

In [170]:

```
# This cell runs automatically. Don't edit it.  
from CSE142L.notebook import *  
from cfiddle import *  
setup_lab()
```

Double Click to edit and enter your

1. Name
2. Student ID
3. @ucsd.edu email address

# Lab 1: The Performance Equation

**Welcome to the first lab of CSE142L!**

The main goals of this lab are:

1. To get you set up with the lab environment
  - A. Github
  - B. Github Classroom
  - C. Docker
  - D. UCSD's Datahub cluster
  - E. Jupyter Notebook
2. Gain experience with the performance equation
3. Collect a list of **Interesting Questions** that you probably cannot answer now, but will be able to by the end of the course.

This lab will be completed on your own.

Check Gradescope for due date(s).



## 1 FAQ and Updates

None yet.





## 2 Using in the Correct Environment on DataHub

**Use the right Datahub environment** There is a different environment for each lab on DataHub, and you must use the correct environment when working on the corresponding lab.

To get into the right environment when you start a new lab, you should:

1. Connect to Datahub. If it takes you to the menu of environments, select the appropriate one.
2. Otherwise, click "Control Panel" in the upper right.
3. Then click "Stop my Server"
4. Click "Start my Server" which should take you to the menu of environments.



## 3 Pre-Lab Reading Quiz

Part of this lab is a pre-lab quiz. The pre-lab quiz is on Canvas. It is due **on Wednesday before Section A meets** (check Canvas for the time). It's not hard, but it does require you to read over the lab before class. If you are having trouble accessing it, make sure you are **logged into Canvas**.

### 3.1 How To Read the Lab For the Reading Quiz

The goal of reading the lab before starting on it is to make sure you have a preview of:

1. What's involved in the lab.
2. The key concepts of the lab.
3. What you can expect from the lab.
4. Any questions you might have.

These are the things we will ask about on the quiz. You *do not* need to study the lab in depth. You *do not* need to run the cells.

You should read these parts carefully:

- Paragraphs at the top of section/subsections
- The description of the programming assignment
- Any other large blocks of text
- The "About Labs in This Class" section (Lab 1 only)

You should skim these parts:

- The questions.

You can skip these parts:

- The "About Labs in This Class" section (Labs other than Lab 1)
- Commentary on the output of code cells (which is most of the lab)
- Parts of the lab that refer to things you can't see (like cell output)
- Solution to completeness questions.

## 3.2 Taking the Quiz

You can find it here: <https://canvas.ucsd.edu/courses/40763/quizzes>  
(<https://canvas.ucsd.edu/courses/40763/quizzes>)

The quiz is "open lab" -- you can search, re-read, etc. the lab.

You can take the quiz 3 times. Highest score counts.

## 4 Browser Compatibility

Your best option is to use Chrome, and that is the only browser the course officially supports. Here's the status for other browsers:

1. Chrome -- well tested. Preferred option. **Required for initial authentication**
2. Firefox -- seems ok, but not thoroughly tested.
3. Edge -- seems ok, but not thoroughly tested.
4. Safari -- not tested. Probably doesn't work.
5. Internet Explorer -- not tested. Probably doesn't work.

At the moment, the authentication step must be done in Chrome. You usually *will not* have to re-authenticate between labs, so if things work OK for the first lab, things will probably work here.

## 5 About Labs In This Class

*This section is the same in all the labs. It's repeated here for your reference.*

Labs are a way to **learn by doing**. This means you *must do*. I have built these labs as Jupyter notebooks so that the "doing" is as easy and seamless as possible.

In this lab, what you'll do is answer questions about how a program will run and then compare what really happened to your predictions. Engaging with this process is how you'll learn. The questions that the lab asks are there for several purposes:

1. To draw your attention to specific aspects of an experiment or of some results.
2. To push you to engage with the material more deeply by thinking about it.
3. To make you commit to a prediction so you can wonder why your prediction was wrong or be proud that you got it right.
4. To provide some practice with skills/concepts you're learning in this course.
5. To test your knowledge about what you've learned.

The questions are graded in one of three ways:

1. "Correctness" questions require you to answer the question and get the correct answer to get full credit.
2. "Completeness" questions require you to answer (even if incorrectly) all parts of the question to get credit.
3. "Optional" questions are...optional. They are there if you want to go further with the material.

Some of the "Completeness" problems include a solution that will be hidden until you click "Show Solution". To get the most from them, try them on your own first.

Many of the "Completeness" questions ask you to make predictions about the outcome of an experiment and write down those predictions. To maximize your learning, think carefully about your prediction and commit to it. **You will never be penalized for making an incorrect prediction.**

You are free to discuss "Completeness" and "Optional" questions with your classmates. You must complete "Correctness" questions on your own.

If you have questions about any kind of question, please ask during office hours or during class.



## 5.1 How To Succeed On the Labs

Here are some simple tips that will help you do well on this lab:

1. Read/skim through the entire lab *before* class. If something confuses you, you can ask about it.
2. Start early. Getting answers on piazza can take time. So think through the lab questions (and your questions about them) carefully.
  - A. Go through the lab once (several days before the deadline), do the parts that are easy/make sense
  - B. Ask questions/think about the rest
  - C. Come back and do the rest.
3. Start early. The DSMLP cluster gets busy and slow near deadlines. "The cluster was slow the night of the deadline" is not an excuse for not getting the lab done and it is not justification for asking for an extension.
4. Follow the guidelines below for asking answerable questions on piazza.

You may think to yourself: "If I start early enough to account for all that, I'd have to start right after the lab was assigned!" Good thought!



**The Cluster Will Get Slow** DSMLP and our cloud machines will get crowded and slow *before every deadline*. This is completely predictable. DSMLP can also get crowded due to deadlines in other courses. You need to start early so you can avoid/work around these slowdowns. Unless there's some kind of complete outage, we will not grant extensions because the servers are crowded.

## ▼ 5.2 Getting Help

You might run into trouble while doing this lab. Here's how to get help:

1. Re-read the instructions and make sure you've followed them.
2. Try saving and reloading the notebook.
3. If it says you are not authenticated, go to the [the login section of the lab](#) and (re)authenticate.
4. If you get a `FileNotFoundError` make sure you've run all the code cells above your current point in the lab.
5. If you get an exception or stack dump, check that you didn't accidentally modify the contents of one of the python cells.
6. If all else fails, post a question to piazza.

## 5.3 Posting Answerable Questions on Piazza

If you want useful answers on piazza, you need to provide information that is specific enough for us to provide a useful answer. Here's what we need:

1. Which part of which lab are you working on (use the section numbers)?
2. Which problem (copy and paste the *text* of the question along with the number).

If it's question about instructions:

1. Try to be as specific as you can about what is confusing or what you don't understand (e.g., "I'm not sure if I should do X or Y.")

If it's a question about an error while running code, then we need:

1. If you've committed anything, your github repo url.
2. If you've submitted a job with `cse142` you *must* provide the job id. It looks like this: `544e0cf2-4771-43c3-86f8-1c30d7af601f` . With the id, we can figure out just about anything about your job. Without it, we know nothing.
3. The *entire* output you received. There's no limit on how long an piazza post can be. Give us all the information, not just the last few lines. We like to scroll!

For all of the above **paste the text** into the piazza question. Please **do not provide screen captures**. The course staff refuses to type in job ids found in screen shots.

**We Can't Answer Unanswerable Questions** If you don't follow these guidelines (especially about the github repo and the job id), we will probably not be able to answer

## ▼ 5.4 Keeping Your Lab Up-to-Date

`pull-updates` is causing more trouble than it's worth, so it's been removed for this lab.

~~Occasionally, there will be changes made to the base repository after the assignment is released. This may include bug fixes and updates to this document. We'll post on piazza when an update is available.~~

~~In those cases, you can use `./pull-updates` to pull the changes from upstream and merge them into your code. You'll need to do this at a shell. It won't work properly in the notebook. Save your notebook in the browser first.~~

~~Then, change to your lab directory and do~~

~~`./pull-updates`~~

~~Then, reload this page in your browser.~~

## ▼ 5.5 Writing Code Outside Jupyter Notebook

The code for some programming assignments could get pretty long. If you'd like, you can develop outside of Jupyter Notebook.

You can do this by removing the call to `code()` and replacing it with a file name. Then `build()` will use the source code in the file.

**Don't overwrite your code:** `code()` does some checks to try to avoid overwriting your code and will throw an exception if it found modifications to files it wrote earlier. This seems to work pretty well, but I wouldn't trust it, so commit often.

## ▼ 5.6 Using VSCode

You can also develop remotely using Microsoft VSCode. You can find instructions from campus about how to do this on Datahub under "Visual Studio (VS) Code" at this link:

[https://support.ucsd.edu/services?id=kb\\_article\\_view&sysparm\\_article=KB0032269&sys\\_kb\\_id=01322d481b5ed514d1b0a935604bc](https://support.ucsd.edu/services?id=kb_article_view&sysparm_article=KB0032269&sys_kb_id=01322d481b5ed514d1b0a935604bc)  
([https://support.ucsd.edu/services?id=kb\\_article\\_view&sysparm\\_article=KB0032269&sys\\_kb\\_id=01322d481b5ed514d1b0a935604bc](https://support.ucsd.edu/services?id=kb_article_view&sysparm_article=KB0032269&sys_kb_id=01322d481b5ed514d1b0a935604bc))

The TAs report that this works fine.

A few things to note:

1. That page lists several ways of starting docker containers on the campus servers. The configuration for this class is a little unusual, and none of the other methods listed on that page have been tested for this class. I suspect they don't work, and we won't be fixing them.
2. You'll need to be on campus or on the campus VPN.
3. Using VSCode is not officially supported in this class. If it doesn't work for you, the TAs may be willing help you and you might have luck submitting a ticket to campus, but if you can't get it to work, you'll need to fall back on working through Jupyter Notebook.

## ▼ 5.7 How To Use This Document

You will use Jupyter Notebook to complete this lab. You should be able to do much of this lab without leaving Jupyter Notebook. The main exception will be some parts of the some of the programming assignments. The instructions will make it clear when you should use the terminal.

### 5.7.1 Logging In

If you haven't already, you can go to [the login section of the lab](#) and follow the instructions to login into the course infrastructure.

### 5.7.2 Running Code

Jupyter Notebooks are made up of "cells". Some have Markdown-formatted text in them (like this one). Some have Python code (like the one below).

For code cells, you press `shift-return` to execute the code. Try it below:

In [ ]:

```
print("I'm in python")
```

Code cells can also execute shell commands using the `!` operator. Try it below:

In [ ]:

```
!echo "I'm in a shell"
```

## ▼ 5.7.3 Telling What The Notebook is Doing

The notebook will only run one cell at a time, so if you press `shift-return` several times, the cells will wait for one another. You can tell that a cell is waiting if it there's a `*` in the [ ] to the left the cell:

In [\*]:



You'll can also tell *where* the notebook is executing by looking at the table of contents on the left. The section with the currently-executing cell will be red:

Fig. How to use the document

- 1.5.1 Logging In
- 1.5.2 Running Code
- 1.5.3 Telling What The Notebook is Doing
- 1.5.4 The Embedded Code

## ▼ 5.7.4 What to Do If Jupyter Notebook It Gets Stuck

First, check if it's actually stuck: Some of the cells take a while, but they will usually provide some visual sign of progress. If *nothing* is happening for more than 10 seconds, it's probably stuck.

To get it unstuck, you stop execution of the current cell with the "interrupt button":



You can also restart the underlying python instance (i.e., the confusingly-named "kernel" which is not the same thing as the operating system kernel) with the restart button:



Once you do this, all the variables defined by earlier cells are gone, so you may get some errors. You may need to re-run the cells in the current section to get things to work again.

You can also try reloading the web page. That will leave Python kernel intact, but it can help with some problems.

## ▼ 5.7.5 Common Errors and Non-Errors

1. If you get `sh: 0: getcwd() failed: no such file or directory`, restart the kernel.
2. If you get `INFO:MainThread:numexpr.utils>Note: NumExpr detected 40 cores but "NUMEXPR_MAX_THREADS" not set, so enforcing safe limit of 8.` . It's not a real error. Ignore it.
3. If you get a prompt asking `Do you want to cancel them and run this job?` but you can't reply because you can't type into an output cell in Jupyter notebook, replace `cse142 job run` with `cse142 job run --force` . (see useful tip below.)
4. If you get an `Error: Your request failed on the server: 500 Server Error: Internal Server Error for url=http://cse142l-dev.wl.r.appspot.com/file` , trying running the job again.
5. Sometimes `cse142 job run` will just sit there and seemingly do nothing. Weirdly, interrupting the kernel (button above) seems to jolt it awake and cause it to continue.
6. These errors while display CFGs are harmless:



```
Cannot determine entrypoint, using 0x00002560.
Warning: run r2 with -e bin.cache=true to fix relocations in disassembly
Cannot determine entrypoint, using 0x00001140.
Warning: run r2 with -e bin.cache=true to fix relocations in disassembly
```

```
/opt/conda/lib/python3.10/site-packages/pandas/plotting/_matplotlib/core.py:1114: UserWarning: No data for colormapping provided via 'c'. Parameters 'cmap' will be ignored
scatter = ax.scatter(
```



7. If you get `http.cookiejar.LoadError: '/home/yourusername/.djrcookies.txt does not look like like a Netscape format cookies file. remove the file and re-authenticate.`

8. If you get

```
You already have one or more jobs submitted or running.
a26fc9cc-ba36-4f49-89ea-1f36b16b5ea4    you@ucsd.edu    CREATED
2022-10-07 23:46:18.709330+00:00    true
Do you want to cancel them and run this job? [y/N]:
```

You can run `cse142 job run --lab intro --take NOTHING true` and it should fix it.

9. If you get a big list of files that ends like this:

```
.cfiddle/builds/build/MORE_INCLUDES_-I_cse142L_CSE141pp-Tool-Mon
eta_moneta__nibble/nibble_29.so
.cfiddle/builds/build/MORE_INCLUDES_-I_cse142L_CSE141pp-Tool-Mon
eta_moneta__nibble/nibble_76.so
If you want to upload more than 200 files, pass '--input-file-co
unt-limit '.
```

It means you have too many files in your local directory. You can delete some of them or do what the error says and pass a large value to `--input-file-count-limit` (although that will make running jobs quite slow for you). A good candidate for deletion is your `.cfiddle` folder. You remove but you may need re-run some of your `build()` cells afterwards:

In [ ]:

```
#!/rm -rf .cfiddle
```

10. If you get this

```
SourceCodeModified: The contents of foo.cpp have changed since cfiddle wrote them last. Aborting to prevent loss of work.
```

This means that `cfiddle's code()` function detected a change to the file mentioned (`foo.cpp` in the error above) and it is refusing to overwrite it, so it doesn't destroy your changes. This can happen, for example, if you've edited the file in VSCode. You can either

1. Delete the file (``rm foo.cpp``) and re-execute the cell.
2. Delete the file, and replace the argument to ``code()`` with the new contents of the file, so you can keep editing in Jupyter notebook.
3. Keep editing the file externally and replace the call to ``code()`` with the file name.

## ▼ 5.7.6 Useful Tips

1. If you need to edit a cell, but you can't you can unlock it by pressing this button in the tool bar (although you probably shouldn't do this because it might make the lab work incorrectly. A better choice is to copy and paste the cell, *and then* unlock the copy):



## ▼ 5.7.7 The Embedded Code

The code embedded in the lab falls into two categories:

1. Code you need to edit and understand.
2. Code that you do not need to edit or understand -- it's just there to display something for you.

For code in the first category, the lab will make it clear that you need to study, modify, and/or run the code. If we don't explicitly ask you to do something, you don't need to.

Most of the code in the second category is for drawing graphs. You can just run it with shift-return to see the results. If you are curious, it's mostly written with `Pandas` and `matplotlib`. These cells should be un-editable. However, if you want to experiment with them, you can copy *the contents* of the cell into a new cell and do whatever you want (If you copy the cell, the copy will also be uneditable).

**Most Cells are Immutable** Many of the cells of this notebook are uneditable. The only ones you should edit are some of the code cells and the text cells with questions in them.

**Pro Tip** The "carrot" icon in the lower right (shown below) will open a scratch pad area. It can be a useful place to do math (or whatever else you want).



## ▼ 5.7.8 Showing Your Work

Several questions ask you to show your work for calculations. We don't need anything fancy. Many of the questions ask you to compute something based on results of an experiment. Your experimental results will be different than others', so your answer will be different as well.

To make it possible to grade your work (and give you partial credit), we need to know where your answer came from. This is why you need to show your work. For instance this would be fine as answer to "On average, how many weeks do you have per lab?":

Weeks in quarter/# of labs =  $10/5 = 2$  weeks/lab

2 significant figures is sufficient in all cases, but you can include more, if you want.

If you are feeling fancy, you can use LaTeX, but it's not at all required.

When it's appropriate, you can also paste in images. However, Jupyter Notebook is flaky about it. Save frequently.

### ▼ 5.7.9 Saving Your Work and Making Sure Your Connected to the Server

In theory, Jupyter Notebook saves automatically. However, a few things can go wrong:

If your Datahub server shuts down, you can still edit your notebook, but you won't be able to save it.

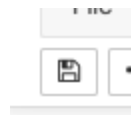


You can tell your server has stopped if there's a red box in the upper right that says "Not Connected":



If this happens, you should stop working, restart your server and reload the lab.

In any case, it's a good idea to save frequently:



## 5.7.10 Answering Questions

Throughout this document, you'll see some questions (like the one below). You can double click on them to edit them and fill in your answer. Try not to mess up the formatting (so it's easy for us to grade), but at least make sure your answer shows up clearly. When you are done editing, you can `shift-return` to make it pretty again.

A few tips, pointers, and caveats for answering questions:

1. The answers are all in [github-flavored markdown](https://guides.github.com/features/mastering-markdown/) (<https://guides.github.com/features/mastering-markdown/>) with some html sprinkled in. Leave the html alone.
2. Many answers require you to fill in a table, and many of the `|` characters will be missing. You'll need to add them back.
3. The HTML needs to start at the beginning of a line. If there are spaces before a tag, it won't render properly. If you accidentally add white space at the beginning of a line with an html tag on it, you'll need to fix it.
4. Text answers also need to start at the beginning of a line, otherwise they will be rendered as code.
5. Press `shift-return` or `option-return` to render the cell and make sure it looks good.
6. There needs to be a blank line between html tags and markdown. Otherwise, the markdown formatting will not appear correctly.

You'll notice that there are three kinds of questions: "Correctness", "Completeness", and "Optional". You need to provide an answer to the "Completeness" questions, but you won't be graded on its correctness. You'll need to answer "Correctness" questions correctly to get credit. The "Optional" questions are optional.

Give it a try:

### Question 1 (Completeness)

**What will you do with your Jupyter Notebook to turn it in (Delete *all* incorrect answers)? Then fill in the table. Fix the formatting of the last line and the closing tag.**

1. Print it out on paper and slide it under the professor's door.
2. Follow the directions at the end of the lab to produce a pdf
3. Read it aloud to a TA during office hours
4. Submit via gradescope.

Random Fact	Value	Second choice
Your favorite color		
Favorite food on campus		

Random Fact	Value	Second choice
Time of day		

This answer is formatted poorly.  
</div>



Show Solution



## 6 Logging In To the Course Tools

In the course you will use some specialized tools to let you perform detailed measurements of program behavior. To use them you need to login with your @ucsd.edu email address using the instructions below. **You need to use the email address that appears on the course roster. That's the email address we created an account for. In almost all cases, this is your @ucsd.edu email address.**

You'll probably only have to do this once this quarter, but if you get an error about not being authenticated, just re-authenticate. You can return to this notebook (or any other of the lab notebooks) to login at any time.

Here's what to do:

1. Enter your @ucsd.edu email address (without the '<>') in quotes after login below. It'll take a few seconds to load.
2. Click the google "G" login button below and login with your @ucsd.edu email address.
3. **Click the google button regardless of whether it says "sign in" or "signed in". Then be sure to select your @ucsd.edu account if it shows you multiple google acocunts**
4. You'll see a very long string numbers an letters appear above. Click "Copy it" to copy it.

**Note:** If it doesn't give you a choice about which account to log into and authentication fails, that means you are logged into a single Google account and that account is *not* your @ucsd.edu account. You'll have to log into your @ucsd.edu through Gmail or through Chrome's account manager and then try again.

**Use Chrome** The login process doesn't seem to work properly with Safari or Firefox. Use Chrome to login. You can use any of the other compatible browsers you want for the doing the rest of the lab, and it should be fine.

In [3]:

```
login("Your @ucsd.edu email address")
```

Next step: Paste it below between the quote marks. Press shift-return .

In [4]:

```
token("eyJhbGciOiJSUzI1NiIsImtpZCI6IjQ1NmI1MmM4MWUzNmZlYWQyNTkyMzFhNjk0")
```

It should have replied with

You are authenticated as <your email>

You are now logged in! Try submitting a job:

In [6]:

```
!cse142 job run --take NOTHING "echo Hello World"
```

If you see "Hello World", you're all set. Proceed with the lab!

Delete your token from the above cell ( `token("...")` ). Because your token is essentially your username and password combined, you should treat it like a password or ssh private key. **Sharing your token with another student or possessing another student's token is an AI violation.**

In [ ]:



## 7 Grading

This is a one or two week lab depending on which section you are in (due to the holiday).

Your grade for this lab will be based on your completion and submission of this notebook.

Part	value
Reading quiz	3%
Jupyter Notebook	95%
Post-lab survey	2%

We will grade 5 of the "completeness" problems. They are worth 3 points each. We will grade all of the "correctness" questions.

Check Gradescope for the due dates.

Instructions for submitting the lab are at the end of the lab.

No late work or extensions will be allowed for anything less than severe system outages.



## 8 Academic Integrity Agreement

To continue in the class, you need to agree to the following:

At UCSD, academic integrity[1] means that you have the courage, even when it is difficult, to only submit academic work that is honest, responsible, respectful, fair, and trustworthy. When you excel with integrity in computer science, it means that you:

**Honest** submit work that is a truthful demonstration of your knowledge and abilities (rather than the knowledge and abilities of another)

**Responsible** manage your time so that you are not pressured to complete an assignment at the last minute

**Respectful** acknowledge the contributions of others to your work by citing them when you've used their words or ideas (e.g., after I've spoken to classmates or after I've used portions of a code written by another if permitted)

**Fair** complete your academic work according to stated standards and expectations even when it takes longer or re struggling

**Trustworthy** can be trusted to be honest, responsible, respectful, and fair even when no one is watching you.

When you act contrary to these values, you are cheating. Cheating undermines trust between students and professors, the value of the UCSD degree, and your learning/development of skills.

While we can't list every behavior that would be cheating, we can give you some illustrative examples like the following:

- Submitting any source code written by another person or copied from another person, submitting homework answers which were produced by another student.
- Submitting code/homework you have previously submitted to another course for credit without first obtaining permission from the instructor. The same restriction holds for publicly available code/homework solutions that you haven't written. Taking notes taken during any discussions with classmates about an assignment is prohibited.
- Using words or text written by someone else without citing text appropriately. Every figure or sentence fragment must be appropriately decorated with quotation marks or indentation to indicate very clearly that someone else wrote the text. In addition, the passage must be labeled with a citation or citation number which refers to a footnote or bibliographic entry. Citing a paper once is not enough. Remember: citations should be used to illuminate a viewpoint which you hold. They are not a substitute for expressing your own ideas in your own words.
- Submitting any portion(s) of an assignment you have previously submitted for credit in another course.
- Copying from a neighbor during an exam or using an unauthorized aid to help you on your exam.
- Altering a graded exam or assignment and resubmitting it for regrade

- Allowing someone else to complete an assignment or exam for you, or allowing them to pretend to be you in class (e.g., by signing an attendance form or clicking for you).
- Making available to others source code, documentation, or notes useful for completing an assignment. You should neither produce, procure, nor accept such material. This includes students in current, past, and future offerings of the course, and applies to electronic transmissions including email, web pages, ftp, and so on, as well as hard copy such as source code listings.
- Possessing (at any time) source code, data, or answers to homeworks or lab questions created by another student. Having had any of these items in your possession (e.g., in your directory on the campus servers or on your personal computer) at any time, constitutes cheating. You should never accept these materials from anyone for any reason.
- Running other students' code from your account or allowing another student submit code using your credentials.

If the behavior you are considering isn't listed here, don't assume that it is allowed. Rather, you should always do independent work unless told otherwise. And before completing your academic work in a certain way, you should ask is it honest, respectful, responsible, fair, and trustworthy. You can also ask yourself "Would I be okay if my methods were exposed to the TA, professors, and fellow students?" If the answer is no, you shouldn't do it.

**If you have any questions about what is and isn't cheating, be sure to discuss them with the instructor.**

Any student who cheats, thereby undermining integrity, will be reported to the Academic Integrity Office. Students who cheat face various disciplinary sanctions as well as academic penalty imposed by the instructor in the course. *Academic penalties include, but are not limited to, receiving a grade of 0 for the assignment or test in question, and receiving an 'F' for the course.*

[1] For more information on academic integrity, including how you can excel with integrity, as well as information on sanctioning guidelines for cheating, visit the Academic Integrity Office website at: <http://academicintegrity.ucsd.edu> (<http://academicintegrity.ucsd.edu>)

## Question 2 (Correctness - 1pts)

### Please affirm your adherence to this agreement

Type 'I excel with integrity' here: [ type it (leave the brackets) ]

By submitting this file, I, [ Your Name ], a student enrolled in CSE142L affirm the principle of academic integrity and commit to excel with integrity by completing all academic assignments in the manner expected as described above, informing the instructor of suspected instances of academic misconduct by my peers, and fully engaging in the class and its related assignments for the purpose of learning.

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## 9 Skills to Learn

1. Get access to docker
2. git /GitHub basics
3. Navigating a Jupyter Notebook
4. Understanding cfiddle code
5. Think about code and predict its behavior.
6. Optimize a simple algorithm.

## 10 Building and Running Code

In these labs you will be doing most of your coding within Jupyter Notebook using an interactive code analysis system called `cfiddle` (<https://github.com/NVSL/cfiddle>). `Cfiddle` is a Python library that lets you write code, compile it, analyze it, execute it, and analyze its performance all in a Jupyter Notebook.

### 10.1 Compiling and Inspecting Code with CFiddle

Here's how you compile Hello World with `Cfiddle`:

In [ ]:

```
hello_world = build(code(r"""
#include"cfiddle.hpp" // Connect to Cfiddle's performance measurement m
#include<iostream>

extern "C" void hello_world(int k) {
    start_measurement(); // start measuring
    uint64_t sum = 0;
    for(int i = 0; i < k; i++) {
        sum += i;
    }

    std::cout << "Hello World! The sum of the day is " << sum << ".\n";
    end_measurement(); // stop measuring
}

""", file_name="hello_world.cpp"))
```

The `code()` function takes a Python string, writes it to a file and returns the filename. The `r` before the string makes it a "raw" string which disables Python's escape sequence processing. `extern "C"` makes it easier to find `hello_world`. We'll learn more about it in the next lab.

The `hello_world` is now *an array* of Python objects that represent compiled code. Usually, we'll just be interested in the first element of the array. For instance, we can look at the code

```
In [ ]: print(hello_world)
```

```
In [ ]: hello_world[0].source()
```

Or it's assembly:

```
In [ ]: hello_world[0].asm("hello_world")
```

We can also configure how the code is compiled. For instance, we can compile it with optimizations by setting the `build_parameters` argument:

```
In [ ]: hello_world_twice = build("hello_world.cpp", build_parameters=arg_map(C
```

the `arg_map` is a utility function that generates all combinations of its arguments (we'll see more of this later), but for now the important part is that it will compile with and without optimizations. The result is that the `hello_world` has two entries in: The first is the compiled code without optimizations and second is the code with optimizations. We can now compare the assembly for the two:

```
In [ ]: compare([hello_world_twice[0].asm("hello_world"), hello_world_twice[1].
```

### Question 3 (Optional)

**What did the compiler do differently in the two versions? (We'll investigate this in detail in Lab 2)**



## 10.2 Running Code With CFiddle

CFiddle can also run code. Let's run `hello_world()` :

In [ ]:

```
results = run(hello_world, function="hello_world", arguments=arg_map(k=
```

The function argument tells `cfiddle` which function to invoke, and arguments get passed to the function.

Running the code took a while, because, for this class, `cfiddle` is configured to run your code on our servers in the cloud, but you can see the output buried in there:

Hello World! The number of the day is 6.

`results` now has some measurements about the execution of `hello_world()`. We can render them as "data frame" with is a fancy 2D array with labeled columns (Technically it's a [Pandas data frame](https://pandas.pydata.org/docs/reference/api/pandas.DataFrame.html) (<https://pandas.pydata.org/docs/reference/api/pandas.DataFrame.html>)):

In [ ]:

```
results.as_df()
```

These numbers don't mean much to you at the moment, but, for instance, `ET` is the execution time. So we can see that running `hello_world` didn't take very long when `k` is 4. It's so short, in fact, that I doubt measurement is accurate. We need at least 0.3 seconds of execution for an accurate measurement.

We can also run `hello_world()` with different parameters and different compilation options. We can pass a list of values to `arg_map()` that vastly increase `k`, and we'll use `hello_world_twice` which holds two different compiled versions of our function.

In [ ]:

```
# this one takes a while. But less than a minute.
results = run(hello_world_twice, function="hello_world", arguments=arg_
```

Here, `cfiddle` ran `hello_world()` a total of 6 times for the 6 combinations of compiler flags and values for `k`: Three times without optimizations with `k == 400000000, 800000000, 1600000000` and three times with optimizations. And we have the results to prove it:

In [ ]:

```
results.as_df()
```

### Question 4 (Correctness - 2pts)

How much speedup does `-O4` provide compared to `-O0` for the smallest and largest value of `k`?

### Question 5 (Completeness)

Add another value of  $k$  to the cell above that is 2x the highest value. Re-run the experiment and analyze the speedup. What happened? Why (give your best idea)?

## 10.3 Local Execution

If you're debugging or running simple experiments you can run code locally. It'll be faster but performance measurements will be pretty meaningless because the datahub machines are shared across many users.:

In [ ]:

```
with local_execution():  
    results = run(hello_world_twice, function="hello_world", arguments=
```

## 11 Analyzing Program Behavior with the Performance Equation

**Keep track of questions you have.** This lab is more about collecting questions than finding answers. I've called out some interesting questions throughout the lab. The last question of the lab asks for *other* questions you had while examining the data you'll collect below, so keep track of them as you work through the lab.

### 11.1 Meet the Code!

In this lab we are going to analyze a few simple functions in `microbench.cpp`. The first is `baseline_int()`:

In [171]:

```
microbench = build("microbench.cpp")
microbench[0].source("baseline_int")
```

100%

1/1 [00:00&lt;00:00, 61.81it/s]

Out[171]:

```
extern "C" uint64_t * baseline_int(unsigned long int size, uint reps) {
    uint64_t * array = new uint64_t[size];
    for(uint i = 0; i < size; i++) {
        array[i] = 0;
    }
    start_measurement();
    for(uint r = 0 ; r < reps; r++) {
        for (uint j = 0; j < 3; j++) {
            for(uint i= 1 ; i < size; i++) {
                array[i] += i/(1+j)+array[i - 1];
            }
        }
    }
    end_measurement();
    return array;
}
```

`baseline_int()` initializes `array` and then does some multiplies and additions to update it's contents. It's not a useful computation, so don't spend time trying to figure out what it does. The `reps` argument controls how many times we'll run it.

## ▼ 11.2 Measuring The Performance Equation

Now that we know how to take measurements, we can try to understand `baseline_int()` 's performance. We will do this using the performance equation:

$$ET = IC * CPI * CT$$

So, we'll need to measure `IC` (instruction count), `CPI` (cycles per instruction), `CT` (cycle time), and `ET`, and we'll do that using "performance counters". Performance counters are the CPU provides to count events that occur in the processor.

`cfiddle` has support for for reading performance counters built in, we just need to tell it to collect data. We can do that by passing some extra options to `run()`. Here's what they mean:

1. Passing `run_options` to `run()` lets us control the execution environment for the program. `MHz` set's the clock speed.
2. `perf_counters` controls which performance counters we collect.

In this case, `PERF_COUNT_HW_INSTRUCTIONS`, and `PERF_COUNT_HW_CPU_CYCLES` count the number of instructions executed and the number of CPU cycles, respectively. There are hundreds of other events we can count. We'll use some of them later in the class.

In [ ]:

```
r = run(microbench,
       function="baseline_int",
       arguments=arg_map(size=[1024*1024], reps = 25),
       run_options=arg_map(MHz=[3500]),
       perf_counters=["PERF_COUNT_HW_INSTRUCTIONS", "PERF_COUNT_HW_CPU_CYCLES"])
```

Here's the raw output

In [ ]:

```
r.as_df()
```

The key fields are `ET` (execution time) and the results of the performance counter measurements: `PERF_COUNT_HW_INSTRUCTIONS` and `PERF_COUNT_HW_CPU_CYCLES`.

### Question 6 (Correctness - 4pts)

Using the value names in the table above, give expressions for `IC`, `CT`, and `CPI`. (You cannot use `ET` in your expressions.)

- `IC =`
- `CPI =`
- `CT =`

Then we'll do some calculations on the output and print them nicely. `PE_calc()` applies the same math you figured out for the question above (assuming you got it right) to the results. We're going to use this many times throughout the course.

In [ ]:

```
PE_calc(r.as_df())
```

We are going to see a lot of data like this so, let's be clear about what they mean:

function	size	requestedMHz	realMHz	cycles	IC	CPI	CT	ET
function	size of array	MHz requested	Measured MHZ	Processor Cycles	dynamic instructions executed	Cycles/instruction	Cycle Time	Execution Time

Note that `MHz` and `realMHz` don't quite match. This can be due to noise in measurements or the processor running faster than we requested.

## 11.3 Meet Your Processor

Let's gather a little bit of information about the CPU we'll be using. We can just ask the OS, but we need to ask on our servers in the cloud.

To do that we'll need to run a command line program remotely. For that we'll use `cse142 job run`. It runs a command line remotely (this is the same mechanism that `cfiddle` uses internally). The `--force` option cancels any jobs you currently have running, and `--take NOTHING` keeps it from copying any local files to the server. `lscpu` is the command to run.

In [ ]:

```
!cse142 job run --force --take NOTHING lscpu
```

As you can see it's a E-2146G CPU running at 3.5GHz. The model number probably doesn't mean anything to you, but that's what [google is for](https://ark.intel.com/content/www/us/en/ark/products/134866/intel-xeon-e-2146g-processor-12m-cache-up-to-4-50-ghz.html) (<https://ark.intel.com/content/www/us/en/ark/products/134866/intel-xeon-e-2146g-processor-12m-cache-up-to-4-50-ghz.html>).

### Question 7 (Correctness - 2pts)

**Based on the program output and link above, fill out the table below. Some of the information is in the output above, some you'll need to google for. "Technology node" is roughly the size of the smallest transistors used in designing the chip.:**

Parameter	Value
How many physical cores?	[YOUR ANSWER HERE]
How many total threads?	
Base processor frequency	
Max turbo boost frequency	
Process technology node (nm)	
L1 data cache size (d)	
L1 instruction cache size (i)	
L2 Cache size	
L3 Cache size	
Year introduced	
What does the 'E' mean in the model number?	
What does the 'G' mean in the model number?	

## Question 8 (Optional)

**Here's some other interesting questions to look into about our processor:**

1. What's the Intel code name for our processor's microarchitecture? When was it introduced?
2. What's the maximum clock rate at which processors with this microarchitecture can run?
3. What's the most cores available on a single die with this microarchitecture?
4. What major revision of Intel's microarchitecture is it a part of? How old is this basic design?
5. What do all the things under `Flags` mean?

Here are some resources:

1. [https://en.wikipedia.org/wiki/List\\_of\\_Intel\\_CPU\\_microarchitectures](https://en.wikipedia.org/wiki/List_of_Intel_CPU_microarchitectures) ([https://en.wikipedia.org/wiki/List\\_of\\_Intel\\_CPU\\_microarchitectures](https://en.wikipedia.org/wiki/List_of_Intel_CPU_microarchitectures))
2. <https://en.wikichip.org/wiki/WikiChip> (<https://en.wikichip.org/wiki/WikiChip>)

## ▼ 11.4 Instruction Count

Let's see how changing the instruction count ( `IC` ) affects performance. There are two ways we can increase instruction count for `baseline_int()` :

1. We can increase the number of times the outer loop runs with `reps` .
2. We can increase the size of `array` with the `size` parameter.

### ▼ 11.4.1 Running The Experiment Multiple Times

Let's run the `baseline_int()` with three values for `reps` : 25, 50, and 100.

**Answer the question *before* you look at the results.** The goal of this question (and many more to follow) is for you to predict the answer and then see if the results match your intuition. Don't be discouraged if you frequently get the prediction wrong: They are intentionally challenging and some of them have intentionally non-intuitive results. Also, a major goal of the lab is to highlight behavior that seems non-intuitive so you can improve your intuition.

Kick off the experiment below and the answer this question:



### Question 9 (Completeness)

In a moment, you'll see four graphs that show how each term of the performance equation changes as we increase `reps` .

What *shape* do you think each curve will have (linear? curved?) and what *direction* will it go (increasing? Decreasing? flat?)? For each term, predict the ratio between it's value at 100 and 25 (i.e., `value_at_100/value_at_25` ).

	IC	CPI	CT	ET
Shape	[PUT YOUR ANSWERS IN THE TABLE]			
Direction				
100 vs 25 ratio				

In [ ]:

```
#This takes a while... (like a minute)
d = run(microbench, function="baseline_int",
        arguments=arg_map(size=1024*1024, reps=[25,50,100]),
        run_options=arg_map(MHz=3500),
        perf_counters=["PERF_COUNT_HW_INSTRUCTIONS", "PERF_COUNT_HW_CPU"],
        results = PE_calc(d.as_df())
display(results)
```

In [ ]:

```
plotPE(df=results, lines=True, what=[ ('reps', "IC"), ("reps", "CPI"),
```

### Question 10 (Correctness - 4pts)

Use the "scratch pad" (little triangle thing in the lower-right of this window) or a calculator to compute the actual ratio of the values at `reps = 100` and `reps = 25` and enter them in the table below. How does these results differ from what you expected? (We won't check that you actually used the scratch pad, but you should. It's a very useful tool.)

	IC	CPI	CT	ET
100 vs 25 reps ratio				

	IC	CPI	CT	ET
Differences compared to your expectations (if any):				

## ▼ 11.4.2 Increasing the Size of array

Instead of increasing `reps` we can increase the size of `array` so we'll run the same computation over more data. We will vary the size over a very large range: 10,000 to 20,480,000.

Here's the code again, for your reference

In [ ]:

```
display(microbench[0].source("baseline_int"))
```

Kick off the cell below to collect the data. While it's running answer this question:

### Question 11 (Completeness)

This time, the four graphs will show each term of the performance equation changes as we increase `size`. What *shape* do you think each curve will have (linear? curved?) and what *direction* will it go (increasing? Decreasing? flat?)? For each term, predict the ratio between it's value at 5,120,000 and its value at 320,000.

	IC	CPI	CT	ET
Shape				
Direction				
5,120,000 vs 320,000 ratio				

In [ ]:

```
#This takes a while...
d = run(microbench, function="baseline_int",
        # this call to exp_range generates a sequence, a, where a_0 = 1
        arguments=arg_map(size=exp_range(10000, 20480000, 2), reps=20),
        run_options=arg_map(MHz=3500),
        perf_counters=["PERF_COUNT_HW_INSTRUCTIONS", "PERF_COUNT_HW_CPU"
```

In [ ]:

```
df = PE_calc(d.as_df())
display(df)
plotPE(df=df, lines=True, what=[ ('size', "IC"), ("size", "CPI"), ("size", "CT"), ("size", "ET")])
```

### Question 12 (Correctness - 4pts)

**Compute the actual ratio of the values at 5,120,000 and 320,000. How does these results differ from what you expected? What is the speedup for a an array of 5,120,000 vs an array of 320,000?**

	IC	CPI	CT	ET
5,120,000 vs 320,000 ratio				
Differences compared to your expectations				

**Speedup for 5,120,000 vs 320,000-element arrays:**

**Interesting Question:** Why does increasing the size of array change CPI? And why does this change occur so quickly and at only over a certain range of sizes?



## ▼ 11.5 Cycle Time

Next, we'll take a look at how clock rate affects performance. Before we do, though, let's see what our options are for clock rate on our machine. We'll use `cpupower` for this, a common Linux utility. Again, we'll run it with `cse142 job run` to run it in the cloud:

In [ ]:

```
!cse142 job run --force 'cpupower frequency-info -n'
```

As you can see, the processors in our target systems can run between 800MHz and 3501MHz at mostly 200MHz increments.

Let's see how that affects things by plotting execution time as a function of clock speed (we are skipping 3501MHz for the moment. We'll come back to it.). The readings for the current clock speed may vary from run to run. It just ends up at whatever the last experiment left it at.

Kick off the cell below to collect the data. While it's running answer this question:

### Question 13 (Completeness)

We are going to plot four graphs that show how each term of the performance equation changes as we increase clock rate. What *shape* do you think each curve will have (linear? curved?) and what *direction* will it go (increasing? Decreasing? flat?)? For each term, predict the ratio between its value at 3500MHz and its value at 1800Mhz.

	IC	CPI	CT	ET
Shape				
Direction				
3500MHz/1800Mhz ratio				

In [ ]:

```
#This takes a while...
ct_data = run(microbench, function="baseline_int",
              arguments=arg_map(size=1024*1024*32, reps=1),
              run_options=arg_map(MHz=[800, 1000, 1200, 1400, 1600, 1800,
              perf_counters=["PERF_COUNT_HW_INSTRUCTIONS", "PERF_COUNT_HW
```

In [ ]:

```
display(ct_data.as_df())
ct_df = PE_calc(ct_data.as_df())
display(ct_df)
plotPE(df=ct_df, lines=True, what=[ ('requestedMHz', "IC"), ("requested
```

### Question 14 (Correctness - 4pts)

Compute the actual ratio of the values at 3500MHz and 1800MHZ. How do these results differ from what you expected? How much speedup does double doubling the clock rate provide?

	IC	CPI	CT	ET
3500MHz/1800Mhz ratio				
Differences compared to your expectations:				

**How much speedup does doubling the clock rate from 1800MHz to 3500MHz provide? (show your work):**

**Interesting question:** How can clock rate affect CPI ?



The data above had `size = 1024*1024*32` and `reps = 1`. Let's call that "1-big". Note that, for 1-big, IC did not change significantly with clock speed (although, why did it change at all?).

Consider another configuration called "32-small" with `size = 1024*1024` and `reps = 32`. Kick off the cell below and answer this question:

### Question 15 (Completeness)

**How will 1-big's behavior compare to 32-small's? Fill out the table with your predictions:**

Expression	Value
1-big IC / 32-small IC (Your estimate)	
1-big ET @ 3500MHz/ 1-big ET @ 800Mhz (calculated from the data above)	
32-small ET @ 3500MHz/32-small ET @ 800Mhz (your estimate)	

In [ ]:

```
#This takes a while...
ct_data2 = run(microbench, function="baseline_int",
               arguments=arg_map(size=1024*1024, reps=32),
               run_options=arg_map(MHz=[800, 3500]),
               perf_counters=["PERF_COUNT_HW_INSTRUCTIONS", "PERF_COUNT_HW_
```

In [ ]:

```
display(ct_data2.as_df())
ct2_df = PE_calc(ct_data2.as_df())
display(ct2_df)
```

### Question 16 (Correctness - 1pts)

Fill in the table to see how your predictions did.

Expression	Actual values
1-big IC / 32-small IC	
32-small ET @ 3500MHz/32-small ET @ 800Mhz	

Interesting question: Why doesn't clock rate always affect CPI ?"



## 11.6 Cycles Per Instruction

Unlike IC and CT we can't set CPI directly, but we can adjust the code and see how CPI changes. We'll do this in two ways. First, we'll change the data type we are operating on. Then, we'll change the compiler options. Finally, we'll restructure the code.

### 11.6.1 Floating Point vs Integer Operations

`baseline_double()` (on the left) that is identical to `baseline_int()` (on the right) but uses 64-bit floating point values (of type `double`) instead of 64-bit integers (`uint64_t`):

In [172]:

```
compare([microbench[0].source(show="baseline_double"), microbench[0].source(show="baseline_int")])
```

```
extern "C" double *baseline_double(
    unsigned long int size, uint r
    eps) {
    double * array = new double[size];
    for(uint i = 0; i < size; i++) {
        array[i] = 0;
    }
    start_measurement();
    for(uint r = 0 ; r < reps; r++)
        for (double j = 0; j < 3; j++) {
            for(uint i = 1 ; i < size; i++) {
                array[i] += i/(1+j)+array[i - 1];
            }
        }
    end_measurement();
    return array;
}
```

```
extern "C" uint64_t * baseline_int(
    unsigned long int size, uint reps) {
    uint64_t * array = new uint64_t[size];
    for(uint i = 0; i < size; i++) {
        array[i] = 0;
    }
    start_measurement();
    for(uint r = 0 ; r < reps; r++) {
        for (uint j = 0; j < 3; j++) {
            for(uint i = 1 ; i < size; i++) {
                array[i] += i/(1+j)+array[i - 1];
            }
        }
    }
    end_measurement();
    return array;
}
```

Kick off the the cell below to run both functions, and answer this question:

### Question 17 (Completeness)

How do you think each term in the performance equation will change for `baseline_double()` compared to `baseline_int()` ?

IC:

CPI:

CT:

ET:

In [ ]:

```
cpi_data = run(microbench, function=["baseline_int", "baseline_double"],
              arguments=arg_map(size=1024*1024, reps=25),
              run_options=arg_map(MHz=3500),
              perf_counters=["PERF_COUNT_HW_INSTRUCTIONS", "PERF_COUNT_
```

In [ ]:

```
cpi_df = PE_calc(cpi_data.as_df())
display(cpi_df)
plotPEBar(df=cpi_df, what=[ ('function', "IC"), ("function", "CPI"), ('
```

### Question 18 (Completeness)

**How did the results for each term in the PE differ from your predictions (if they did)?**

**IC:**

**CPI:**

**CT:**

**ET:**

### Question 19 (Optional)

**In `microbench.cpp` there are also `baseline_char()` and `baseline_float()`. Add those functions to the experiment above. What do you find?**

**Interesting question:** How and why do the datatypes we use change IC and CPI ?





## ▼ 11.6.2 The Compiler's Effect

microbench.cpp contains the following function:

In [173]:

```
display(microbench[0].source("baseline_int_04"))
```

```
extern "C" uint64_t *__attribute__((optimize(4))) baseline_int_04 (unsigned long int size, uint reps) {
    uint64_t * array = new uint64_t[size];
    for(uint i = 0; i < size; i++)
        array[i] = 0;

    start_measurement();
    for(uint r = 0 ; r < reps; r++) {
        for (uint j = 0; j < 3; j++) {
            for(uint i= 1 ; i < size; i++) {
                array[i] += i/(1+j)+array[i - 1];
            }
        }
    }
    end_measurement();
    return array;
}
```

It's identical to `baseline_int()` except that for the `__attribute__((optimize(4)))` which is a little bit of `gcc` magic to optimize this functions as much as it can (it's the equivalent of passing `-O4` on the command line but just for this function).

Let's see how optimizations affect performance. Kick off the experiment in the cell below and answer this question while it runs:

### Question 20 (Completeness)

How do you think each term in the performance equation will change for `baseline_int()` compared to `baseline_int_04()` ?

IC:

CPI:

CT:

ET:

In [ ]:

```

cpi_opt_data = run(microbench, function=["baseline_int", "baseline_int_0"],
                  arguments=arg_map(size=8388608, reps=10),
                  run_options=arg_map(MHz=3500),
                  perf_counters=["PERF_COUNT_HW_INSTRUCTIONS", "PERF_COUNT_

```

In [ ]:

```

cpi_opt_df = PE_calc(cpi_opt_data.as_df())
display(cpi_opt_df)
plotPEBar(df=cpi_opt_df, what=[ ('function', "IC"), ("function", "CPI")

```

### Question 21 (Completeness)

**Based on the data above, describe in words what affect the optimizations had on the code and the value of each term of the PE.**

IC:

CPI:

CT:

ET:

### ▼ 11.6.3 Code Structure

These two functions increment all the elements in an array by 1.0 , but they do it slightly different ways.

In [174]:

```
compare([microbench[0].source(show="matrix_row_major"),
        microbench[0].source(show="matrix_column_major")])
```

```
extern "C" uint64_t *__attribute__((optimize(4))) matrix_row_major(
    unsigned long int size, uint reps) {
    double * array = new double[size];

    start_measurement();
    for(uint r = 0 ; r < reps; r++)
        for(uint i = 0; i < size/ROW_SIZE; i++) {
            for (int k = 0; k < ROW_SIZE; k++) {
                array[i*ROW_SIZE + k] += 1.0; // This Line
            }
        }
    end_measurement();
    return (uint64_t*)array;
}
```

```
extern "C" uint64_t *__attribute__((optimize(4))) matrix_column_major(
    unsigned long int size, uint reps) {
    #define ROW_SIZE 1024
    double * array = new double[size];

    start_measurement();

    for(uint r = 0 ; r < reps; r++)
        for (int k = 0; k < ROW_SIZE; k++) {
            for(uint i = 0 ; i < size/ROW_SIZE; i++) {
                array[i*ROW_SIZE + k] += 1.0; // This Line
            }
        }
    end_measurement();
    return (uint64_t*)array;
}
```

We'll run both versions and compare their performance. Kick off the experiment below and answer this question:

### Question 22 (Completeness)

If **size** is equal to 8,388,608 how many times will "This Line" execute in each function? Do you think one will be faster than the other? Why?

How many times does This Line execute in `matrix_row_major()` :

How many times does This Line execute in `matrix_column_major()` :

Is there any difference in the "Big-O" running time of these two functions?

Do you think one will be faster than the other? Why?

In [ ]:

```

    cpi_order_data = run(microbench, function=["matrix_row_major", "matrix_
    arguments=arg_map(size=32*1024*1024, reps=1),
    run_options=arg_map(MHz=3500),
    perf_counters=["PERF_COUNT_HW_INSTRUCTIONS", "PERF

```

In [ ]:

```

#data_cell
cpi_order_df = PE_calc(cpi_order_data.as_df())
display(cpi_order_df)
plotPEBar(df=cpi_order_df, what=[ ('function', "IC"), ("function", "CPI

```

### Question 23 (Completeness)

Calculate the speedup of `matrix_row_major` over `matrix_column_major`. Why is this result surprising?

Speedup:

Why is this surprising?:

**Interesting Question:** Why does the order in which the program performs calculations affect CPI ?



## 12 Amdahl's Law

Recall from CSE142 that Amdahl's Law limits the speed up an optimization can provide. It's given as

$$S_{\text{tot}} = \frac{1}{\left(\frac{x}{S}\right) + (1-x)}$$

Where  $S$  is the speedup provided by the optimization,  $x$  is the fraction of execution time affected by the optimization, and  $S_{\text{tot}}$  is total speedup. There is also a more general form of Amdahl's law that covers multiple optimizations operating on two *disjoint* portions of a

program:

$$S_{\text{tot}} = \frac{1}{\left(\frac{x_1}{S_1}\right) + \left(\frac{x_2}{S_2}\right) + (1-x_1-x_2)}$$

It can be generalized to handle any number of optimizations operating on disjoint portions of the program.

We are going to explore both of these using functions similar to those we studied above. The code is in `amdahl.cpp`. Take a moment to compare them to the code we saw earlier in `microbench.cpp` and understand what's different.

In [175]:

```
amdahl = build("amdahl.cpp")  
amdahl[0].source()
```

100%

1/1 [00:00&lt;00:00, 66.46it/s]

```

Out[175]: #include <cstdlib>
#include <unistd.h>
#include <unordered_set>
#include<algorithm>
#include<cstdint>
#include"cfiddle.hpp"

#define CLINK extern "C"
#define OPT(a) __attribute__((a))

extern "C" void init(uint64_t * array, uint64_t size) {
    for(uint i= 0 ; i < size; i++) {
        array[i] = 0;
    }
}

extern "C" void baseline_int(uint64_t * array, uint64_t size) {
    for (uint j = 0; j < 3; j++) {
        for(uint i= 1 ; i < size; i++) {
            array[i] += i/(1+j)+array[i - 1];
        }
    }
}

extern "C" void __attribute__((optimize(4))) baseline_int_04 (uint64_t
* array, uint64_t size) {
    for (uint j = 0; j < 3; j++) {
        for(uint i= 1 ; i < size; i++) {
            array[i] += i/(1+j)+array[i - 1];
        }
    }
}

volatile int ROW_SIZE = 1024;
extern "C" void __attribute__((optimize(4))) matrix_column_major(uint6
4_t * array, uint64_t size) {

    for (int k = 0; k < ROW_SIZE; k++) {
        for(uint i= 0 ; i < size/ROW_SIZE; i++) {
            array[i*ROW_SIZE + k] += 1.0; // This Line
        }
    }
}

extern "C" void __attribute__((optimize(4))) matrix_row_major(uint64_t
* array, uint64_t size) {

    for(uint i= 0; i < size/ROW_SIZE; i++) {
        for (int k = 0; k < ROW_SIZE; k++) {
            array[i*ROW_SIZE + k] += 1.0; // This Line
        }
    }
}

```

```
}
}
```

The cell below contains the source for a function that calls several of these functions in sequence and has them operate on the same array of data.

`start_measurement()`, `restart_measurement()`, and `end_measurement()` break up the function into sections that we'll measure independently. The argument to `start_measurement()` and `restart_measurement()` are a "tag" to label those phases in the output.

The cell will compile the code and then run it.

In [ ]:

```
amdahl = build(code(r"""
#include"cfiddle.hpp"
extern "C" void init(uint64_t * array, uint64_t size);
extern "C" void baseline_int(uint64_t * array, uint64_t size);
extern "C" void baseline_int_04 (uint64_t * array, uint64_t size);
extern "C" void matrix_column_major(uint64_t * array, uint64_t size);
extern "C" void matrix_row_major(uint64_t * array, uint64_t size);

extern "C" uint64_t *__attribute__((optimize(0))) everything(unsigned
    uint64_t * array = new uint64_t[size];

    start_measurement("init phase");

    init(array, size);

    restart_measurement("matrix phase");

    matrix_column_major(array, size);

    restart_measurement("baseline phase");

    baseline_int(array,size);

    end_measurement();

    return array;
}
"""), build_parameters=arg_map(MORE_SRC="amdahl.cpp"))

amdahl_data = run(amdahl, function="everything",
    arguments=arg_map(size=32*1024*1024),
    run_options=arg_map(MHz=3500),
    perf_counters=["PERF_COUNT_HW_INSTRUCTIONS", "PERF_COUNT_
```



In [ ]:

```
amdahl_df = PE_calc(amdahl_data.as_df())  
display(amdahl_df)
```

Imagine that you are a manager and your team is tasked with the improving the performance of `everything()` . Members of your team propose two different approaches:

1. Option 1: Replacing `baseline_int()` with `baseline_int_04()`
2. Option 2: Replacing `matrix_column_major()` with `matrix_row_major()` .

To answer the question below, it'll be helpful to have some of the data from our earlier experiments:

In [ ]:

```
display(cpi_order_df)  
display(cpi_opt_df)
```

### Question 24 (Completeness)

**Based on the data you've collected, calculate the two speedups below and then decide which of the two options will give the best overall speedup for `everything()` (Show your work).**

**Speedup of `baseline_int_04()` vs `baseline_int()` :**

**Speedup of `matrix_row_major()` vs `matrix_column_major()` :**

**Which of the two options will give the best speedup?**

### Question 25 (Correctness - 4pts)

**Based on the data you collected earlier in this lab for the performance of `baseline_int()` , `baseline_int_04()` , `matrix_column_major()` , and `matrix_row_major()` , use Amdahl's law to predict the speedup of each approach described below. (Show your work, including the values of  $x$  and  $S$  and how you computed them.)**

**Option 1 (Replacing `baseline_int()` with `baseline_int_04()` )**  
**Speedup:**

x =

S = (Hint: you computed this in the previous question)

S\_tot =

**Option 2 (Replacing `matrix_column_major()` with `matrix_row_major()` ) Speedup:**

x =

S = (Hint: you computed this in the previous question)

S\_tot =

**Option 3 (Replace both) Speedup:**

This one is more challenging. Show your work.

Modify the code for `everything()` above to implement all three versions (Note: it might be wise to save a copy of the original code). Store the results for each experiments in different variables (e.g., `opt_baseline` instead of `amdahl_data` ) and use the cell below to print them out `display()` out. Make it clear either through comments in the code below or the data itself which version is which:

In [ ]:

```
# display(your_data)
```

### Question 26 (Correctness - 2pts)

**What was the actual speedup for each option? Did Amdahl's law get it right?**

**Option 1 speedup:**

**Option 2 speedup:**

**Option 3 speedup:**

**Did Amdahl's Law get it right?:**

In [ ]:



## 13 Recap

This lab has collected real data to understand the how the performance equation, the power equation, and Amdahl's law apply to some simple programs. This exploration presented the following questions:

- Why does increasing the size of array change CPI ? And why does this change occur so quickly?
- How can clock rate affect CPI ?
- Why doesn't clock rate always affect CPI ?
- How and why do the datatypes we use change IC and CPI ?
- Why does the order in which the program performs calculations affect CPI ?

Throughout the rest of the course and the labs, we'll find answers to most of these and see how can use those answers to make better use of modern processors.

### *Question 27 (Completeness)*

**Which of these questions do you find most interesting and why?**

### *Question 28 (Completeness)*

**Give three other questions you have after completing this lab.**

1. question 1
2. question 2
3. question 3

### *Question 29 (Optional)*

**What's the seed for the first hidden test case? This one is pretty hard.**

## 14 Turning In the Lab

For each lab, there are two different assignments on gradescope:

1. The lab notebook.
2. The programming assignment.

There's also a pre-lab reading quiz on Canvas and a post-lab survey which is embedded below.

### 14.1 The CSE142L Emergency Lab Submission Form

We do not accept late submissions. However, sometimes things go wrong at submission time. To accommodate this, we have the [Emergency Lab Submission Form](https://docs.google.com/forms/d/e/1FAIpQLScTowUZOFsQ6vp36PEKFWeyOEpN_FrEwo8pIKzv/) ([https://docs.google.com/forms/d/e/1FAIpQLScTowUZOFsQ6vp36PEKFWeyOEpN\\_FrEwo8pIKzv/](https://docs.google.com/forms/d/e/1FAIpQLScTowUZOFsQ6vp36PEKFWeyOEpN_FrEwo8pIKzv/)). It allows us to deal with submission problems in a fair and uniform way.

Here's the process:

1. If you are having trouble submitting, commit your work, and fill out this form *before the deadline*. THERE WILL BE NO EXCEPTIONS GRANTED.
2. The commit has you provide for your github repo must be dated before the deadline.
3. You can continue to try to submit via the normal gradescope.
4. If you aren't able to successfully submit via gradescope, then submit a regrade request during the regrade period.
5. We will review the contents of your github repo, the gradescope submission URLs, and the job IDs you provide.
6. If there was some problem with the infrastructure, you can receive up to full credit. If there was a problem on your side (e.g., not generating the PDF properly), you can earn up to 90% credit.

We will not address these issues on Piazza or via email.

### 14.2 Reading Quiz

The reading quiz is an online assignment on Canvas. It's due before the class when we will assign the lab.

### 14.3 The Note Book

You need to turn in your lab notebook and your programming assignment separately.

After you complete the lab, you will turn it in by creating a version of the notebook that only contains your answers and then printing that to a pdf.

**Step 1:** Save your workbook!!!

```
In [ ]: !for i in 1 2 3 4 5; do echo Save your notebook!; sleep 1; done
```

**Step 2:** Run this command:

```
In [ ]: !turnin-lab Lab.ipynb  
!ls -lh Lab.turnin.ipynb
```

The date in the above file listing should show that you just created `Lab.turnin.ipynb`

**Step 3:** Click on this link to open it: [./Lab.turnin.ipynb](#) ([./Lab.turnin.ipynb](#))

**Step 4:** Hide the table of contents by clicking the



**Step 5:** Select "Print" from *your browser's* "file" menu. Print directly to a PDF.

**Step 6:** Make sure all your answers are visible and not cut off the side of the page.

**Step 7:** Turn in that PDF via gradescope.

**Print Carefully** It's important that you print directly to a PDF. In particular, you should *not* do any of the following:

1. **Do not** select "Print Preview" and then print that. (Remarkably, this is not the same as printing directly, so it's not clear what it is a preview of)
2. **Do not** select "Download as-> PDF via LaTeX". It generates nothing useful.

Once you have your PDF, you can submit it via gradescope. In gradescope, you'll need to show us where all your answers are. Please do this carefully, if we can't find your answer, we can't grade it.

## ▼ 14.4 Lab Survey

Please fill out this survey when you've finished the lab. You can only submit once. Be sure to press "submit", your answers won't be saved in the notebook.

In [176]:

```
from IPython.display import IFrame
IFrame('https://docs.google.com/forms/d/e/1FAIpQLScwdcKJcNbleIfp7upzmkk')
```

Out[176]:

## CSE142L Lab Survey

sjswanson@ucsd.edu [Switch account](#)

\* Indicates required question

Email \*

☐

Record sjswanson@ucsd.edu as the email to be included with my responses