Back

Print view Next

Print all

≡ Index of pages

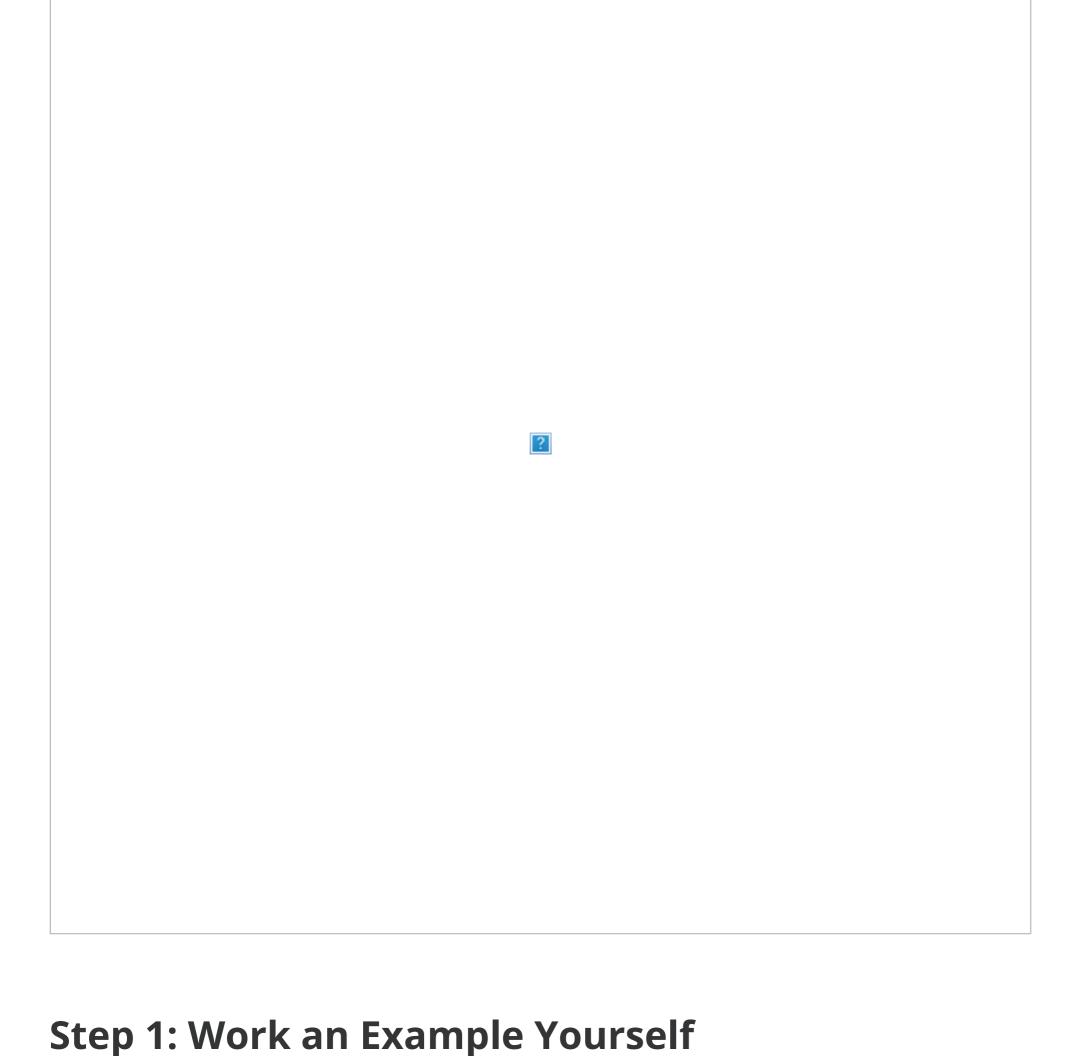


The First Four Steps:

1.0.2 Stepping Through an Algorithm.m4v

Stepping Through an Algorithm When we write an algorithm by hand, it should be precise enough that anyone can execute it and get exactly the right answer. You will see this in the next video. Note

that once we translate algorithms to code, we will see that we can execute code by hand in much the same way.

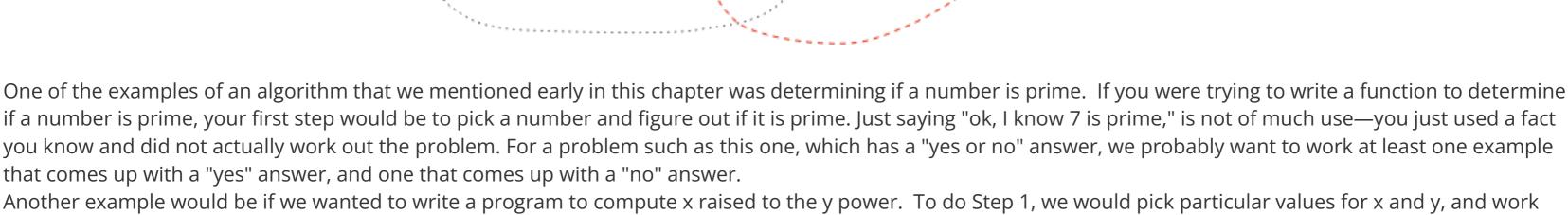


precisely you can draw a diagram of the situation if applicable), the easier the remainder of our steps will be. A good example of the sort of picture you might draw would be the diagrams drawn in many science classes (especially physics classes). The figure shows multiple copies of the box for this step layered one on top of the other, as you may need to perform this step multiple times to generalize the algorithm properly.

(2) Write down (3) Generalize instance your steps (5) Translate to exactly what your yourself from (2) you just did Code

The first step in trying to design an algorithm is to work at least one instance of the problem—picking specific values for each parameter—yourself (by hand). Often

this step will involve drawing a diagram of the problem at hand, in order to work it precisely. The more precisely you can perform this problem (including the more



Another example would be if we wanted to write a program to compute x raised to the y power. To do Step 1, we would pick particular values for x and y, and work them by hand. We might try x = 3 and y = 4, getting an answer of $3^4 = 81$. If you get stuck at this step, it typically means one of two things. The first case is that the problem is ill-specified—it is not clear what you are supposed to do. In such a situation, you must resolve how the problem should be solved before proceeding. In the case of a classroom setting, this resolution may require asking your professor

need to think harder about what the right answers should be and refine your definition of the problem. The second case where Step 1 is difficult is when you lack domain knowledge—the knowledge of the particular field or discipline the problem deals with. In our primality example, if you did not remember the definition of a prime number, that would be an example of lacking domain knowledge—the problem domain is

or TA for more details. In an industrial setting, asking your technical lead or customer may be required. If you are solving a problem of your own creation, you may

mathematics, and you are lacking in math knowledge. No amount of programming expertise nor effort ("working harder") will overcome this lack of domain knowledge. Instead, you must consult a source of domain expertise—a math textbook, website, or expert. Once you have the correct domain knowledge, you can proceed with solving your instance of the problem. Note that domain knowledge may come from domains other than math. It can come from any field, as programming is useful for processing any sort of information. Sometimes, domain knowledge may come from particular fields of computer science or engineering. For example, if you intend to write a program that determines the meaning of English text, the relevant domain field is actually a sub-field of computer science, called Natural Language Processing. Here the domain knowledge would

unlikely to contain such information. **Step 2: Write Down What You Just Did**

For this step, you must think about what you did to solve the problem, and write down the steps to solve that particular instance. Another way to think about this step,

is to write down a clear set of instructions that anyone else could follow to reproduce your answer for the particular problem instance that you just solved. If you do

multiple instances in Step 1, you will repeat Step 2 multiple times as well, once for each instance you did in Step 1. If an instruction is somewhat complex, that is all

be the specific techniques developed to write programs that deal with natural language. A source of domain knowledge on English (an English professor or textbook) is

right, as long as the instruction has a clear meaning—later, we will turn these complex steps into their own programming problems, which will get solved separately.

The difficult part of Step 2 is thinking about exactly what you did to accomplish the problem. The difficulty here is that it is very easy to mentally gloss over small details, "easy" steps, or things that you do implicitly. This difficulty is best illustrated by the peanut butter and jelly exercise we mentioned earlier. Implicit assumptions about what to do, or relying on common sense lead to imprecise or omitted steps. The computer will not fill in any steps you omit, thus you must be careful to think

Returning to our example of computing x to the y, we might write down the following steps for x = 3 and y = 4: • Multiply 3 by 3 • You get 9 • Multiply 3 by 9

• Multiply 3 by 27 o You get 81

• 81 is your answer.

o You get 27

through all the details.

- The steps are very precise—and leave nothing to guess work. Anyone who can perform basic arithmetic can follow these steps to get the right answer. Computers are very good at arithmetic, so none of these steps is even complex enough to require splitting into a sub-problem.
- **Step 3: Generalize Your Steps**

Having solved one or more problems from the class we are interested in and written down the particular steps we executed to solve them, we are ready to try to generalize those steps into an algorithm. In our Step 2 steps, we solve particular instances, but now we need to find the pattern that allows us to solve the whole

class. This generalization typically requires two activities. First, we must take particular values that we used and replace them with mathematical expressions of the

• You get 9 • Multiply x by 9

not always use 3—we are using 3 specifically because it is the value that we picked for x. We can generalize this slightly by replacing this occurrence of 3 with x: • Multiply x by 3

parameters. Looking at our Step 2 steps for computing 3^4, we would see that we are always multiplying 3 by something in each step. In the more general case, we will

o You get 27

• Multiply x by 27

• 81 is your answer.

• You get 81

- The second common way to generalize steps is to find repetition—the same step repeated over and over. Often the number of times that the pattern repeats will depend on the parameters. We must generalize how many times to do the steps, as well as what the steps are. Sometimes, we may find steps which are almost repetitive, in which case we may need to adjust our steps to make them exactly repetitive. In our 3⁴ example, our multiplication steps are almost repetitive—both multiply x by "something," but that "something" changes (3 then 9 then 27). Examining the steps in more detail, we will see that the "something" we multiply is the answer from the previous step. We can then give it a name (and an initial value) to make all of these steps the same:
 - Start with n = 3• n = Multiply x by n • n = Multiply x by n

 \bullet n = Multiply x by n

• n is your answer

• n is your answer

• n is your answer

is beyond the scope of this course.

• If y is 0 then

• Otherwise:

have written it.

• 1 is your answer

not to jump to the conclusion that it repeats x times because x = 3—that is just a coincidence in this case. In this case, it repeats y - 1 times. The reason for this is that we need to multiply four 3s together, and we already have one in n at the start, so we need y - 1 more. This would lead to the following generalized steps: • Start with n = 3• Count up from 1 to y-1 (inclusive), for each number you count, \circ n = multiply x by n

We need to make one more generalization of a specific value to a function of the parameters. We start with n = 3; however, we would not always want to start with

Now, we have the same exact step repeated three times. We can now contemplate how many times this step repeats as a function of x and/or y. We must be careful

• Start with n = x• Count up from 1 to y-1 (inclusive), for each number you count, \circ n = multiply x by n

3. In the general case, we would want to start with n = x.

- Sometimes you may find it difficult to see the pattern, making it hard to generalize the steps. When this happens, returning to Steps 1 and 2 may help. Doing more instances of the problem will provide more information for you to consider, possibly giving you insight into the patterns of your algorithm. This process is often referred to as writing 'pseudo-code', as you are working to design an algorithm programmatically with no particular target language. Nearly all programmers make use
- of this method to ensure their algorithm is correct before writing any actual code. **Step 4: Test Your Algorithm**
- After Step 3, we have an algorithm that we think is right. However, it is entirely possible that we have messed up along the way. The primary purpose of Step 4 is to ensure our steps are actually right before we proceed. To accomplish this, we test our algorithm with different values of the parameters than the ones we used to design our algorithm. We execute our algorithm by hand and compare the answer it obtains to the right answer. If they differ, then we know our algorithm is wrong.

The more test cases (values of parameters) we use, the more confident we can become that our algorithm is correct. Unfortunately, it is impossible to ensure that our

algorithm is correct by testing. The only way to be completely sure that your algorithm is correct is to formally prove its correctness (using a mathematical proof), which

One common type of mistake is mis-generalizing in Step 3. As we just discussed, one might think that the steps repeated x times because x = 3 and the steps repeated

3 times. If we had written that down in Step 3, our algorithm would only work when x = y - 1; otherwise we would count the wrong number of times and get the wrong

answer. If that were the case, we would hopefully detect the problem by testing our algorithm by hand in Step 4. When we detect such a problem, we must go back and

re-examine the generalizations we made in Step 3. Often, this is best accomplished by returning to Steps 1 and 2 for whatever test case exposed the problem. Re-doing accordingly.

Steps 1 and 2 will give you a concrete set of steps to generalize differently. You can then find where the generalization you came up with before is wrong, and revise it Another common type of mistake is that there are cases we did not consider in designing our algorithm. In fact, in our x^y example, we did not consider what happens when y = 0, and our algorithm handles this case incorrectly. If you execute the algorithm by hand with x = 2, y = 0, you should get $2^0 = 1$; however, you will get an answer of 2. Specifically, you will start with n = x = 2. We would then try to count up from 1 to 0 - 1 = -1, of which there are no numbers, so we would be done counting right away. We would then give back *n* (which is 2) as our answer.

To fix our algorithm, we would go back and revisit Steps 1 and 2 for the case that failed (x = 2, y = 0). This case is a bit tricky since we just know that the answer is 1

may seem nice, it actually makes it a little more difficult to incorporate it into our generalized steps. We might be tempted to write generalized steps like these:

without doing any work (x^0 =1 for any x). The fact that the answer requires no work makes Step 2 a little different—we just give an answer of 1. While this simplicity

 \circ Start with n = x \circ Count up from 1 to y-1 (inclusive), for each number you count, \blacksquare n = Multiply x by n o n is your answer These steps check explicitly for the case that gave us a problem (y = 0), give the right answer for that case, then perform the more general algorithm. For some

problems, there may be corner cases which require this sort of special attention. However, for this problem, we can do better. Note that if you were unable to see the

Instead, a better approach would be to realize that if we count no times, we need an answer of 1, so we should start n at 1 instead of at x. In doing so, we need to count

• Start with n = 1• Count up from 1 to y (inclusive), for each number you count, o n = Multiply x by n • n is your answer

Whenever we detect problems with our algorithm in Step 4, we typically want to return to Steps 1 and 2 to get more information to generalize from. Sometimes, we

generalization). If we see how to fix the problem, it is fine to fix it right away without redoing Steps 1 and 2, but if you are stuck, you should redo those steps until you

you should test potential corner cases right around the boundaries of these conditions. For example, if your algorithm makes a decision based on whether or not x < 3

, you might want to test with x = 2, x = 3, and x = 4. You can limit your "pencil and paper" testing somewhat, since you will do more testing on the actual code once you

may see the problem right away (e.g., if we made a trivial arithmetic mistake, or if executing the problematic test case by hand gives us insight into the correct

find a solution. Whatever approach you take to fixing your algorithm, you should re-test it with all the test cases you have already used, as well as some new ones. Determining good test cases is an important skill that improves with practice. For testing in Step 4, you will want to test with cases which at least produce a few different answers (e.g., if your algorithm has a "yes" or "no" answer, you should test with parameters which produce both "yes" and "no"). You should also test any corner cases—cases where the behavior may be different from the more general cases. Whenever you have conditional decisions (including limits on where to count),

1.0.3 Testing an Algorithm for a Numerical Sequence.m4v

1 more time (to y instead of to y - 1)—to multiply by x one more time:

Testing an Algorithm for a Numerical Sequence

better solution and were to take the above approach, it is not wrong *per se*, but it is not the best solution.

Day 1, Part 1

Next