

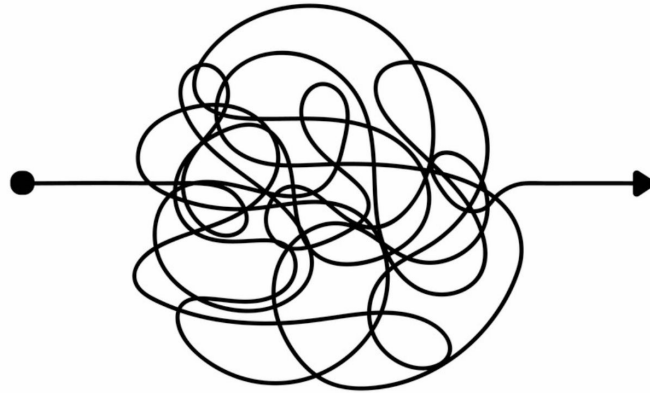
# How Data Scientists Use Computational Notebooks for Real-Time Collaboration

**April Yi Wang** | Anant Mittal | Chris Brooks | Steve Oney

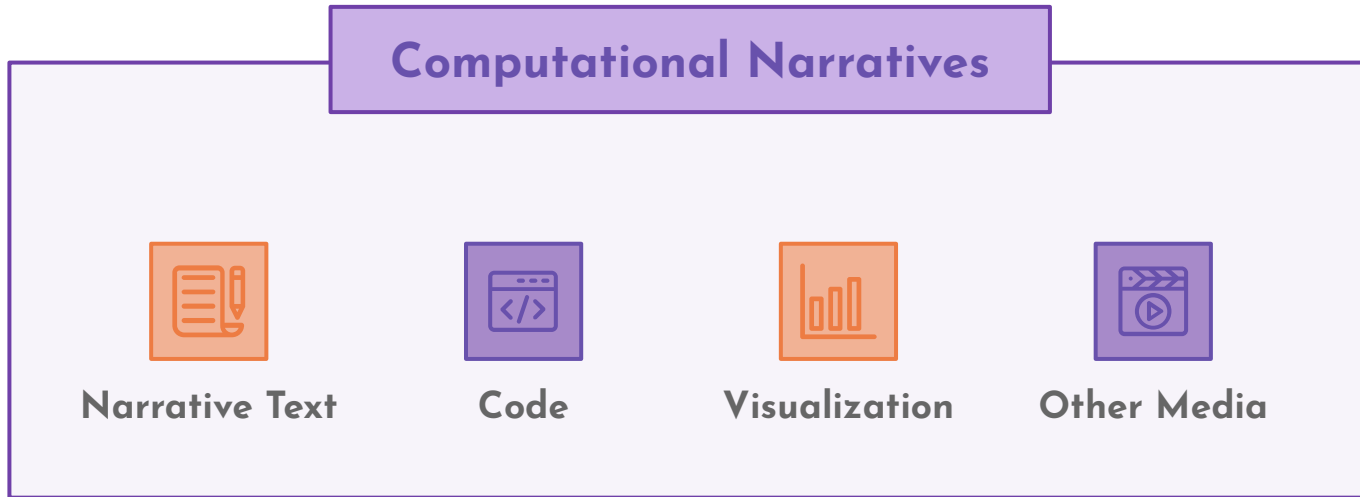
University of Michigan



# The Story Behind Data Analysis



# The Story Behind Data Analysis



Home House Price Prediction Incognito

localhost:8888/notebooks/House%20Price%20Prediction.ipynb

Jupyter House Price Prediction (unsaved changes)

File Edit View Insert Cell Kernel Widgets Help Trusted Python 3

Code nbdiff Share

## House Price Prediction

### Import Library and Dataset

```
In [2]: import numpy as np # linear algebra
import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)
from matplotlib import pyplot as plt
import seaborn as sns

df_train = pd.read_csv('https://researchdatafiles.s3.amazonaws.com/house-d
df_train.describe()
```

Out[2]:

	Id	LotFrontage	LotArea	OverallQual	OverallCond	YearBuilt	YearRemc
count	1460.000000	1201.000000	1460.000000	1460.000000	1460.000000	1460.000000	1460.0
mean	730.500000	70.049958	10516.828082	6.099315	5.575342	1971.267808	1984.8
std	421.610009	24.284752	9981.264932	1.382997	1.112799	30.202904	20.6
min	1.000000	21.000000	1300.000000	1.000000	1.000000	1872.000000	1950.0
25%	365.750000	59.000000	7553.500000	5.000000	5.000000	1954.000000	1967.0
50%	730.500000	69.000000	9478.500000	6.000000	5.000000	1973.000000	1994.0
75%	1095.250000	80.000000	11601.500000	7.000000	6.000000	2000.000000	2004.0
max	1460.000000	313.000000	215245.000000	10.000000	9.000000	2010.000000	2010.0

### Histogram for SalePrice

## Jupyter Notebook

Jupyter notebooks consist of "cells" — typically small chunks of code or narrative text in the Markdown format.

Users can execute cells (typically, but not necessarily, from top to bottom) and observe their outputs.

# Writing and Sharing Computational Notebooks in Various Contexts

## Data Science Education

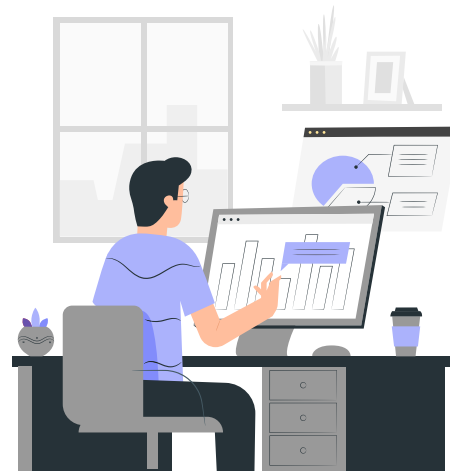
Kross and Guo, CHI 19

## Open Science

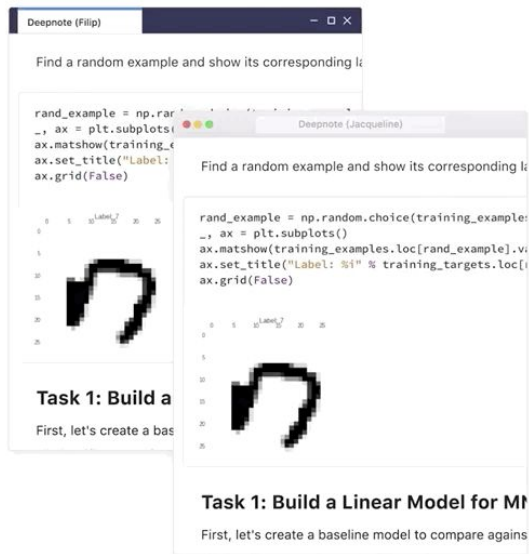
Randles et al., JCDL 17

## Professional Data Analytics

Kery et al., CHI 18



# From Sharing to Synchronous Editing



Deepnote (Filip)

Find a random example and show its corresponding label

```
rand_example = np.random.choice(training_examples.loc[training_targets['Label'] == '9'].index)
_, ax = plt.subplots(1, 1)
ax.imshow(training_examples.loc[rand_example].v)
ax.set_title("Label: %s" % training_targets.loc[rand_example].Label)
ax.grid(False)
```

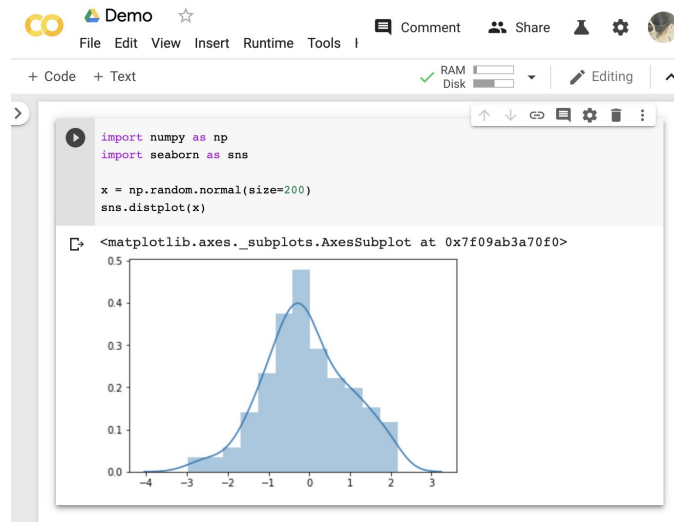
Deepnote (Jacqueline)

Find a random example and show its corresponding label

```
rand_example = np.random.choice(training_examples.loc[training_targets['Label'] == '9'].index)
_, ax = plt.subplots(1, 1)
ax.imshow(training_examples.loc[rand_example].v)
ax.set_title("Label: %s" % training_targets.loc[rand_example].Label)
ax.grid(False)
```

**Task 1: Build a Linear Model for MNIST**  
First, let's create a baseline model to compare against

Deepnote



Demo

File Edit View Insert Runtime Tools Help

+ Code + Text

```
import numpy as np
import seaborn as sns

x = np.random.normal(size=200)
sns.distplot(x)
```

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f09ab3a70f0>

0.5  
0.4  
0.3  
0.2  
0.1  
0.0

-4 -3 -2 -1 0 1 2 3

Google colab

# Issues with Synchronous Editing

- Reluctant to write together when collaboratively constructing a document
- Social embarrassment to be watched by others when typing

~ Wang et al. CSCW'17



# Issues with Synchronous Editing

**a Bill of Rights and Principles for Online, Networked Learners in the Digital Age**

**Editing recommendations:**

- Use a distinct color for any new text you add
- Use strikethrough rather than deleting

**Contributor colors:** Philipp (@schmittph), Piet (@bagabod), Doug Betshaw (@dabebshaw), Audrey Walters (@audreywalters), Steve Finkler, Cathy Davidson, Greg Borenstein, Scott Johnson, Rami Bacheld, Rachel Werthe-Chaney, William Torgerson

**If you want to edit,** send a tweet to any of the editors. This is ONLY to prevent spam, all requests will be answered asap.

Tweet at @framersrights

**Preamble**

Work on this Bill of Rights & Principles began in Palo Alto, California, on December 14, 2012. We convened a group of people passionate about learning, about serving today's students, and about using every tool we could imagine to respond better to the needs of students in a global, interactive, digitally-connected world. (moved to footnote)

The Internet has made it possible for anyone on the planet many people in wealthier countries to be a student, a teacher, and a creative collaborator at virtually no cost. Newer New technologies that can catalyze learning are bubbling up in less time than it takes to read this sentence. Some have emerged from universities, some from publicly-funded research, some from the private sector, some from individuals and digital communities. In the past year, Massive Open Online Courseware, or MOOCs, have become the darling of the moment—lauded by the media,

**Comments:**

- Philipp Schmitt** (12:01 PM on 28 Jul 2013)  
Based on the various threads about the title - I've taken out the "right" language, but left "online" I didn't get the sense that there was agreement on getting rid of it, and I think we risk making this too broad- and meaningless - if we try to cover everything).
- Cathy Davidson** (7:01 AM on 27 Jul 2013)  
Why not Rights and Principles for Networked Learners—that applies to online but also interactive face to face learning?
- Philipp Schmitt** (8:03 AM on 28 Jul 2013)  
I think two reasons speak for calling it "online learners": (1) We are only having this conversation because of incredible interest in ONLINE courses, spearheaded by the MOOCs. (2) focus is important, it sets us clearly state who we are and what we stand for. I think Show more
- Larry Goldberg** (8:01 PM on 26 Jul 2013)  
Suggestion on inclusion of people with disabilities: use more commonly used terminology such as "physical, sensory, cognitive or intellectual disability" instead of "bodily ability." a change I've never ran across.

```
server
├── etc
├── node_modules
├── src
│   ├── components
│   ├── handlers
│   └── interfaces
├── JS
│   ├── customer.js
│   ├── forgot-password.js
│   ├── login.js
│   ├── order.js
│   ├── phone.js
│   ├── vendor-data.js
│   └── zip-code.js
├── modules
├── utility
├── JS_Store
├── JS
├── app.js
├── JS
├── index.js
├── JS
├── socket.js
└── babel

Index.js
import fs from 'fs';
import http from 'http';
import socketIO from 'socket.io';

// Set up HTTP server
const HOSTNAME = '127.0.0.1';
const PORT = 1337;

http.createServer(function receiveRequest(req, res) {
  res.writeHead(200, {
    'Content-Type': 'text/plain'
  });

  res.end('Hello World!\n');
}).listen(PORT, HOSTNAME, () => {
  console.log('Server running at http://${hostname}:${port}');
});

// Set up web socket
```

## Collaborative Writing

Wang et al. CSCW'17

D'Angelo et al. CSCW'18

## Collaborative Programming

Goldman et al. UIST'11

Oney et al. CSCW'18



**What about collaborative data science?**

data science  $\neq$  writing + coding

**RQ1** What **tools and strategies** do data scientists currently use for collaboration?

**RQ2** Compared to **working on individual notebooks** in a collaborative setting, how does **synchronous notebook editing** change the way data scientists collaborate in computational notebooks?

**RQ3** What **challenges**, if any, do data scientists perceive in synchronous notebook editing?

## Study 1 Formative Survey

**RQ1** What **tools and strategies** do data scientists currently use for collaboration?

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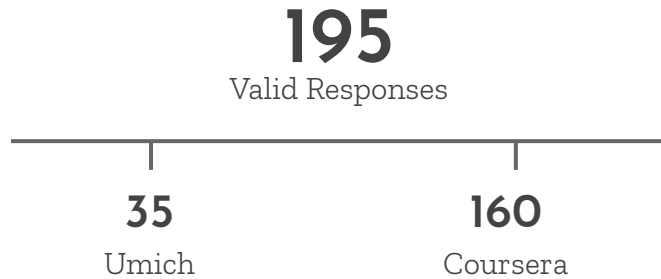
**RQ3** What **challenges**, if any, do data scientists perceive in synchronous notebook editing?

## Study 2 Observational Study

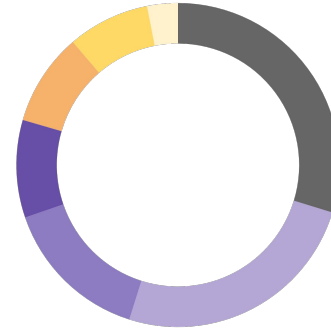
## Study 1 Formative Survey

### Demographic

#### Data Source



#### Job Roles



- Students
- Data Scientists
- Software Engineers
- Researchers
- Managers
- Business Analysts
- Others

#### Experience with Data Science



## Choices of Tools

<b>Programming</b>	Jupyter Notebooks (88.72%), IDEs (51.79%), <b>Google Colab (12.31%)</b>
<b>Communication</b>	Emails (79.49%), Face-to-face Communication (68.72%)
<b>Project Management</b>	Version Control Tools (49.74%), Task Tracking Tools (21.03%)

## Strategies for Keeping a Shared Understanding

<b>Discussions and Meetings</b>	Weekly meeting among team members;
<b>Frequently Check-in</b>	Communicate actively and frequently;
<b>Documentation</b>	Keep notes in Google Docs; ... comments in code;
<b>Organization</b>	Divide up the work into definable parts;
<b>Shared Assets</b>	Common repository for files;
<b>Others</b>	Code review to ensure code matched intent

**RQ1** What **tools and strategies** do data scientists currently use for collaboration?

### **Traditional Collaboration Setting**

Working on individual Jupyter notebooks

### **Emerging Collaboration Setting**

Working on notebooks with synchronous editing

## Study 1 Formative Survey

**RQ2** Compared to working on individual notebooks in a collaborative setting, how does synchronous notebook editing change the way data scientists collaborate in computational notebooks?

**RQ3** What challenges, if any, do data scientists perceive in synchronous notebook editing?

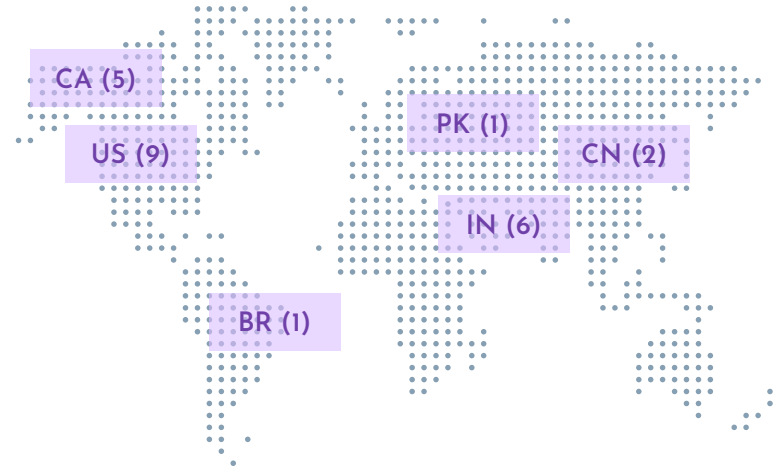
## Study 2 Observational Study



## Study 2 Observational Study

### Participants

- 24 participants (12 from the survey)
- Randomly assigned to pairs
- Work collaboratively on a predictive modeling problem remotely



# Study Setup

## Non-Shared Condition

Participants worked on individual notebooks

- ✓ Exchange the **notebook file**
- ✓ Set up a **git** repository
- ✓ Send **code snippets** through other tools if necessary

## Shared Condition

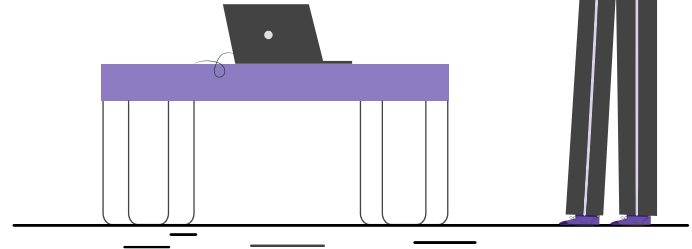
Synchronous editing was supported.

- ✓ Share **notebook edits and actions** (e.g., moving cursor, adding cells) in real-time
- ✓ Execute code on a **single interpreter**
- ✓ Update **output and runtime variables** among collaborators

## Study 2 Observational Study

### Task

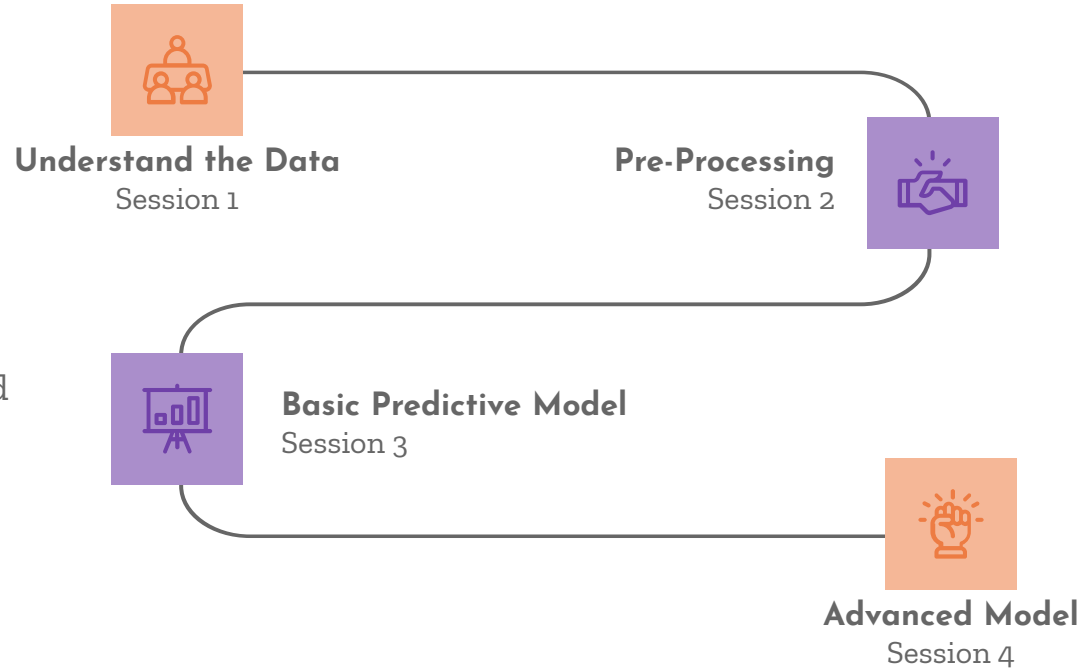
- Predict house sale prices using 80 features (e.g., lot size, year built)
- Additional incentives for the group with the lowest error score
- Submit prediction results as well as one Jupyter notebook report
- Choose from text-messaging (Slack) or video-conferencing (Google Hangouts) for communication



## Study 2 Observational Study

### Procedure

The study consisted of **four sessions**, each of which lasted **an hour**.



## Study 2 Observational Study

### Collaboration Style

Collaboration Style	GID	Definition
<b>Single Authoring</b>		One team member contributed the majority of ideas and did the majority of the implementation, while the others did not contribute much.
<b>Pair Authoring</b>		
<b>Divide and Conquer</b>		
<b>Competitive Authoring</b>		

## Study 2 Observational Study

### Collaboration Style

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## Study 2 Observational Study

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<b>Competitive Authoring</b>		Team members wrote the code for the same purpose and reached the consensus to use the code by whomever finished first.



### Collaboration Style

Collaboration Style	GID	Definition
<b>Single Authoring</b>	S2, S5	One team member contributed the majority of ideas and did the majority of the implementation, while the others did not contribute much.
<b>Pair Authoring</b>	S6	One team member did the majority of implementation while the others contributed ideas, engaged in discussions and reviewed the results.
<b>Divide and Conquer</b>	N2, N5, S1, S3, S4	Members divided the task into subgoals and explored the subgoals independently.
<b>Competitive Authoring</b>	N1, N3, N4, N6	Team members wrote the code for the same purpose and reached the consensus to use the code by whomever finished first.

### Communication Channels

	Non-Shared Condition	Shared Condition
Choices of Tools	Text Messaging (6/6)	Text Messaging (3/6) Video Conferencing (3/6)

Participants in the **non-shared condition** send files, code snippets, and output more often.

- Working in the **shared notebook** may **reduce the communication costs** by establishing a shared context.

## Study 2 Observational Study

### Final Submissions

→ Groups in the **shared condition** achieved a **better prediction result**.

- Non-Shared Condition
- Shared Condition

Error Score



→ Groups in the **shared condition** explored **more alternative models**.

Number of Alternative Models\*

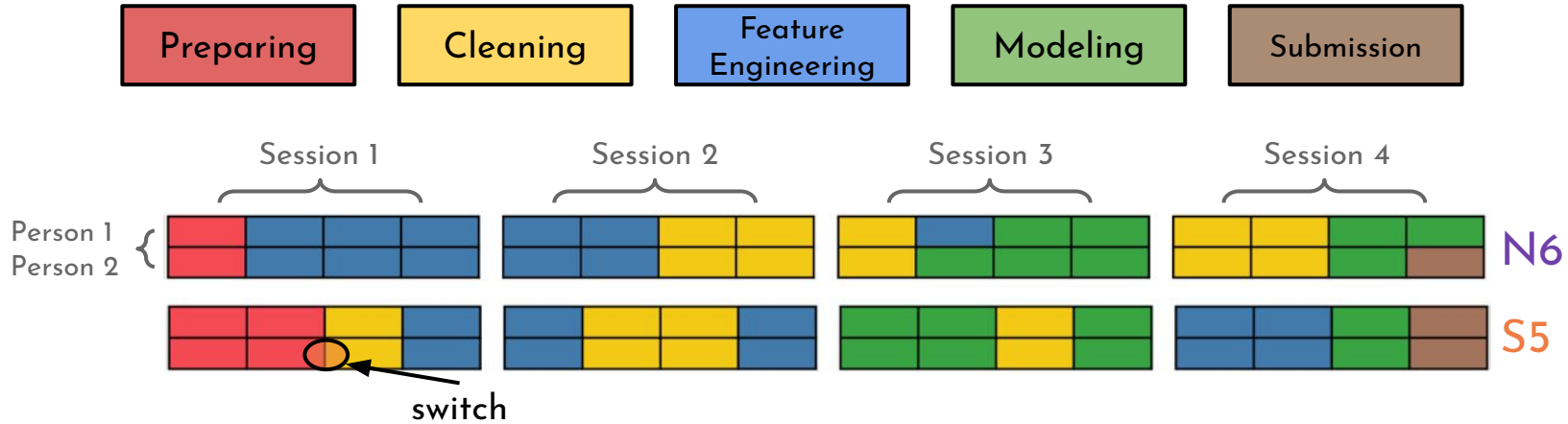


Lines in the Notebook\*



## Study 2 Observational Study

### Work Across Phases

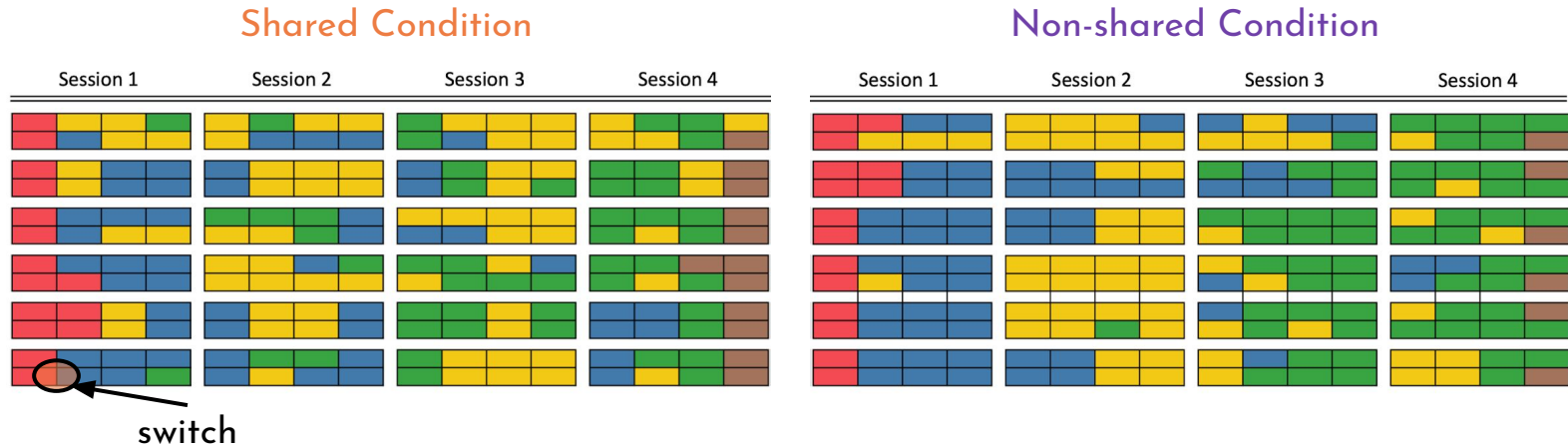


Participants in the **shared condition** switched more frequently ( $p < 0.001$ ).

→ Working on the same notebook provides collaborators with convenience to branch through tasks

## Study 2 Observational Study

### Work Across Phases



Participants in the **shared condition** switched more frequently ( $p < 0.001$ ).

→ Working on the same notebook provides collaborators with convenience to branch through tasks

## Benefits of Synchronous Editing in Notebook

- Reducing communication costs
- Flexibility to branch through tasks
- Enabling explorations of more alternative models
- Leading to a better prediction result

## Challenges of Synchronous Editing



# Challenges of Synchronous Editing

## 1. Interference with each other



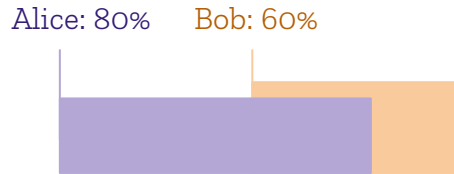
*"... **When using Jupyter Notebook together, it's hard to keep track of variable names.** Everyone might use a different name and may cause issues. For example, my teammate used train\_df as name, and later changed it to something else, but I wanted him to keep using the original name..." (P2 from S1)*



# Challenges of Synchronous Editing

## 2. Lack of Strategic Coordination

Why competitive authoring happens in the non-shared condition?



Why single authoring happens in the shared condition?



"... I feel I am not splitting work well enough. **I was thinking about how to get the work done and just tried the ideas on myself...**" (P11 from S2)

# Challenges of Synchronous Editing

S3 wrote down subtasks in the notebook.

## Pre-processing and cleaning

### Steps

1. Replace discrete values with indices
2. Remove data samples with too many missing features
3. Normalize continuous variables
4. Compute correlation, or use other techniques to select features

```
In [24]: 1 def count_nans(data):  
2         for name in data:  
3             count_nan = len(data[name]) - data[name].count()  
4             print(name, 'num of nans:', count_nan)
```

```
In [74]: 1 def label_encoding(data):  
2         for name in data:  
3             if data[name].dtype == 'object':
```

# Challenges of Synchronous Editing

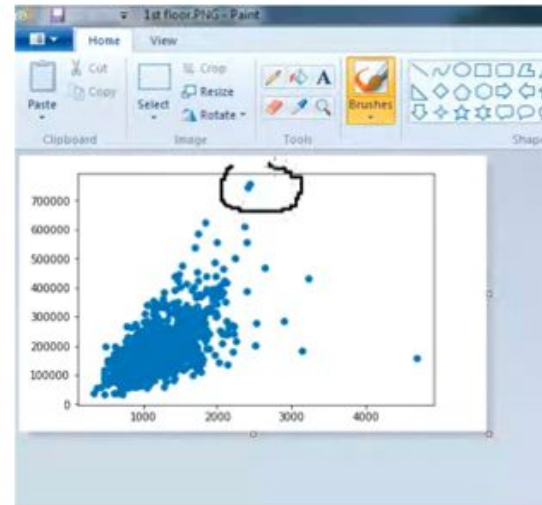
## 3. Contextual Chatting

P14 and P15 were looking at the scatterplots of independent variables together.

2:18 PM  
In my opinion there are outliers in all of our features  
there are 1 or 2 points that outliers

2:19 PM  
which ones?

P14 downloaded the graph, opened MS Paint, annotated the graph and sent it back to P15.



## Challenges of Synchronous Editing

1. Interference with each other
2. Lack of Strategic Coordination
3. Contextual Chatting
4. Lack of Awareness
5. Problems with the Linear Structure
6. Privacy Concerns

## Study 2 Observational Study

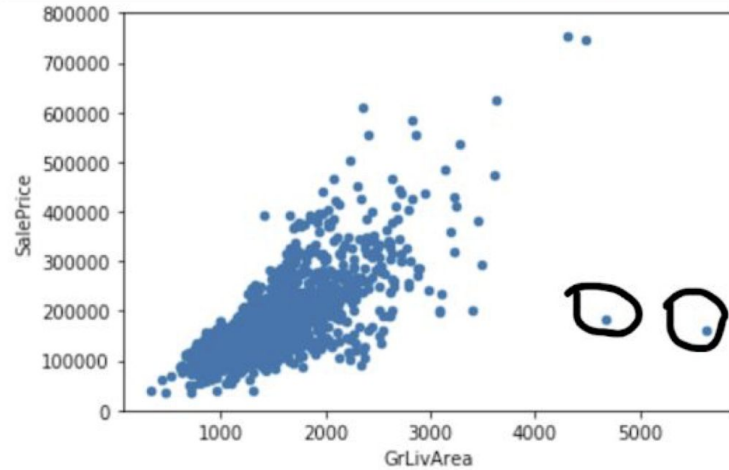
- Working on the same notebook results in **different collaboration styles** compared to working on individual notebooks.
- Synchronous editing tools improve collaboration by helping data scientists **maintain a shared context** and **improve work efficiency**.
- However, the current real-time collaborative editing features may lead to **several problems** (e.g., interference with each others' work, unbalanced contributions).

## Extending Our Understanding of Collaborative Editing Across Contexts

- Collaborators may hold different programming backgrounds and domain knowledge
- Different roles in collaborative data science

## Example: How to deal with the two dots?

```
In [8]: # scatter plot grlivarea/saleprice
label = 'GrLivArea'
data = pd.concat([df_train['SalePrice'], df_train[label]], axis=1)
data.plot.scatter(x=label, y='SalePrice', ylim=(0,800000));
```



## Design Implications

- Improve Awareness of Collaborators' Activity
- Provide Access Control
- Enable Discussions within Notebooks



### Limitations

- Generalizability
  - the type of data science problems
  - the expertise of collaborators
  - the size of the team
  - the synchronicity of the collaboration

# How Data Scientists Use Computational Notebooks for Real-Time Collaboration

What tools and strategies do data scientists currently use for collaboration?

Study 1 - Formative Survey on Collaborative Data Science

Traditional Collaboration Setting + Emerging Collaboration Setting

How does synchronous notebook editing change the way data scientists collaborate?

What challenges do data scientists perceive in synchronous notebook editing?

Study 2 - Observational Study on Collaborative Data Science

Having synchronous editing is great for collaborative data science, but not perfect!

42

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