; for j := 1	to 254 - mov	j, 1	
jloop:	cmp	j, 254	
	ja	jloopDone	
; sum := 0;			
	l to 1 do fo	or l := -1 to 1 do	
	mov	ax, InSeg	;Gain access to InSeg.
		,	

	mov	es, ax	
	mov	sum, 0 k, -1	
kloop:	mov cmp	k, 1	
HICOP	ja	kloopDone	
	mov	l, -1	
lloop:	cmp	1, 1	
	ja	lloopDone	
; sum := sum	+ datain [i	.+k][j+l]	
	mov	bx, i	
	add	bx, k	
	shl	bx, 8	;Multiply by 256.
	add	bx, j	
	add	bx, l	
	mov	al, es:DataIn[bx]	
	mov	ah, 0	
	add	Sum, ax	
		_	
	inc	1	
	jmp	lloop	
lloopDone:	inc	k	
TTOOPDONG	jmp	kloop	
		-	
; dataout [i]	[j] := (sum	n + datain[i][j]*7)	div 16;
kloopDone:	mov	bx, i	
	shl	bx, 8	;*256
	add	bx, j	
	mov	al, es:DataIn[bx]	
	mov	ah, 0	
	imul	ax, 7	
	add shr	ax, sum ax, 4	;div 16
	BIII	ax, i	
	mov	bx, OutSeg	
	mov	es, bx	
	mov	bx, i	
	shl	bx, 1 bx, 8	
	add	bx, j	
	mov	es:DataOut[bx], a	1
	inc	j jloop	
	jmp	JIOOD	
jloopDone:	inc	i	
	jmp	iloop	
iloopDone: ; Copy the ou	itput data t	to the input buffer	·.
	mov	i, 0	
iloop1:	cmp	i, 250	
	ja	iDonel	
17 1.	mov	j, 0	
jloop1:	cmp ja	j, 255 jDonel	
	مار		
	mov	bx, i	;Compute index into both
	shl	bx, 8	; arrays using the formula
	add	bx, j	; i*256+j (row major).
	mov	cx, OutSeg	;Point at input segment.
	mov	es, cx	
	mov];Get DataIn[i][j].
		THE CA	
	mov	cx, InSeg	;Point at output segment.

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	mov mov	es, cx es:DataIn[bx], al	;Store into DataOut[i][j]
	inc jmp	j jloopl	;Next iteration of j loop.
jDone1:	inc jmp	i iloopl	;Next iteration of i loop.
iDone1:	inc jmp	h hloop	
hloopDone:	print byte	"Writing result",	cr,lf,0
; Okay, write	the data t	o the output file:	
	mov mov lea int jnc print	ah, 3ch cx, 0 dx, OutName 21h GoodCreate	;Create output file. ;Normal file attributes.
	byte jmp	"Could not create Quit	<pre>output file.",cr,lf,0</pre>
GoodCreate:	mov push mov lea mov mov int pop cmp je print byte jmp	bx dx, OutSeg ds, dx dx, DataOut cx, 256*251 ah, 40h 21h bx ax, 256*251 GoodWrite	<pre>le handle. ;Where the data can be found. ;Size of data file to write. ;Write operation. ;Retrieve handle for close. ;See if we wrote the data. e file properly",cr,lf,0</pre>
GoodWrite:	mov int	ah, 3eh 21h	;Close operation.
Quit: Main	ExitPgm endp		;DOS macro to quit program.
cseg	ends		
sseg stk sseg	segment byte ends	para stack `stack 1024 dup ("stack	
zzzzzzseg LastBytes zzzzzseg	segment byte ends end	para public `zzzz 16 dup (?) Main	zz'

This assembly code is a very straight-forward, line by line translation of the previous Pascal code. Even beginning programmers (who've read and understand Chapters Eight and Nine) should easily be able to improve the performance of this code.

While we could run a profiler on this program to determine where the "hot spots" are in this code, a little analysis, particularly of the Pascal version, should make it obvious that there are a lot of nested loops in this code. As Chapter Ten points out, when optimizing code you should always start with the innermost loops. The major change between the code above and the following assembly language version is that we've unrolled the innermost loops and we've replaced the array index computations with some constant

computations. These minor changes speed up the execution by a factor of six! The assembly version now runs in six seconds rather than 36. A Microsoft C++ version of the same program with comparable optimizations runs in eight seconds. It required nearly four hours to develop, test, and debug this code. It required an additional hour to apply these same modifications to the C version⁷.

```
; IMGPRCS2.ASM
; An image processing program (First optimization pass).
; This program blurs an eight-bit grayscale image by averaging a pixel
; in the image with the eight pixels around it. The average is computed
; by (CurCell*8 + other 8 cells)/16, weighting the current cell by 50%.
; Because of the size of the image (almost 64K), the input and output
; matrices are in different segments.
; Version #1: Straight-forward translation from Pascal to Assembly.
; Version #2: Three major optimizations. (1) used moved instruction rather
          than a loop to copy data from DataOut back to DataIn.
          (2) Used repeat. until forms for all loops. (3) unrolled
;
;
          the innermost two loops (which is responsible for most of
;
         the performance improvement).
:
:
        Performance comparisons (66 MHz 80486 DX/2 system).
;
;
        This code-
                                       6 seconds.
;
        Original ASM code-
Borland Pascal v7.0-
                                      36 seconds.
;
                                    45 seconds.
:
        Borland C++ v4.02-
                                        29 seconds.
;
        Microsoft C++ v8.00-
                                       21 seconds.
;
         « Lots of omitted code goes here, see the previous version»
;
              print
              byte
                          "Computing Result", cr, lf, 0
; for h := 1 to iterations-
                         h, 1
              mov
hloop:
; Copy the input data to the output buffer.
; Optimization step #1: Replace with movs instruction.
              push
                         ds
                         ax, OutSeg
              mov
              mov
                         ds, ax
                         ax, InSeg
              mov
              mov
                        es, ax
              lea
                         si, DataOut
                         di, DataIn
              lea
              mov
                         cx, (251*256)/4
        rep
              movsd
                         ds
              aoa
; Optimization Step #1: Convert loops to repeat..until form.
; for i := 1 to 249 -
              mov
                        i, 1
iloop:
; for j := 1 to 254 -
```

^{7.} This does not imply that coding this improved algorithm in C was easier. Most of the time on the assembly version was spent trying out several different modifications to see if they actually improved performance. Many modifications did not, so they were removed from the code. The development of the C version benefited from the past work on the assembly version. It was a straight-forward conversion from assembly to C.

mov i, 1 iloop: ; Optimization. Unroll the innermost two loops: bh, byte ptr i; i is always less than 256. mov bl, byte ptr j;Computes i*256+j! mov push ds ax, InSeq mov ;Gain access to InSeq. mov ds, ax mov cx, 0 ;Compute sum here. ah, ch mov mov cl, ds:DataIn[bx-257];DataIn[i-1][j-1] al, ds:DataIn[bx-256];DataIn[i-1][j] mov add cx, ax al, ds:DataIn[bx-255];DataIn[i-1][i+1] mov cx, ax add mov al, ds:DataIn[bx-1];DataIn[i][j-1] add cx, ax al, ds:DataIn[bx+1];DataIn[i][j+1] mov add cx, ax al, ds:DataIn[bx+255];DataIn[i+1][i-1] mov add cx, ax al, ds:DataIn[bx+256];DataIn[i+1][j] mov add cx, ax al, ds:DataIn[bx+257];DataIn[i+1][j+1] mov add cx, ax al, ds:DataIn[bx];DataIn[i][i] mov shl ax, 3 ;DataIn[i][j]*8 cx, ax add shr cx, 4 ;Divide by 16 mov ax, OutSeq ds. ax mov mov ds:DataOut[bx], cl ds pop inc j cmp i. 254 jbe jloop inc i i, 249 cmp jbe iloop inc h ax, h mov cmp ax, Iterations jnbe Done jmp hloop Done: print "Writing result", cr, lf, 0 byte ; «More omitted code goes here, see the previous version»

The second version above still uses memory variables for most computations. The optimizations applied to the original code were mainly language-independent optimizations. The next step was to begin applying some assembly language specific optimizations to the code. The first optimization we need to do is to move as many variables as possible into the 80x86's register set. The following code provides this optimization. Although this only improves the running time by 2 seconds, that is a 33% improvement (six seconds down to four)!

```
; IMGPRCS.ASM
```

; An image processing program (Second optimization pass).

; This program blurs an eight-bit grayscale image by averaging a pixel ; in the image with the eight pixels around it. The average is computed ; by (CurCell*8 + other 8 cells)/16, weighting the current cell by 50%. ; Because of the size of the image (almost 64K), the input and output ; matrices are in different segments. ; Version #1: Straight-forward translation from Pascal to Assembly. ; Version #2: Three major optimizations. (1) used moved instruction rather than a loop to copy data from DataOut back to DataIn. (2) Used repeat..until forms for all loops. (3) unrolled the innermost two loops (which is responsible for most of the performance improvement). ; Version #3: Used registers for all variables. Set up segment registers ; once and for all through the execution of the main loop so the code didn't have to reload ds each time through. Computed : index into each row only once (outside the j loop). ; ; ; ; Performance comparisons (66 MHz 80486 DX/2 system). ; This code-4 seconds. ; 6 seconds. 1st optimization pass-; Original ASM code-36 seconds. : ; «Lots of delete code goes here» print "Computing Result", cr, lf, 0 byte ; Copy the input data to the output buffer. hloop: ax, InSeq mov mov es, ax ax, OutSeg mov mov ds, ax si, DataOut lea lea di, DataIn mov cx, (251*256)/4 movsd rep ds:InSeq, es:OutSeq assume ax, InSeg mov mov ds, ax ax, OutSeg mov mov es, ax cl, 249 mov iloop: bh, cl ;i*256 mov bl, 1 ;Start at j=1. mov ch, 254 ;# of times through loop. mov jloop: dx, 0 ;Compute sum here. mov ah, dh mov mov dl, DataIn[bx-257] ;DataIn[i-1][j-1] al, DataIn[bx-256] ;DataIn[i-1][j] mov add dx, ax al, DataIn[bx-255] mov ;DataIn[i-1][j+1] add dx, ax al, DataIn[bx-1] ;DataIn[i][j-1] mov add dx, ax al, DataIn[bx+1] mov ;DataIn[i][j+1] dx, ax add mov al, DataIn[bx+255] ;DataIn[i+1][j-1] add dx, ax mov al, DataIn[bx+256] ;DataIn[i+1][j] add dx, ax al, DataIn[bx+257] ;DataIn[i+1][j+1] mov

:

	add	dx, ax	
	mov shl add shr mov	al, DataIn[bx] ax, 3 dx, ax dx, 4 DataOut[bx], dl	;DataIn[i][j] ;DataIn[i][j]*8 ;Divide by 16
	inc dec jne	bx ch jloop	
	dec jne	cl iloop	
	dec jne	bp hloop	
Done:	print byte	"Writing result",cr,lf,0	

«More deleted code goes here, see the original version»

Note that on each iteration, the code above still copies the output data back to the input data. That's almost six and a half megabytes of data movement for 100 iterations! The following version of the blurring program unrolls the hloop twice. The first occurrence copies the data from DataIn to DataOut while computing the blur, the second instance copies the data from DataOut back to DataIn while blurring the image. By using these two code sequences, the program save copying the data from one point to another. This version also maintains some common computations between two adjacent cells to save a few instructions in the innermost loop. This version arranges instructions in the innermost loop to help avoid data hazards on 80486 and later processors. The end result is almost 40% faster than the previous version (down to 2.5 seconds from four seconds).

```
; IMGPRCS.ASM
; An image processing program (Third optimization pass).
; This program blurs an eight-bit grayscale image by averaging a pixel
; in the image with the eight pixels around it. The average is computed
; by (CurCell*8 + other 8 cells)/16, weighting the current cell by 50%.
; Because of the size of the image (almost 64K), the input and output
; matrices are in different segments.
; Version #1: Straight-forward translation from Pascal to Assembly.
 Version #2: Three major optimizations. (1) used moved instruction rather
;
         than a loop to copy data from DataOut back to DataIn.
         (2) Used repeat..until forms for all loops. (3) unrolled
         the innermost two loops (which is responsible for most of
         the performance improvement).
 Version #3: Used registers for all variables. Set up segment registers
;
         once and for all through the execution of the main loop so
         the code didn't have to reload ds each time through. Computed
:
         index into each row only once (outside the j loop).
 Version #4: Eliminated copying data from DataOut to DataIn on each pass.
;
         Removed hazards. Maintained common subexpressions. Did some
;
         more loop unrolling.
;
;
        Performance comparisons (66 MHz 80486 DX/2 system, 100 iterations).
;
        This code-
                                        2.5 seconds.
;
        2nd optimization pass-
                                        4 seconds.
;
;
        1st optimization pass-
                                        6 seconds.
        Original ASM code-
                                        36 seconds.
;
        «Lots of deleted code here, see the original version»
;
```

print "Computing Result", cr, lf, 0 bvte assume ds:InSeq, es:OutSeq ax, InSeq mov mov ds, ax ax, OutSeg mov mov es, ax ; Copy the data once so we get the edges in both arrays. cx, (251*256)/4 mov si, DataIn di, DataOut lea lea movsd rep ; "hloop" repeats once for each iteration. hloop: ax, InSeq mov ds, ax mov mov ax, OutSeg mov es, ax ; "iloop" processes the rows in the matrices. cl, 249 mov iloop: mov bh, cl ;i*256 bl, 1 ;Start at i=1. mov mov ch, 254/2 ;# of times through loop. si, bx mov mov dh, 0 ;Compute sum here. mov bh, 0 ah, 0 mov ; "jloop" processes the individual elements of the array. ; This loop has been unrolled once to allow the two portions to share ; some common computations. jloop: ; The sum of DataIn [i-1][j] + DataIn[i-1][j+1] + DataIn[i+1][j] + ; DataIn [i+1][j+1] will be used in the second half of this computation. ; So save its value in a register (di) until we need it again. mov dl, DataIn[si-256] ;[i-1,j] ;[i-1,j+1] al, DataIn[si-255] mov bl, DataIn[si+257] mov ;[i+1,j+1] add dx, ax al, DataIn[si+256] mov ;[I+1,j] add dx, bx bl, DataIn[si+1] ;[i,j+1] mov dx, ax add al, DataIn[si+255] ;[i+1,j-1] mov di, dx ;Save partial result. mov add dx, bx bl, DataIn[si-1] mov ;[i,j-1] add dx, ax al, DataIn[si] ;[i,j] mov add dx, bx bl, DataIn[si-257] ;[i-1,j-1] mov ax, 3 ;DataIn[i,j] * 8. shl add dx, bx dx, ax add shr ax, 3 ;Restore DataIn[i,j]. dx, 4 ;Divide by 16. shr

add

mov

di, ax

DataOut[si], dl

; Okay, process the next cell over. Note that we've got a partial sum ; sitting in DI already. Don't forget, we haven't bumped SI at this point, ; so the offsets are off by one. (This is the second half of the unrolled ; loop.)

mov mov mov	dx, di bl, DataIn[si-254] al, DataIn[si+2]	;Partial sum. ;[i-1,j+1] ;[i,j+1]
add	dx, bx	
mov	bl, DataIn[si+258]	;[i+1,j+1];
add	dx, ax	
mov	al, DataIn[si+1]	;[i,j]
add	dx, bx	
shl	ax, 3	;DataIn[i][j]*8
add	si, 2	;Bump array index.
add	dx, ax	
mov	ah, O	;Clear for next iter.
shr	dx, 4	;Divide by 16
dec	ch	
mov	DataOut[si-1], dl	
jne	jloop	
dec	cl	
jne	iloop	
-	-	
dec	dq	
je	Done	
-		

; Special case so we don't have to move the data between the two arrays. ; This is an unrolled version of the hloop that swaps the input and output

; arrays so we don't have to move data around in memory.

mov mov mov mov assume	ax, OutSeg ds, ax ax, InSeg es, ax es:InSeq, ds:OutSeg
assume	es:InSeg, ds:OutSeg

hloop2:

iloop2:	mov mov mov mov mov mov mov	cl, 249 bh, cl bl, 1 ch, 254/2 si, bx dh, 0 bh, 0 ah, 0
jloop2:	mov mov add mov add mov add mov	<pre>dl, DataOut[si-256] al, DataOut[si-255] bl, DataOut[si+257] dx, ax al, DataOut[si+256] dx, bx bl, DataOut[si+1] dx, ax al, DataOut[si+255] di, dx</pre>
	mov add mov add mov add mov shl add add shr shr mov	<pre>dx, bx bl, DataOut[si-1] dx, ax al, DataOut[si] dx, bx bl, DataOut[si-257] ax, 3 dx, bx dx, ax ax, 3 dx, 4 DataIn[si], dl</pre>

mov	<pre>dx, di bl, DataOut[si-254] dx, ax al, DataOut[si+2] dx, bx bl, DataOut[si+258] dx, ax al, DataOut[si+1] dx, bx ax, 3 si, 2 dx, ax ah, 0 dx, 4 ch DataIn[si-1], dl</pre>
jne	jloop2
dec jne	cl iloop2
dec je jmp	bp Done2 hloop

; Kludge to guarantee that the data always resides in the output segment.

Done2:

		mov	ax,	InSeg
		mov	ds,	ax
		mov	ax,	OutSeg
		mov	es,	ax
		mov	cx,	(251*256)/4
		lea	si,	DataIn
		lea	di,	DataOut
	rep	movsd		
Done:		print		
		byte	"Wr	iting result",cr,lf,0
;	«Lots	of deleted	l cod	le here, see the original program»

This code provides a good example of the kind of optimization that scares a lot of people. There is a lot of cycle counting, instruction scheduling, and other crazy stuff that makes program very difficult to read and understand. This is the kind of optimization for which assembly language programmers are famous; the stuff that spawned the phrase "never optimize early." You should never try this type of optimization until you feel you've exhausted all other possibilities. Once you write your code in this fashion, it is going to be very difficult to make further changes to it. By the way, the above code took about 15 hours to develop and debug (debugging took the most time). That works out to a 0.1 second improvement (for 100 iterations) for each hour of work. Although this code certainly isn't optimal yet, it is difficult to justify more time attempting to improve this code by mechanical means (e.g., moving instructions around, etc.) because the performance gains would be so little.

In the four steps above, we've reduced the running time of the assembly code from 36 seconds down to 2.5 seconds. Quite an impressive feat. However, you shouldn't get the idea that this was easy or even that there were only four steps involved. During the actual development of this example, there were many attempts that did not improve performance (in fact, some modifications wound up reducing performance) and others did not improve performance enough to justify their inclusion. Just to demonstrate this last point, the following code included a major change in the way the program organized data. The main loop operates on 16 bit objects in memory rather than eight bit objects. On some machines with large external caches (256K or better) this algorithm provides a slight improvement in performance (2.4 seconds, down from 2.5). However, on other machines it runs slower. Therefore, this code was not chosen as the final implementation:

; IMGPRCS.ASM ; An image processing program (Fourth optimization pass). ; This program blurs an eight-bit grayscale image by averaging a pixel in the image with the eight pixels around it. The average is computed ; by (CurCell*8 + other 8 cells)/16, weighting the current cell by 50%. ; Because of the size of the image (almost 64K), the input and output ; matrices are in different segments. ; Version #1: Straight-forward translation from Pascal to Assembly. ; Version #2: Three major optimizations. (1) used moved instruction rather than a loop to copy data from DataOut back to DataIn. (2) Used repeat..until forms for all loops. (3) unrolled the innermost two loops (which is responsible for most of the performance improvement). Version #3: Used registers for all variables. Set up segment registers ; once and for all through the execution of the main loop so the code didn't have to reload ds each time through. Computed index into each row only once (outside the j loop). Version #4: Eliminated copying data from DataOut to DataIn on each pass. : Removed hazards. Maintained common subexpressions. Did some more loop unrolling. Version #5: Converted data arrays to words rather than bytes and operated ; on 16-bit values. Yielded minimal speedup. Performance comparisons (66 MHz 80486 DX/2 system). : This code-2.4 seconds. : ; 3rd optimization pass-2.5 seconds. 2nd optimization pass-; 4 seconds. 1st optimization pass-6 seconds. ; Original ASM code-36 seconds. ; .xlist include stdlib.a includelib stdlib.lib .list .386 segment:use16 option dseg segment para public 'data' ImgData byte 251 dup (256 dup (?)) "roller1.raw",0 InName byte OutName "roller2.raw",0 byte Iterations 0 word dseq ends ; This code makes the naughty assumption that the following ; segments are loaded contiguously in memory! Also, because these ; sequents are paragraph aligned, this code assumes that these segments ; will contain a full 65,536 bytes. You cannot declare a segment with ; exactly 65,536 bytes in MASM. However, the paragraph alignment option ; ensures that the extra byte of padding is added to the end of each ; segment. para public 'ds1' DataSeg1 segment

Datala DataSegl	byte ends	65535 dup (?)
DataSeg2 Data1b DataSeg2	segment byte ends	para public `ds2' 65535 dup (?)

DataSeg3 Data2a DataSeg3	segment byte ends	para public `ds3′ 65535 dup (?)	
DataSeg4 Data2b DataSeg4	segment byte ends	para public `ds4' 65535 dup (?)	,
cseg	segment assume	para public `code cs:cseg, ds:dseg	2'
Main	proc mov mov meminit	ax, dseg ds, ax	
	mov lea int jnc	ax, 3d00h ;0j dx, InName 21h GoodOpen	pen input file for reading.
	print byte jmp	"Could not open i Quit	input file.",cr,lf,0
GoodOpen:	mov lea	bx, ax	;File handle.
	mov	dx, ImgData cx, 256*251 ah, 3Fh	;Size of data file to read.
	int cmp je	21h ax, 256*251 GoodRead	;See if we read the data.
	print byte jmp	"Did not read the Quit	e file properly",cr,lf,0
GoodRead:	print byte getsm atoi	"Enter number of	iterations: ",0
	free mov	Iterations, ax	
	cmp jle	ax, 0 Quit	
	printf byte dword	"Computing Result Iterations	for %d iterations",cr,lf,0
	oop handles	the first 32,768	bits to sixteen bits. bytes, the second loop
	mov	ax, DataSegl es, ax	
	mov mov mov	es, ax ax, DataSeg3 fs, ax	
	mov mov lea xor	ah, 0 cx, 32768 si, ImgData di, di	;Output data is at ofs zero.
: acolvacy	lodsb		;Read a byte

mov stosw dec

jne

mov

lodsb

fs:[di], ax

CopyLoop di, DataSeg2

сx

CopyLoop:

;Output data is at ofs zero. ;Read a byte iStore a word in DataSeg3
;Store a word in DataSeg1

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CopyLoop1: ; hloop comple ; to Data2a/Da hloop:		es, di di, DataSeg4 fs, di cx, (251*256) - 3 di, di fs:[di], ax cx CopyLoop1 eration on the data ax, DataSeg1 ds, ax ax, DataSeg3 es, ax	2768 ;Read a byte ;Store a word in DataSeg4 ;Store a word in DataSeg2 a moving it from Datala/Datalb
; Process the	first 127 :	rows (65,024 bytes) of the array):
iloop0: jloop0:	mov lea mov mov shl add add add add add add add add add ad	cl, 127 si, Datala+202h ch, 254/2 dx, [si] bx, [si-200h] ax, dx dx, 3 bx, [si-1feh] bp, [si+2] bx, [si+200h] dx, bp bx, [si+202h] dx, [si-202h] di, [si-1fch] dx, [si-202h] di, [si-1fch] dx, [si-21] di, [si+4] dx, [si+1feh] di, [si+204h] bp, 3 dx, bx bp, ax dx, 4 bp, bx es:[si], dx bp, di si, 4 bp, 4 ch es:[si-2], bp jloop0	<pre>;Start at [1,1] ;# of times through loop. ;[i,j] ;[i-1,j] ;[i,j] * 8 ;[i-1,j+1] ;[i,j+1] ;[i+1,j+1] ;[i+1,j-1] ;[i-1,j+2] ;[i,j-1] ;[i,j+2] ;[i,j+1] * 8 ;Divide by 16. ;Store [i,j] entry. ;Affects next store operation! ;Divide by 16. ;Store [i,j+1] entry.</pre>
	add	si, 4	;Skip to start of next row.
		cl iloop0 ows of the array). t. Note that the se	This requires that we switch from egments overlap.
	mov	ax, DataSeg2	
	sub mov mov sub mov	ax, 40h ds, ax ax, DataSeg4 ax, 40h es, ax	;Back up to last 2 rows in DS2 ;Back up to last 2 rows in DS4

;Remaining rows to process.

;Continue with next row. ;# of times through loop. ;[i,j] ;[i-1,j]

;[i,j] * 8

mov

mov

mov mov mov

mov

shl

iloop1: jloop1:

cl, 251-127-1 si, 202h ch, 254/2 dx, [si] bx, [si-200h]

ax, dx dx, 3

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add mov add add	bx, [si-lfeh] bp, [si+2] bx, [si+200h] dx, bp	;[i-1,j+1] ;[i,j+1] ;[i+1,j]
add add mov add add add add shl	bx, [si+202h] dx, [si-202h] di, [si-1fch] dx, [si-2]	
add add shr	dx, bx bp, ax dx, 4	;Divide by 16
add mov add	bp, bx es:[si], dx bp, di	;Store [i,j] entry.
add shr dec	si, 4 bp, 4 ch	;Affects next store operation!
mov jne	es:[si-2], bp jloop1	;Store [i,j+1] entry.
add	si, 4	;Skip to start of next row.
dec jne	cl iloopl	
mov mov assume	ax, dseg ds, ax ds:dseg	
dec je	Iterations Done0	

; Unroll the iterations loop so we can move the data from DataSeg2/4 back ; to DataSeg1/3 without wasting extra time. Other than the direction of the ; data movement, this code is virtually identical to the above.

	mov	ax, DataSeg3
	mov	ds, ax
	mov	ax, DataSegl
	mov	es, ax
	mov	cl, 127
	lea	si, Datala+202h
iloop2:	mov	ch, 254/2
jloop2:	mov	dx, [si]
	mov	bx, [si-200h]
	mov	ax, dx
	shl	dx, 3
	add	bx, [si-lfeh]
	mov	bp, [si+2]
	add	bx, [si+200h]
	add	dx, bp
	add	bx, [si+202h]
	add	dx, [si-202h]
	mov	di, [si-lfch]
	add	dx, [si-2]
	add	di, [si+4]
	add	dx, [si+lfeh]
	add	di, [si+204h]
	shl	bp, 3
	add	dx, bx
	add	bp, ax
	shr	dx, 4
	add	bp, bx
	mov	es:[si], dx
	add	bp, di
	add	si, 4
	shr	bp, 4
	dec	ch
	mov	es:[si-2], bp
		· •

	jne	jloop2
	add	si, 4
	dec	cl
	jne	iloop2
	mov	ax, DataSeg4
	sub	ax, 40h
	mov	ds, ax
	mov	ax, DataSeg2
	sub	ax, 40h
	mov	es, ax
	mov	cl, 251-127-1
iloop3:	mov	si, 202h ch 254/2
jloop3:	mov mov	ch, 254/2 dx, [si]
5 - 1 -	mov	bx, [si-200h]
	mov	ax, dx
	shl	dx, 3
	add mov	bx, [si-lfeh] bp, [si+2]
	add	bp, [si+2] bx, [si+200h]
	add	dx, bp
	add	bx, [si+202h]
	add	dx, [si-202h]
	mov add	di, [si-lfch] dx, [si-2]
	add	di, [si+4]
	add	dx, [si+lfeh]
	add	di, [si+204h]
	shl	bp, 3
	add add	dx, bx bp, ax
	shr	dx, 4
	add	bp, bx
	mov	es:[si], dx
	add	bp, di
	add shr	si, 4 bp, 4
	dec	ch
	mov	es:[si-2], bp
	jne	jloop3
	add	si, 4
	dec	cl
	jne	iloop3
	mov	ax, dseg
	mov	ds, ax
	assume	ds:dseg
	dec	Iterations
	je jmp	Done2 hloop
	Juip	шоор
Done2:	mov	ax, DataSeg1
	mov jmp	bx, DataSeg2 Finish
	Juip	1111011
Done0:	mov	ax, DataSeg3
Finish:	mov mov	bx, DataSeg4 ds, ax
	print	us, ux
	byte	"Writing result", cr, lf, 0
; Convert data	a back to by	yte form and write to the output file:
	mov	ax, dseg
	mov	es, ax

CopyLoop3:	mov lea xor lodsw stosb dec jne	cx, 32768 di, ImgData si, si cx CopyLoop3	;Output data is at offset zero. ;Read a word from final array. ;Write a byte to output array.
CopyLoop4:	mov mov xor lodsw stosb dec jne	ds, bx cx, (251*256) - 3 si, si cx CopyLoop4	32768 ;Read final data word. ;Write data byte to output array.
; Okay, write	e the data t mov mov mov lea int jnc print	ah, 3ch cx, 0 dx, dseg ds, dx dx, OutName 21h GoodCreate	;Create output file. ;Normal file attributes.
GoodCreate:	byte jmp mov push mov mov lea	"Could not create Quit bx, ax bx dx, dseg ds, dx dx, ImgData	e output file.",cr,lf,0 ;File handle. ;Where the data can be found.
	mov mov int pop cmp je print	cx, 256*251 ah, 40h 21h bx ax, 256*251 GoodWrite	;Size of data file to write. ;Write operation. ;Retrieve handle for close. ;See if we wrote the data.
GoodWrite:	byte jmp mov int	"Did not write th Quit ah, 3eh 21h	ne file properly",cr,lf,0 ;Close operation.
Quit: Main	ExitPgm endp		;DOS macro to quit program.
cseg sseg stk sseg	ends segment byte ends	para stack `stack 1024 dup ("stack	
zzzzzzseg LastBytes zzzzzseg	segment byte ends end	para public `zzzz 16 dup (?) Main	222'

Of course, the absolute best way to improve the performance of any piece of code is with a better algorithm. All of the above assembly language versions were limited by a single requirement – they all must produce the same output file as the original Pascal program. Often, programmers lose sight of what it is that they are trying to accomplish and get so caught up in the computations they are performing that they fail to see other possibilities. The optimization example above is a perfect example. The assembly code faithfully preserves the semantics of the original Pascal program; it computes the weighted average

of all interior pixels as the sum of the eight neighbors around a pixel plus eight times the current pixel's value, with the entire sum divided by 16. Now this is a *good* blurring function, but it is not the *only* blurring function. A Photoshop (or other image processing program) user doesn't care about algorithms or such. When that user selects "blur image" they want it to go out of focus. Exactly how much out of focus is generally immaterial. In fact, the less the better because the user can always run the blur algorithm again (or specify some number of iterations). The following assembly language program shows how to get better performance by modifying the blurring algorithm to reduce the number of instructions it needs to execute in the innermost loops. It computes blurring by averaging a pixel with the four neighbors above, below, to the left, and to the right of the current pixel. This modification yields a program that runs 100 iterations in 2.2 seconds, a 12% improvement over the previous version:

; IMGPRCS.ASM ; An image processing program (Fifth optimization pass). ; This program blurs an eight-bit grayscale image by averaging a pixel ; in the image with the eight pixels around it. The average is computed ; by (CurCell*8 + other 8 cells)/16, weighting the current cell by 50%. ; Because of the size of the image (almost 64K), the input and output matrices are in different segments. : : Version #1: Straight-forward translation from Pascal to Assembly. Version #2: Three major optimizations. (1) used moved instruction rather ; than a loop to copy data from DataOut back to DataIn. (2) Used repeat..until forms for all loops. (3) unrolled the innermost two loops (which is responsible for most of the performance improvement). • Version #3: Used registers for all variables. Set up segment registers ; once and for all through the execution of the main loop so the code didn't have to reload ds each time through. Computed index into each row only once (outside the j loop). ; Version #4: Eliminated copying data from DataOut to DataIn on each pass. Removed hazards. Maintained common subexpressions. Did some : more loop unrolling. : Version #6: Changed the blurring algorithm to use fewer computations. ; This version does *NOT* produce the same data as the other : ; programs. ; ; Performance comparisons (66 MHz 80486 DX/2 system, 100 iterations). ; ; This code-2.2 seconds. ; This code-3rd optmization pass-2nd optimization pass-1st optimization pass-Original ASM code-2.5 seconds. ; 4 seconds. ; 6 seconds. ; ; Original ASM code-36 seconds. «Lots of deleted code here, see the original program» ; print "Computing Result", cr, lf, 0 byte ds:InSeg, es:OutSeg assume ax, InSeg mov mov ds, ax ax, OutSeg mov mov es, ax ; Copy the data once so we get the edges in both arrays.

```
mov cx, (251*256)/4
lea si, DataIn
```

lea di, DataOut movsd rep ; "hloop" repeats once for each iteration. hloop: ax, InSeg mov mov ds, ax ax, OutSeg mov mov es, ax ; "iloop" processes the rows in the matrices. cl, 249 mov iloop: bh, cl ;i*256 mov bl, 1 ;Start at j=1. mov ch, 254/2 mov ;# of times through loop. si, bx mov dh, 0 mov ;Compute sum here. mov bh, 0 mov ah, 0 ; "jloop" processes the individual elements of the array. ; This loop has been unrolled once to allow the two portions to share ; some common computations. jloop:

; The sum of DataIn [i-1][j] + DataIn[i-1][j+1] + DataIn[i+1][j] +
; DataIn [i+1][j+1] will be used in the second half of this computation.
; So save its value in a register (di) until we need it again.

mov mov shl mov	<pre>dl, DataIn[si] al, DataIn[si-256] dx, 2 bl, DataIn[si-1] dr. org</pre>	;[i,j] ;[I-1,j] ;[i,j]*4 ;[i,j-1]
add mov	dx, ax al, DataIn[si+1]	;[i,j+1]
add	dx, bx	-
mov	bl, DataIn[si+256]	;[i+1,j]
add	dx, ax	
shl	ax, 2	;[i,j+1]*4
add	dx, bx	•[+ 1 + 1]
mov shr	bl, DataIn[si-255] dx. 3	;[i-1,j+1] ;Divide by 8.
add	ax, bx	DIVIGE by 8.
mov	DataOut[si], dl	
mov	bl, DataIn[si+2]	;[i,j+2]
mov	dl, DataIn[si+257]	;[i+1,j+1]
add	ax, bx	
mov	bl, DataIn[si]	;[i,j]
add	ax, dx	
add	ax, bx	
shr	ax, 3	
dec	ch	
mov	DataOut[si+1], al jloop	
jne	JIOOD	
dec	cl	
jne	iloop	
dec	p	
je	Done	

; Special case so we don't have to move the data between the two arrays.

; This is an unrolled version of the hloop that swaps the input and output ; arrays so we don't have to move data around in memory.

ar r ay b	50	WV C	aon	C	navc	20	IIIO V C	aaca	arouna	T TT	men

mov	ax,	OutSeg
mov	ds,	ax
mov	ax,	InSeg
mov	es,	ax

	2551100	es:InSeg, ds:OutSeg
	assume	estinseg, ustoutseg
hloop2:		
	mov	cl, 249
iloop2:	mov	bh, cl
	mov	bl, 1
	mov	ch, 254/2
	mov	si, bx
	mov	dh, 0
	mov	bh, 0
	mov	ah, 0
jloop2:		
	mov	dl, DataOut[si-256]
	mov	al, DataOut[si-255]
	mov	bl, DataOut[si+257]
	add	dx, ax al, DataOut[si+256]
	mov add	dx, bx
	mov	bl, DataOut[si+1]
	add	dx, ax
	mov	al, DataOut[si+255]
	1110 V	ai, bacadac[51-255]
	mov	di, dx
	add	dx, bx
	mov	bl, DataOut[si-1]
	add	dx, ax
	mov	al, DataOut[si]
	add	dx, bx
	mov	bl, DataOut[si-257]
	shl	ax, 3
	add	dx, bx
	add	dx, ax
	shr	ax, 3
	shr	dx, 4
	mov	DataIn[si], dl
	mov	dx, di
	mov	bl, DataOut[si-254]
	add	dx, ax
	mov	al, DataOut[si+2]
	add	dx, bx
	mov	bl, DataOut[si+258]
	add	dx, ax
	mov	al, DataOut[si+1]
	add	dx, bx
	shl	ax, 3
	add	si, 2
	add	dx, ax
	mov shr	ah, 0 dx, 4
	dec	ch DataIn[si-1], dl
	mov	jloop2
	jne -	
	dec	cl
	jne	iloop2
	dec	bp
	je	Done2
	jmp	hloop

; Kludge to guarantee that the data always resides in the output segment.

Done2:

mov	ax,	InSeg
mov	ds ,	ax
mov	ax,	OutSeg
mov	es,	ax
mov	CX,	(251*256)/4
lea	si,	DataIn
lea	di,	DataOut

```
rep movsd
Done: print
byte "Writing result",cr,lf,0
; «Lots of delete code here, see the original program»
```

One very important thing to keep in mind about the codein this section is that we've optimized it for 100 iterations. While it turns out that these optimizations apply equally well to more iterations, this isn't necessarily true for fewer iterations. In particular, if we run only one iteration, any copying of data at the end of the operation will easily consume a large part of the time we save by the optimizations. Since it is very rare for a user to blur an image 100 times in a row, our optimizations may not be as good as we could make them. However, this section does provide a good example of the steps you must go through in order to optimize a given program. One hundred iterations was a good choice for this example because it was easy to measure the running time of all versions of the program. However, you must keep in mind that you should optimize your programs for the expected case, not an arbitrary case.

25.6 Summary

Computer software often runs significantly slower than the task requires. The process of increasing the speed of a program is known as *optimization*. Unfortunately, optimization is a difficult and time-consuming task, something not to be taken lightly. Many programmers often optimize their programs before they've determined that there is a need to do so, or (worse yet) they optimize a portion of a program only to find that they have to rewrite that code after they've optimized it. Others, out of ignorance, often wind up optimizing the wrong sections of their programs. Since optimization is a slow and difficult process, you want to try and make sure you only optimize your code *once*. This suggests that optimization should be your last task when writing a program.

One school of thought that completely embraces this philosophy is the *Optimize Late* group. Their arguement is that program optimization often destroys the readability and maintanability of a program. Therefore, one should only take this step when absolutely necessary and only at the end of the program development stage.

The *Optimize Early* crowd knows, from experience, that programs that are not written to be fast often need to be completely rewritten to make them fast. Therefore, they often take the attitude that optimization should take place along with normal program development. Generally, the optimize early group's view of optimization is typically far different from the optimize late group. The optimize early group claims that the extra time spent optimizing a program during development requires less time than developing a program and then optimizing it. For all the details on this *religious* battle, see

"When to Optimize, When Not to Optimize" on page 1311

After you've written a program and determine that it runs too slowly, the next step is to locate the code that runs too slow. After identifying the slow sections of your program, you can work on speeding up your programs. Locating that 10% of the code that requires 90% of the execution time is not always an easy task. The four common techniques people use are trial and error, optimize everything, program analysis, and experimental analysis (i.e., use a profiler). Finding the "hot spots" in a program is the first optimization step. To learn about these four techniques, see

"How Do You Find the Slow Code in Your Programs?" on page 1313

A convincing arguement the optimize late folks use is that machines are so fast that optimization is rarely necessary. While this arguement is often overstated, it is often true that many unoptimized programs run fast enough and do not require any optimization for satisfactory performance. On the other hand, programs that run fine by themselves may be too slow when running concurrently with other software. To see the strengths and weaknesses of this arguement, see

"Is Optimization Necessary?" on page 1314

There are three forms of optimization you can use to improve the performance of a program: choose a better algorithm, choose a better implementation of an algorithm, or "count cycles." Many people (especially the optimize late crowd) only consider this last case "optimization." This is a shame, because the last case often produces the smallest incremental improvement in performance. To understand these three forms of optimization, see

• "The Three Types of Optimization" on page 1315

Optimization is not something you can learn from a book. It takes lots of experience and practice. Unfortunately, those with little practical experience find that their efforts rarely pay off well and generally assume that optimization is not worth the trouble. The truth is, they do not have sufficient experience to write truly optimal code and their frustration prevents them from gaining such experience. The latter part of this chapter devotes itself to demonstrating what one can achieve when optimizing a program. Always keep this example in mind when you feel frustrated and are beginning to believe you cannot improve the performance of your program. For details on this example, see

• "Improving the Implementation of an Algorithm" on page 1317