

GPHY491/489: Programming for GIS

K. Arthur Endsley

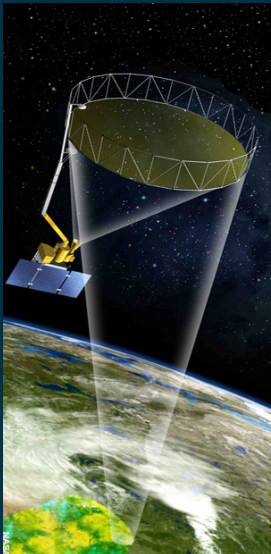
Numerical Terradynamic Simulation Group

W.A. Franke College of Forestry and Conservation

January 16, 2025



Introductions



Hi, My Name is Arthur!

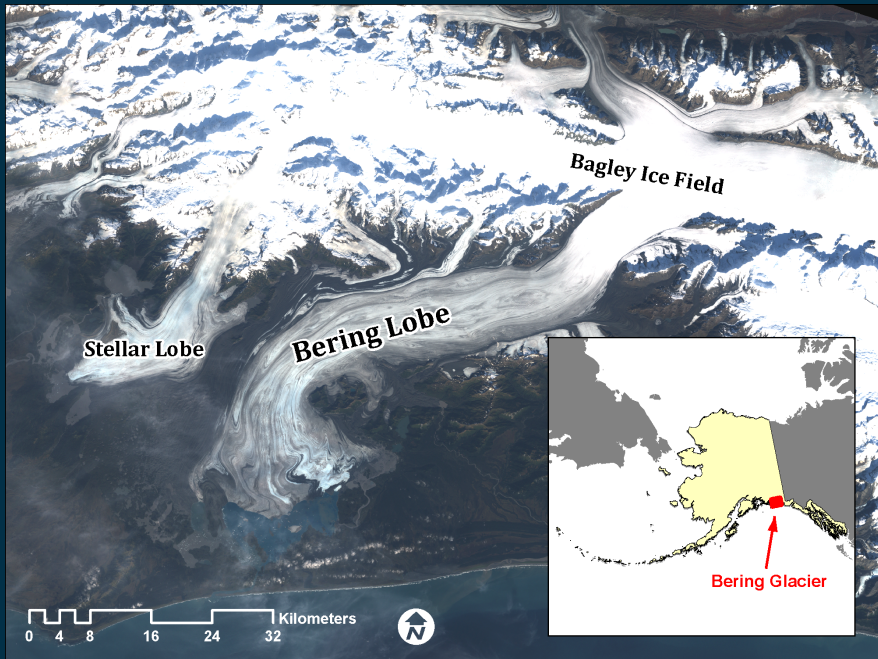
- Taught programming at Lawrence Berkeley National Labs, NASA Langley, the Federal Reserve Board...
- Maintaining the **NASA Soil Moisture Active Passive (SMAP) satellite** mission's Level 4 Carbon (L4C) model: 1-km resolution, global coverage, over 500 million pixels!
- Leading NASA-funded research into how satellite microwave measurements can be used to study forest water stress and fire risk.

Introductions

Teaching Assistant: Bryan Tutt

Outline

- GIS programming workflows: An example
- Some role-playing: How can programming improve GIS tasks?
- Thinking like a Computer Scientist
- Course overview



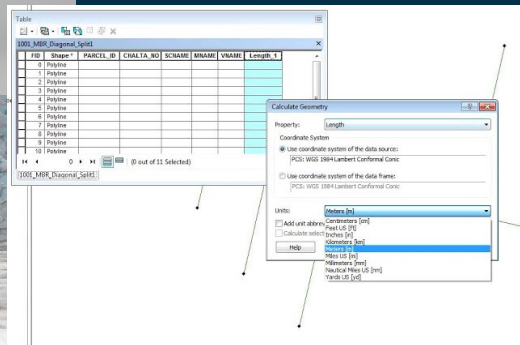
Motivation: Ice Motion Tracking



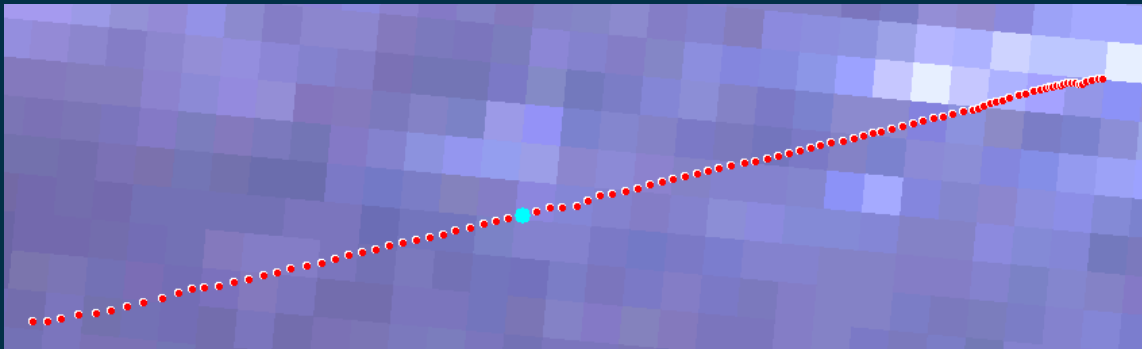
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```
import scipy
scipy.signal.filtfilt(1, 2, gps_coords)
```

Live-Action Role-Playing: GIS Edition



Pair up! There are two roles: **Manager and Analyst**. Take 5 minutes in one role for Prompt A, then **switch roles** and read Prompt B. The Analyst should describe how they would achieve the goal using ArcGIS Desktop.

Questions to consider:

- How will the data be handled?
- What specific tools would you use, in what order?
- How long will it take?
- What are the possible sources of error?

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- Computer code is transferable and re-useable; it can be used to verify that your analysis was done correctly and to obtain the same result.

So: Why Programming?

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- A computer program or script is a documentation of your workflow.
- Computer code is transferable and re-useable; it can be used to verify that your analysis was done correctly and to obtain the same result.
- **\$\$\$:** U.S. average annual salary (2023):¹
 - “GIS Analyst:” \$72,530
 - “GIS Programmer:” \$86,743

¹ZipRecruiter.com

Thinking Like a Computer Scientist

Computational thinking is a *problem-solving* activity.¹

¹Cynthia Selby & John Woolard (2013)

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Abstraction: Representing only what is essential.

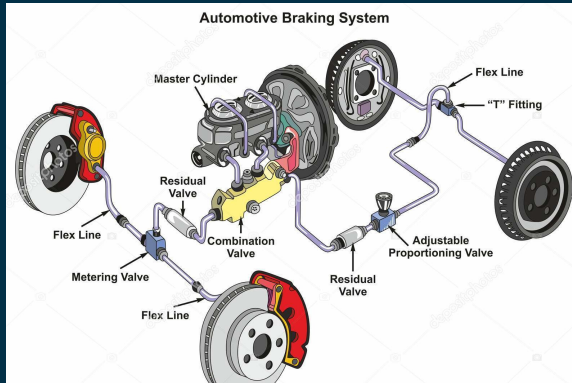


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Computational thinking is a *problem-solving* activity.

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Decomposition: Decomposing a complex problem or system into manageable parts.



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Algorithms: Logical and ordered instructions for carrying out a task.

- Order and precedence; what is required before the next thing?
- Design the fewest number of steps; remove unnecessary steps

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Debugging and Continuous Improvement

- Detect and identify errors
- Start with an initial, acceptable solution
- Then iteratively refine the solution, as needed

Valerie Shute et al. (2017)



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Debugging and Continuous Improvement

Collaboration, Reflection, and Feedback

Valerie Shute et al. (2017)



Thinking Like a Computer Scientist (2)

Don't be afraid to experiment!

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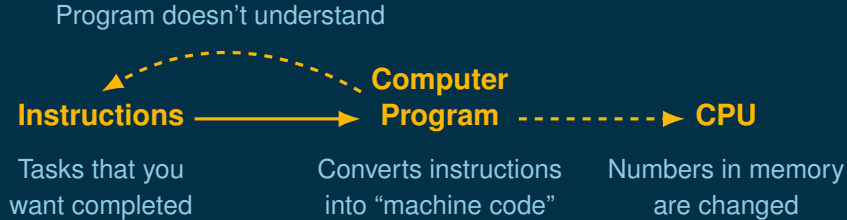
You should be asking yourself:

```
pyplot.scatter(data)
```

```
array.min()
```

- “Is there a `pyplot.histogram()` or `pyplot.lineplot()` function?”
- “Will `array.max()` calculate the maximum?”

Thinking Like a Computer Scientist (3)



(CPU: Central Processing Unit)

Thinking Like a Computer Scientist (4)



Assembly Language

DOSSEG

.MODEL TINY

.DATA

TXT DB "Hello, world!\$"

.CODE

START:

MOV ax, @DATA

MOV ds, ax

MOV ah, 09h

MOV dx, OFFSET TXT

INT 21h

MOV AX, 4C00h

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END START

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2nd-Generation Language (C)

```
include <stdlib.h>
include <stdio.h>

int main(void)
{
    printf("Hello, world!\n");
    return EXIT_SUCCESS;
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3rd-Generation Language (Python)

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print("Hello, world!\n")
```

Thinking Like a Computer Scientist (4)

Compiled Languages

e.g., C, Java, Fortran

- Code is *compiled* into machine code (binary) *before* the program is executed

Interpreted Languages

e.g., Python, R

- Code is translated into machine code automatically when the program is run

Thinking Like a Computer Scientist (4)

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- **Never as fast: Programs are executed as low-level instructions**

Thinking Like a Computer Scientist (4)

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- Programmer must specify data types, how to allocate storage (memory), ...
- Compiled programs can be very fast: Programs executed as binary
- Can take a long time to write and debug a program

Interpreted Languages

e.g., Python, R

- Code is translated into machine code automatically when the program is run
- Interpreter figures out what data types and memory are required
- Never as fast: Programs are executed as low-level instructions
- Much easier to learn; programs can be written quickly

Who is a Computer Programmer?

Who is a Computer Programmer?



Hedy Lamarr (1914-2000)

- Invented frequency-hopping technology, initially for radio-guided torpedoes
- Same technology later used in cellular phones
- **Self-taught!** Collaborated with friend George Antheil to use a player-piano reel for timing the changes in frequency

Who is a Computer Programmer? (2)

**The most famous faces in the tech industry tend to look very similar...
and they're making the world less fair and less free.**

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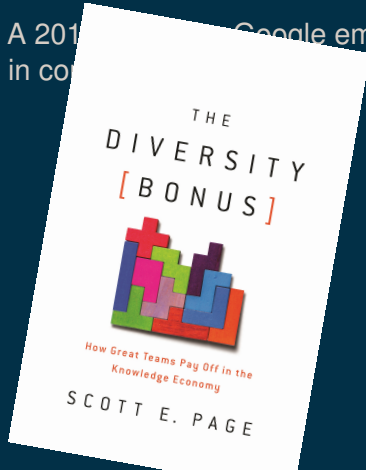
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A 2017 memo by Google employee James Damore claimed that participation and success in computer science is *biologically determined*.

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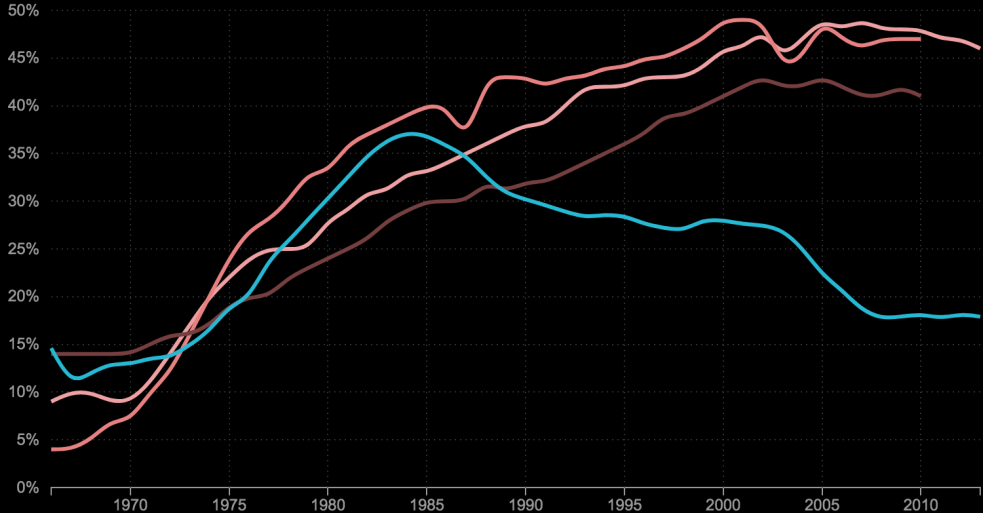


“As the initial pool of problem solvers becomes large, the best-performing [programmers] necessarily become similar... Their relatively greater ability is *more than offset* by their lack of problem-solving diversity.”

What Happened To Women In Computer Science?

% Of Women Majors, By Field

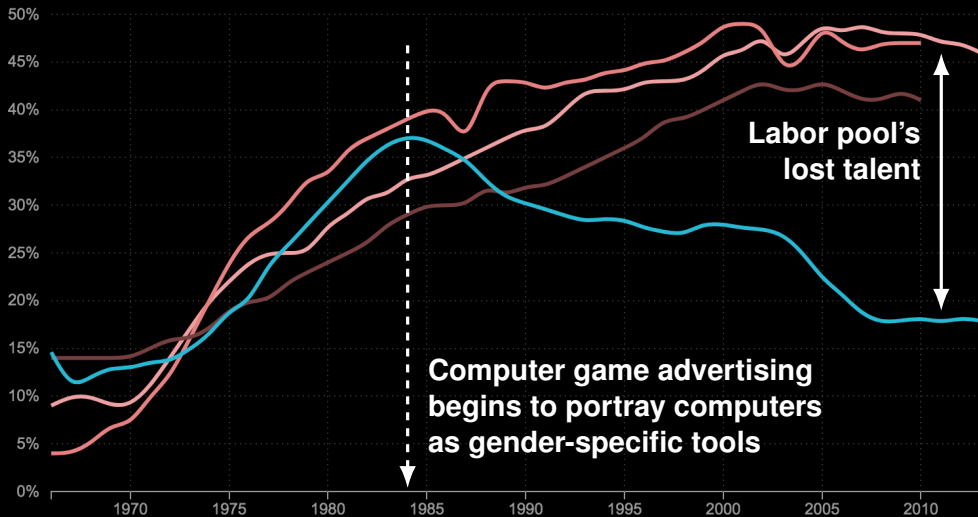
Medical School Law School Physical Sciences Computer science



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**This is a challenging course, but you already have
all the tools you need to be successful.**

Course Overview

How You Will Learn

We'll use two learning strategies in this course:

- 1 Hands-on-Keyboards** with real data
- 2 Peer Programming** with real data

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It's not Chemistry, it's Carpentry!

What You Will Learn

- **Python programming** (currently the most popular language in the world¹)
- **R for data analysis**

¹<https://www.tiobe.com/tiobe-index/>

What You Will Learn (2)

How to choose the best language for the job:

Python

- **Higher performance**
- **Raster and array data processing**
- Batch processing of multiple files
- Machine learning
- **Plotting large raster datasets**
- Process-based or multi-criteria modeling (e.g., habitat suitability)
- Cellular automata/ Agent-based models
- Creating your own algorithm

R

- **Vector data analysis**
- Working with an Attribute Table
- **Plotting vector data and coarse-resolution raster data**
- Geostatistics
- Spatial point pattern analysis
- Spatial autoregressive models
- Using algorithms written by others

How You Will Be Evaluated

Lab Exercises: 60%

For graduate students or for undergraduate extra credit:

Midterm Project: 20%

Final Project: 20%

For Next Time

Make sure to sign up for *both* GPHY 491 and GPHY 489 (Lab).

Lab meets on Wednesdays at 9:00a in Stone Hall 106!

Please listen and read!

- NPR Planet Money Podcast: **“When Women Stopped Coding”**
- Dynamic Ecology blog post: **“Stereotype threat: A summary of the problem”**