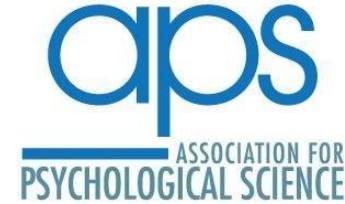




# APS Workshop: Introduction to Python

## San Francisco, CA, 24 May 2018



# Overview

Christian C. Luhmann  
Stony Brook University

# Where to Find these Slides

**[github.com/cluhmann/python-psych-workshop](https://github.com/cluhmann/python-psych-workshop)**

# Who am I?

- B.S. in Computer Science
- Ph.D. in Psychology
- Stony Brook University
- Decision-making, learning, methods (stats & “cognitive modeling”)
- Using Python since ~2002

# Who are You?

- Faculty/students?
- Who has used...
  - Matlab?
  - R?
  - Some other programming language (e.g., Java, C)?
  - SPSS?
  - Eprime?
  - SAS?

# Goals

- Appreciation of the **ends**
  - benefits of Python
  - functionality provided by Python and its ecosystem
  - how to integrate these tools into your existing workflow
- Non-goals of this workshop: **means**
  - Ability to program Python without further consultation
  - Encyclopedic knowledge of packages, APIs, etc.
- Think of this as a open house
  - If you'd like buy, you still need to move all your stuff

# **What I will assume of you...**

- Not much
- You're not terrified of programming
- You use data in your research
- You're looking for tools to conduct efficient, flexible, reproducible (maybe sharable) analyses
- You (maybe) conduct laboratory experiments

# Why?

- Why bother to learn another thing?
  - We already have Matlab, R, etc.
- Why Python?
- Python...
  - is general-purpose
  - is free and open source
  - is eminently readable (i.e., readily learned)
  - has an extensive, well-integrated ecosystem of tools
  - and more!
- This workshop, hopefully, is a more comprehensive answer

# What is Python?

- Developed by Guido van Rossum in the early 1990s
- Python 2.0 was released October 16<sup>th</sup>, 2000
- Python 3.0 was released December 3<sup>rd</sup>, 2008

# Python

- Free and open source
- Cross-platform
- Widely-used and well-supported
- Well-documented
- Multiple options for boosting performance
- **Highly readable**
- Substantial **standard library**
- Vibrant **third-party ecosystem**

# Standard Library

```
>>> abs( -42 )  
42
```

```
>>> pow( 2, 10 )  
1024
```

# Standard Library

```
>>> min([1, 4, 12, 42])  
1
```

```
>>> max([1, 4, 12, 42])  
42
```

```
>>> len([1, 4, 12, 42])  
4
```

```
>>> sum([1, 3, 5])  
9
```

```
>>> sorted([2, 4, 6, 8, 1, 3, 5, 7, 9])  
[1, 2, 3, 4, 5, 6, 7, 8, 9]
```

# Standard Library

```
>>> print('Six times nine is ' + str(6*9))
Six times nine is 54
```

```
>>> file = open('myfile.txt', 'r')
>>> contents = file.read()
>>> print(contents)
First line of my file.
Second line of my file.
Last line of my file.
```

# Python's Ecosystem

Many more...

ggplot

bambi

scikit-learn

statsmodels

seaborn

pymc3

scipy

pandas

matplotlib

numpy

python

# Installing Python

- Anaconda
- Enthought's Canopy
- WinPython (Windows only)
- Each of these projects provides:
  - Python
  - Packages
  - Package manager
  - Editor (IDE)
  - Other tools

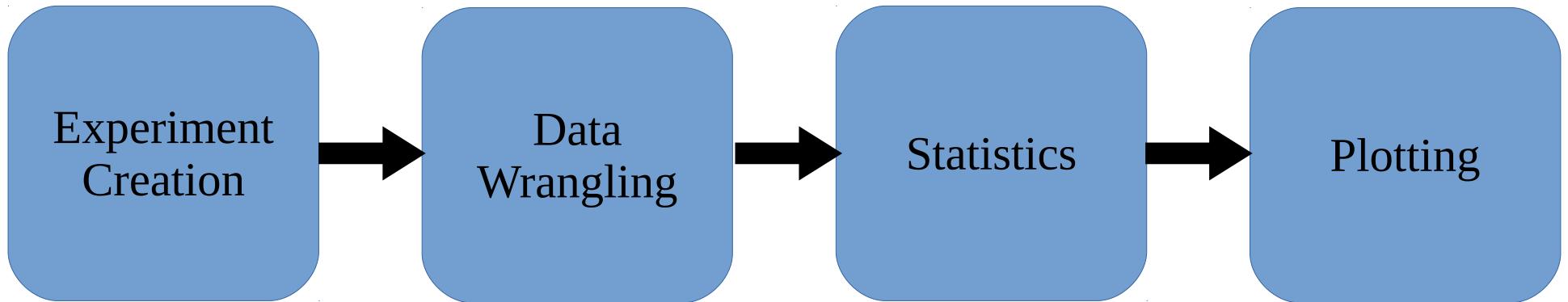
# Anaconda

[www.anaconda.com/download](http://www.anaconda.com/download)

# Installing Python

- Python 2.x or 3.x?
- Python 2.7's end-of-life initially 2015, but postponed to 2020
  - concern that much existing code could not easily be ported to Python 3
- Python 3.x is recommended

# The Pipeline



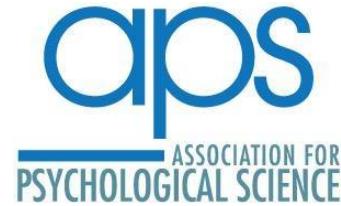
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# APS Workshop: Introduction to Python

## San Francisco, CA, 24 May 2018



# Using Python

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Stony Brook University

# Python

- How do I **do** Python?
- There are **three ways** we will cover today...

# Console

Useful when...

- Playing with very brief pieces of code (i.e., one line at a time)
- No need to save code (or results) for later
- Trying things out
- Learning Python (exploring, trial & error)

Let's try out the console

# Scripts

Useful when...

- Executing large amounts of code
- Confident that your code works the way you want
- No need to “oversee” the code’s operation

Let's see a script

# Notebooks

Useful when...

- You want to play with code
- You want a record of what you do and results (mnemonic)
- You might want to share that record with others
- Teaching

Let's try out a notebook

# Notebooks

- Notebooks may be exported in a variety of formats, including...
  - PDF
  - LaTex
  - HTML
  - Python script
- Useful for supplying others (possibly non-programmers) with information about what you have done/found
  - Students/advisors
  - Collaborators
  - Readers
    - [theatlantic.com/science/archive/2018/04/the-scientific-paper-is-obsolete/556676/](https://theatlantic.com/science/archive/2018/04/the-scientific-paper-is-obsolete/556676/)

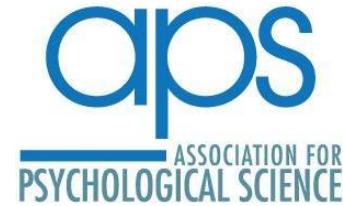
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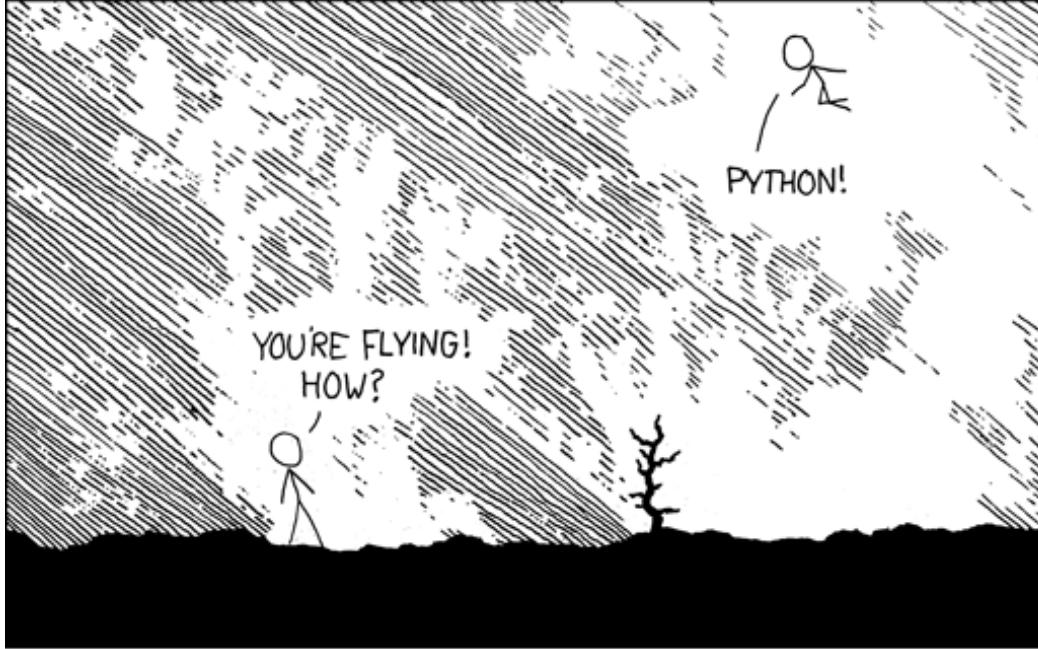
# APS Workshop: Introduction to Python

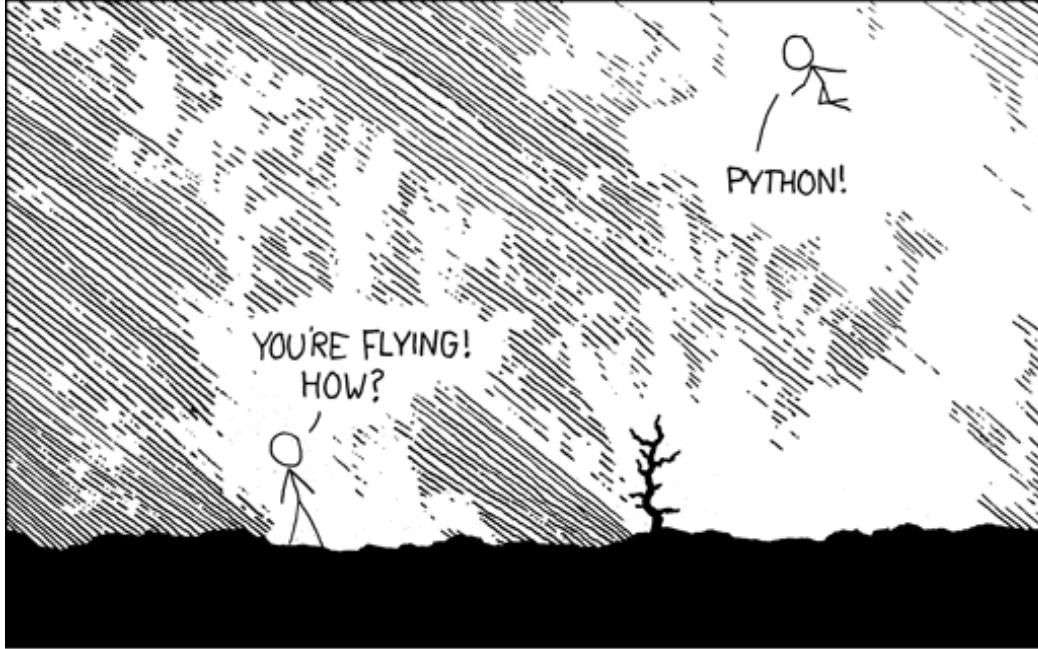
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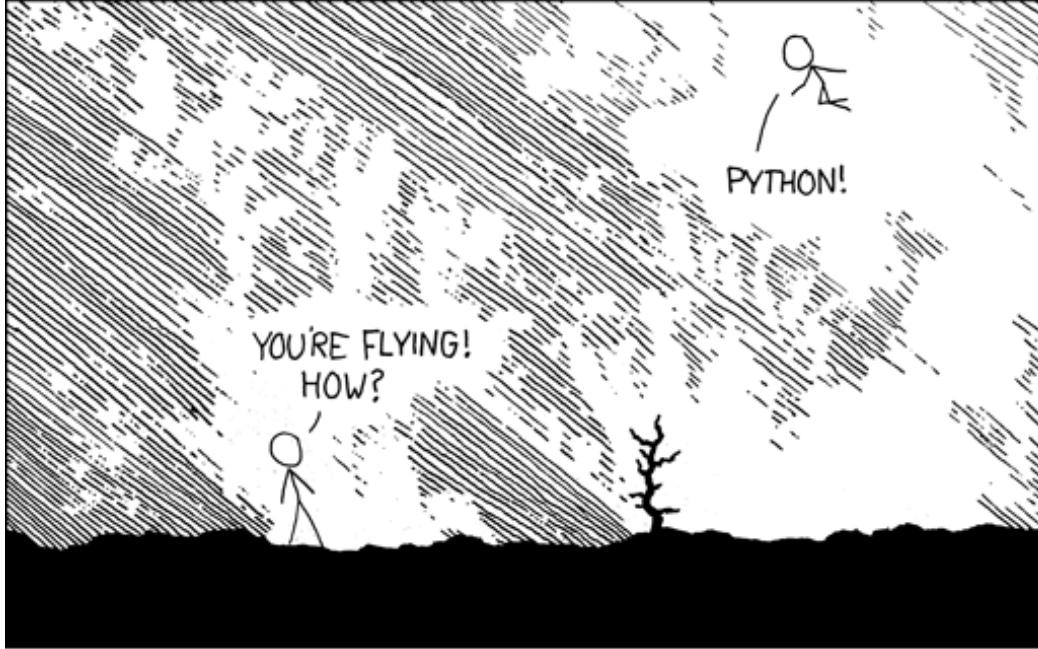


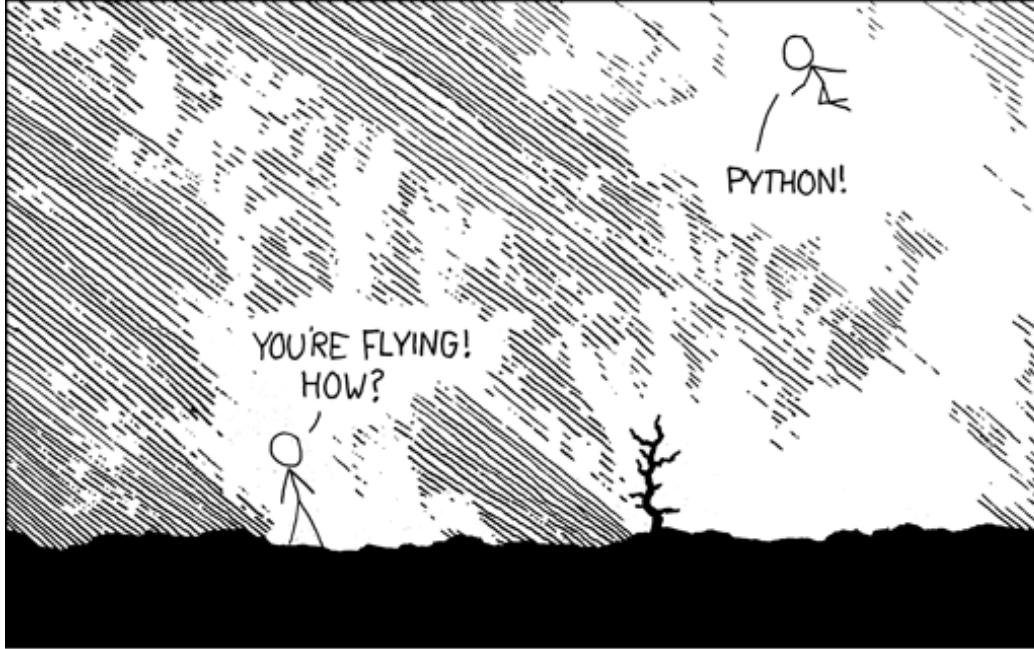
# Python Basics

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Stony Brook University









I LEARNED IT LAST NIGHT! EVERYTHING IS SO SIMPLE!

HELLO WORLD IS JUST

`print "Hello, world!"`

I DUNNO...  
DYNAMIC TYPING?  
WHITESPACE?

COME JOIN US!  
PROGRAMMING IS FUN AGAIN!  
IT'S A WHOLE NEW WORLD UP HERE!

BUT HOW ARE YOU FLYING?

I JUST TYPED  
`import antigravity`

THAT'S IT?

... I ALSO SAMPLED EVERYTHING IN THE MEDICINE CABINET FOR COMPARISON.

BUT I THINK THIS IS THE PYTHON.

```
>>> import this
```

```
The Zen of Python, by Tim Peters
```

Beautiful is better than ugly.

Explicit is better than implicit.

Simple is better than complex.

Complex is better than complicated.

Flat is better than nested.

Sparse is better than dense.

Readability counts.

Special cases aren't special enough to break the rules.

Although practicality beats purity.

Errors should never pass silently.

Unless explicitly silenced.

In the face of ambiguity, refuse the temptation to guess.

There should be one-- and preferably only one --obvious way to do it.

Although that way may not be obvious at first unless you're Dutch.

Now is better than never.

Although never is often better than \*right\* now.

If the implementation is hard to explain, it's a bad idea.

If the implementation is easy to explain, it may be a good idea.

Namespaces are one honking great idea -- let's do more of those!

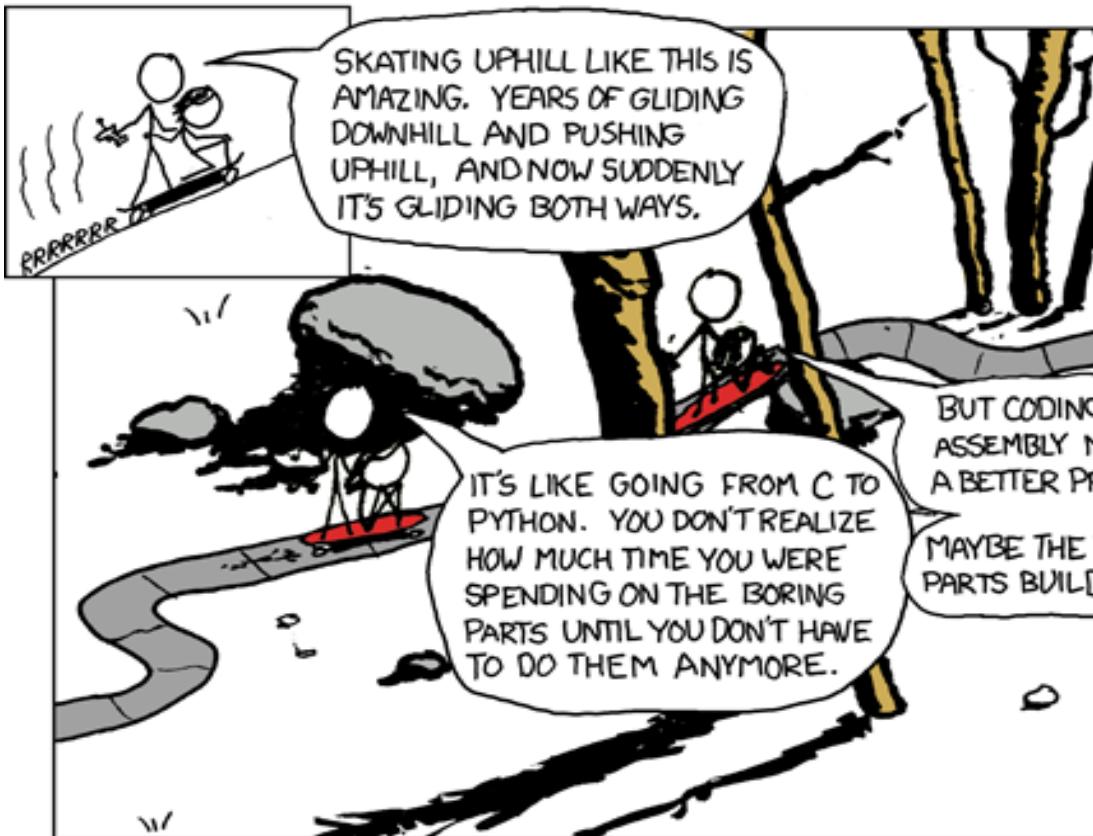
# **Python**

So let's take a closer look at the language itself

# Readability counts

- Every programming language represents a tradeoff between...
  - time it takes **to write** a program (your time)
  - time it takes **to run** a program (computer's time)
- Python prioritizes **your** time

# Python



# Python

For researchers, this means...

- students are less intimidated
- collaboration is easier
- improved transparency

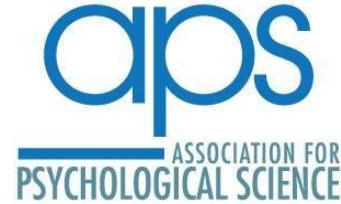
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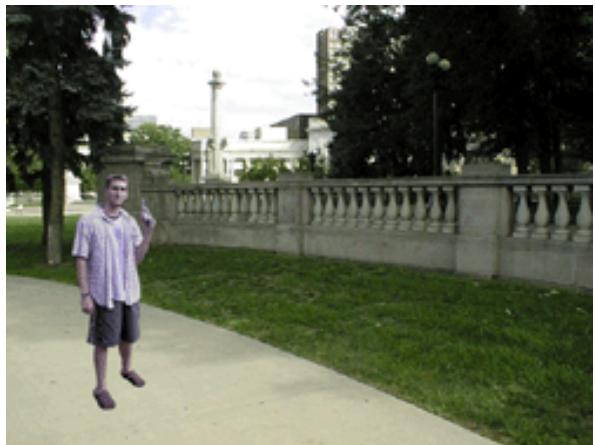


# Our Dataset

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# Dataset

- Joshua Correll's **police officer's dilemma** task
  - [psych.colorado.edu/~jclab/FPST.html](http://psych.colorado.edu/~jclab/FPST.html)
- We will build this task later on (twice!)
- In the meantime, let's follow a data set through our pipeline



Task: to shoot or to not shoot

# Design

		Race	
		White	Black
Object	Gun		
	Not gun		

# Dataset

- We will present multiple images per “cell” of this design
- We will collect both responses (accuracy) and reaction times
- At the beginning of the experiment, we will ask for subjects’ ages

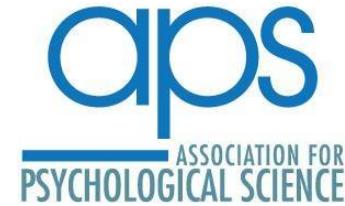
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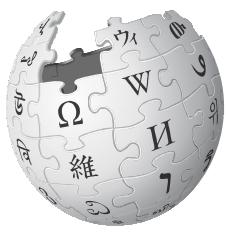
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# Data Wrangling

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# Wrangling

- Data wrangling, sometimes referred to as data munging, is the process of transforming and mapping data from one “raw” data form into another format with the intent of making it more appropriate and valuable for a variety of downstream purposes such as analytics.
- This may include further munging, data visualization, data aggregation, training a statistical model, as well as many other potential uses.

# Relevant Packages

- numpy
- pandas
- matplotlib

# Relevant Packages

- numpy
  - Matrix representation
  - Linear algebra
  - Fast

# numpy

4.1	3.4	2.6
12.6	8.1	1.2
6.2	10.4	5.8

`dtype = float`

# numpy

4	3	2
12	8	1
6	10	5

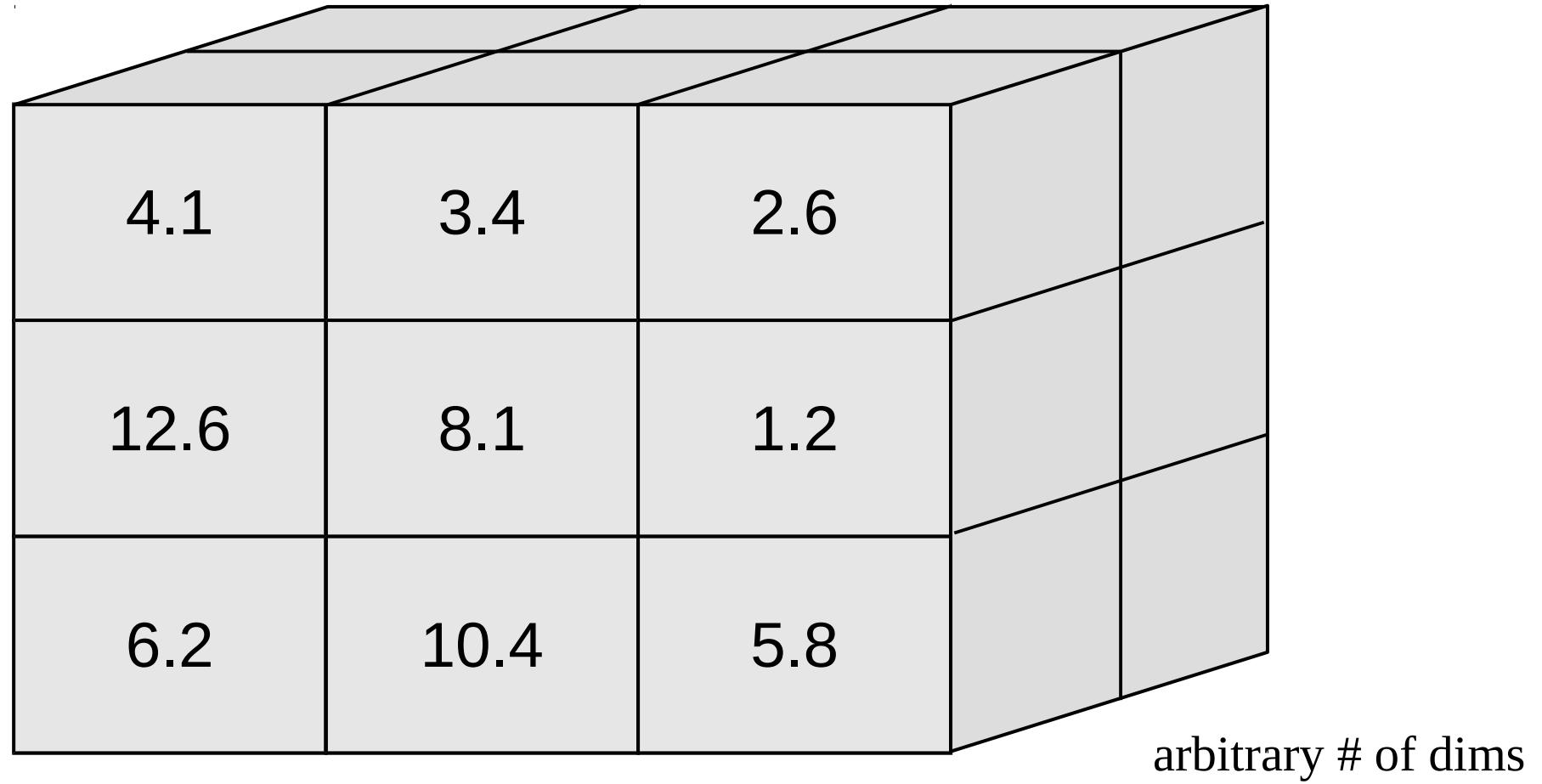
dtype = int32

# numpy

4+9j	3+6j	2+4j
12+4j	8+9j	1+8j
6+8j	10+6j	5+2j

`dtype = complex`

# numpy



# Relevant Packages

- numpy
  - Matrix representation
  - Linear algebra
  - Fast
- pandas
  - R-style dataframe
  - Best for a mixture of heterogenous data types (e.g., subject #, name, DOB)

# Dataframe

# Dataframe

	A	B	C	D	E	
1	Stock Name	Symbol	Shares	Purchase Price	Cost Basis	Curr
2	Apple	AAPL	100	\$90.00	\$9,000.00	
3	Microsoft	MSFT	200	\$32.00	\$6,400.00	
4	Salesforce	CRM	150	\$25.00	\$3,750.00	
5	Oracle	ORCL	250	\$50.00	\$12,500.00	
6	Hewlett Packard Enterprise	HPE	500	\$18.00	\$9,000.00	

# Relevant Packages

- numpy
  - Matrix representation
  - Linear algebra
  - Fast
- pandas
  - R-style dataframe
  - Best for a mixture of heterogenous data types (e.g., subject #, name, DOB)
  - Lots of slicing and dicing options
- matplotlib
  - Matlab-style plotting

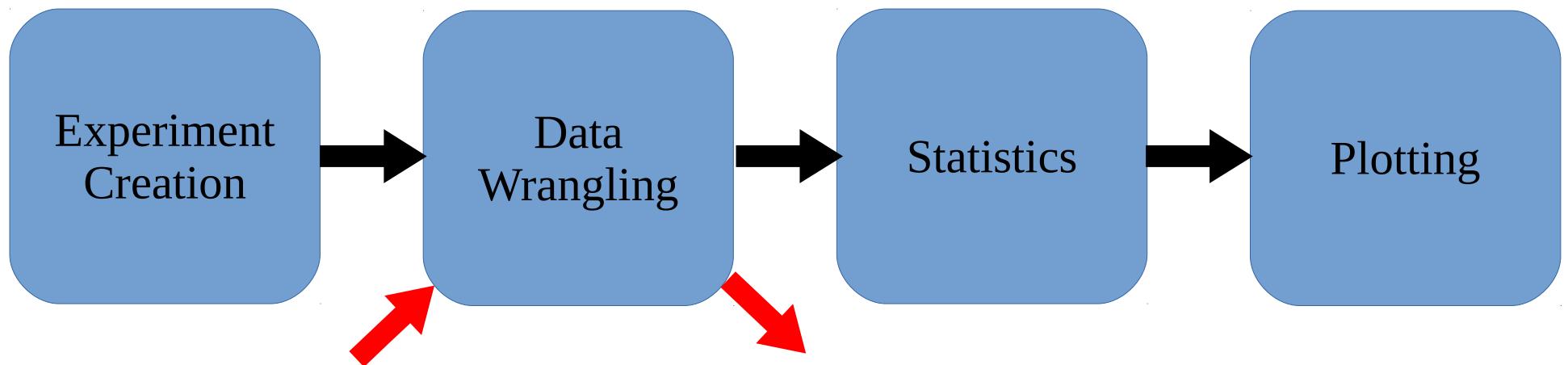
# Wrangling

So let's go wrangle some data

# Pandas

- Pandas can read/write a variety of data formats...
  - CSV
  - JSON
  - HTML
  - Local clipboard
  - MS Excel
  - HDF5 Format
  - Feather Format
  - Parquet Format
  - Msgpack
  - Stata
  - SAS (read only)
  - Python Pickle Format
  - SQL
  - Google Big Query

# The Pipeline



# Take-homes

- You've now seen some **data wrangling** done in Python
- You've seen some of the functionality that **relevant packages** provide
  - pandas
  - jupyter (notebook)
  - matplotlib
- You have some sense of the **flexibility** provided by these tools

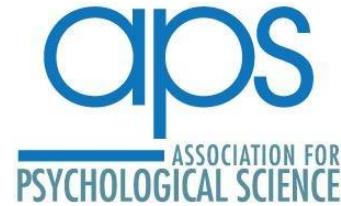
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# Statistics

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# Relevant Packages

- `scipy`
  - Distributions
  - Simple stats (e.g.,  $t$ ,  $\chi^2$ , z,  $r$ , 1-way ANOVA)
- `statsmodels`
- `pymc3`
- `bambi`

# Relevant Packages

- `scipy`
  - Distributions
  - Simple stats (e.g.,  $t$ ,  $\chi^2$ , z,  $r$ , 1-way ANOVA)
- `statsmodels`
  - More complex stats (e.g., GLM, mixed linear models, survival analysis)
- `pymc3`
- `bambi`

# Relevant Packages

- `scipy`
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- `pymc3`
  - Full-featured Bayesian modeling (think Stan)
- `bambi`

# Relevant Packages

- `scipy`
  - Distributions
  - Simple stats (e.g.,  $t$ ,  $\chi^2$ , z,  $r$ , 1-way ANOVA)
- `statsmodels`
  - More complex stats (e.g., GLM, mixed models, survival analysis)
- `pymc3`
  - Full-featured Bayesian modeling (think Stan)
- `bambi`
  - Streamlined, Bayesian GLMs built on top of pymc3 (think brms?)

# **Statistics**

Let's go do some stats!

# scikit-learn

- Machine learning
  - Supervised
    - Classification (e.g., GLM, LDA, SVM, random forests)
    - Regression (e.g., ridge, lasso)
  - Unsupervised
    - Clustering (k-means)
    - Dimension reduction (e.g., PCA)
- All the extras needed to fit, evaluate, and use these tools

# Take-homes

- You have now seen some stats done in Python
- Seen some of the functionality that relevant packages provide
  - pandas
  - jupyter (notebook)
  - matplotlib
- What data exploration looks like and the flexibility these tools provide

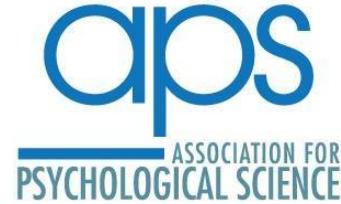
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# Plotting

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# Relevant Packages

- pandas
- matplotlib
  - Matlab-style plotting
  - Extremely flexible, but easiest to generate basic plots
- seaborn
  - pandas-aware plotting
  - Tries to do as much automatically as possible

# Plotting

Let's go plot some stuff!

# Take-homes

- You've now seen some **plotting** done in Python
- You've seen some of the functionality that **relevant packages** provide
  - pandas
  - matplotlib
  - seaborn
- You have some sense of the **flexibility** provided by these tools

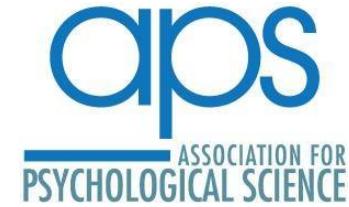
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# Experiment Creation

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Stony Brook University

# Relevant Packages

- **psychopy**
- pyserial
- pyparallel
- pyopengl
- pyglet
- moviepy
- pillow

# PsychoPy

- Originally created by Jon Pierce
- Initiated as a python replacement for Psychtoolbox
  - But it has grown in to much, much more
- It is recommended to install the Psychopy **standalone** alongside any data-centric python installation you might have (e.g., Anaconda)
- The standalone version of PsychoPy includes...
  - Python (2.x or, as of April 2018, 3.x)
  - PsychoPy
  - all the other packages required by PsychoPy

# PsychoPy

- Coder
  - Coder is PsychoPy's IDE (like Spyder)
  - Provides already-written demos and examples
- Builder
  - No programming required
  - Experiments are built using a graphical interface
  - Demos/examples already built
- Can also use PsychoPy as a regular package within python (2.x or 3.x)

# **PsychoPy**

- Running PsychoPy (using the shortcut created during installation) should get this...

# Builder

- Routines
  - Describe timing of stimuli, instructions, responses, etc.
  - A set of events that will always occur together in a fixed order
- Flow
  - Control the way in which the Routines are combined, repeated, and controlled
  - Like a flowchart

# Builder

Let's use Builder to **build** some experiments!

# **Shooter Task – Builder Version**

Specification:

- Show a page of instructions
- Run 10 trials :
  - Show an image
  - Collect a response
  - Show a blank screen (ITI)
- Tell the subject that they're done

# Shooter Task – Experiment Settings

- File->New
- Experiment Settings (blue button to the right of Monitor Center button)
  - Full-screen window: unchecked
  - Window size (pixels): [900,900]
  - Can leave the rest as they are

# Shooter Task – Instructions

- Experiment->New Routine(Shift+Ctrl+N)
  - Name: instructions
- Click on the Text component button (to the right)
  - name: instructText
  - duration: leave blank
  - color: black
  - text: Whatever you want your instructions to say, but ask for a keypress
- Click on the Keyboard component button (to the right)
  - name: instructResp
  - allowedKeys: leave blank
  - store: nothing
  - forceEndTrial: checked
  - storeResponseTime: unchecked

# **Shooter Task – ITI**

- Experiment->New Routine(Shift+Ctrl+N)
  - Name: ITI
- Click on the Text component button (to the right)
  - name: itiText
  - duration: 0.5
  - text: leave blank

# Shooter Task – End of Task

- Experiment->New Routine
  - Name: thanks
- Click on the Text component button (to the right)
  - name: thanksText
  - duration: 10.0
  - color: black
  - text: Whatever you want your instructions to say, but ask for a keypress
- Click on the Keyboard component button (to the right)
  - name: thanksResp
  - allowedKeys: leave blank
  - store: nothing
  - forceEndTrial: checked
  - duration: 10.0
  - storeResponseTime: unchecked

# Shooter Task – Trials

- Experiment->New Routine
  - Name: trial
- Click on the Image component button (to the right)
  - name: trialImg
  - duration: 1.0
  - image: leave blank for now (will change later)
- Click on the Keyboard component button (to the right)
  - name: trialResp
  - allowedKeys: ['1', '2']
  - store: last key
  - storeCorrect: checked
  - correctAns: leave blank for now (will change later)
  - forceEndTrial: checked
  - duration: 2.0
  - storeResponseTime: checked

# **Shooter Task – Flow**

- Insert Routine
  - Select “instructions”
  - Click to the left of “trial”
- Insert Routine
  - Select “thanks”
  - Click to the right of “trial”
- Insert Loop
  - name: trials
  - loopType: random
  - nReps: 1
  - trialListFile: trials.csv

# **trials.csv**

- Comma-delimited text file containing:

img,correctAns

./images/zaba011.jpg,black

./images/zabu02w4.jpg,black

./images/zawa90d3.jpg,white

- First line is the header, telling you what each column is
- Remaining lines describe each trial

# trials.csv

- Need to go back and add in the trial info that will change each time through the loop
- In the trial routine, click on the keyboard component we added earlier
  - Change the correctAns to \$thisTrial.correctAns
- In the trial routine, click on the image component we added earlier
  - Change the image to \$thisTrial.img
  - Change the dropdown box (to the left of image) to “set every repeat”

Asks for the info associated with the “current” line of the trialLoop.csv file. It’s “\$thisTrial” because we called our loop “trials”. If we had named the loop “blocks”, we would want “\$thisBlock.correctAns” instead.

\$thisTrial.correctAns

Asks for the info from the “correctAns” column of the trialLoop.csv file

Same thing here, but now we want to ask for the info from the “img” column.

\$thisTrial.img

# Coder

Let's use Coder to **build** that same experiment!

# Take-homes

- You've now seen an **experiment created** in psychopy (twice!)
  - Builder
  - Code
- You've seen some of the functionality that **psychopy** provides
  - Window creation and management
  - Stimulus creation and manipulation
  - Response collection
  - Timing
  - Much more (other peripherals, logging, staircasing, etc.)
- **Online experiments** are coming to psychopy (stay tuned!)

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# The End(ish)

- Materials available
  - [github.com/cluhmann/python-psych-workshop](https://github.com/cluhmann/python-psych-workshop)
- Thank you
- Questions?
  - Feel free to harass me at APS or contact me later on
  - [christian.luhmann@stonybrook.edu](mailto:christian.luhmann@stonybrook.edu)



BCS1456928