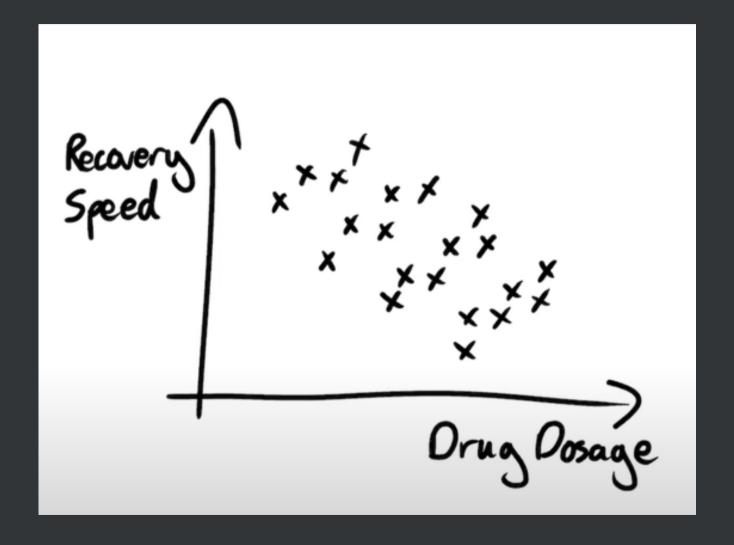
# More Data Tricks and Data Workshop

You are making a comparison between two hospitals:

- Hospital A: Among each 1,000 patients, 900 survived.
- Hospital B: Among each 1,000 patients, 800 survived.

Which hospital will you choose, and why?

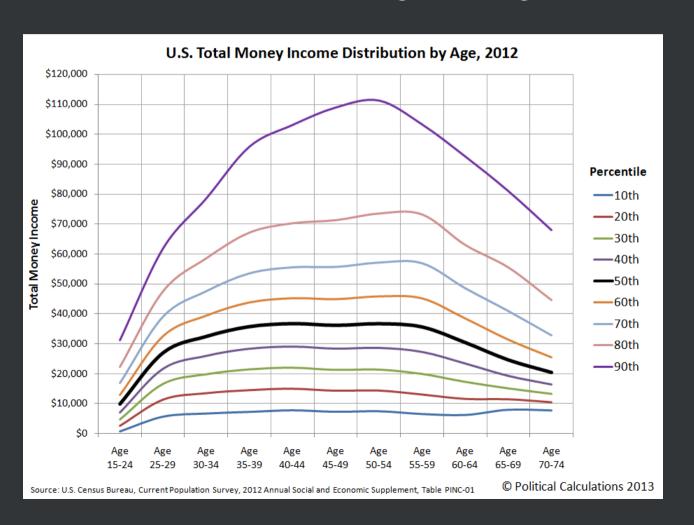
## Do patients recovery more slowly when taking more drugs?



https://www.youtube.com/embed/ebEkn-BiW5k?enablejsapi=1

## More Data Tricks

## How does income change with age?

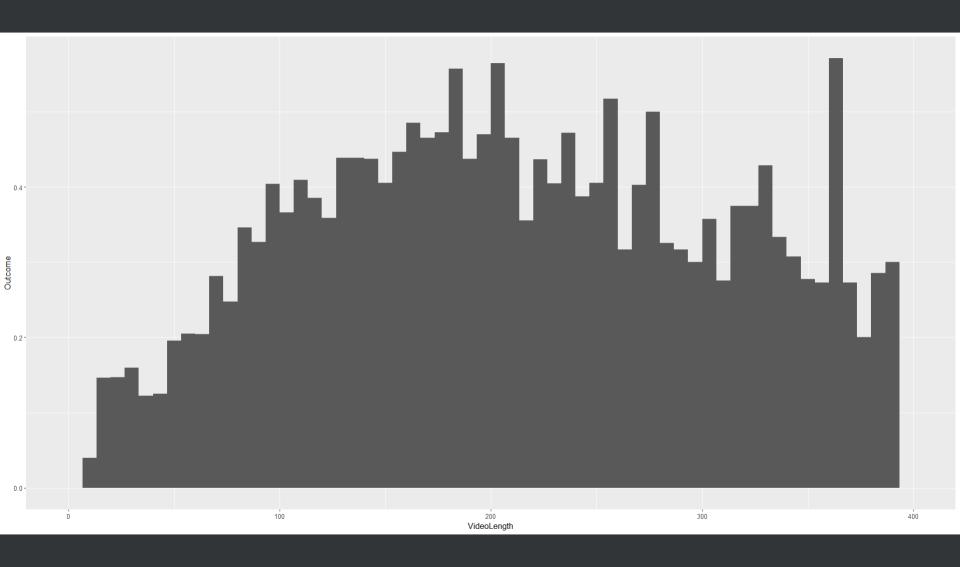


We want to investigate the relationship between video length and the chance of success. Let us prepare the data:

```
1 mydata <-
   read.csv("https://ximarketing.github.io/class/Kickstarter-
   Project.csv", fileEncoding = "UTF-8-BOM")
2 subdata = subset(mydata, IsVideoAvailable == 1)</pre>
```

We want to investigate the relationship between video length and the chance of success. Let us prepare the data:

```
1 library(ggplot2)
2 ggplot(subdata, mapping = aes(VideoLength, Outcome)) +
3 stat_summary_bin(fun.y="mean", geom="bar", bins=60)+xlim(0, 400)
```



The relationship between video length and project success appears to be nonlinear. Shorter videos can enhance the success rate as their length increases; however, excessively long videos do not provide additional benefits to the project.

Let us try the following logistic regression:

$$ext{Pr[Success]} = rac{1}{1 + \exp(-(a + b_1 imes ext{Length} + b_2 imes ext{Length}^2))}$$

## Consider the following code:

```
1 logit <- glm(Outcome ~ VideoLength + I(VideoLength^2), data =
   subdata, family = "binomial")
2 summary(logit)</pre>
```

#### Coefficients:

```
Estimate Std. Error z value Pr(>|z|) (Intercept) -1.192e+00 8.958e-02 -13.307 < 2e-16 *** VideoLength 6.541e-03 8.102e-04 8.074 6.81e-16 *** I(VideoLength^2) -1.056e-05 1.584e-06 -6.666 2.63e-11 ***
```

## Question

Suppose that you want to predict students' performance in exam. Two factors come into play: IQ and Hours of Study.

- A student with a higher IQ is more clever, and gets higher grades on average.
- A student who studies longer hours understands the content better, and gets higher grades on average.

We want to investigate the relationship between the total funding, the creators' experience and the number of products offered. Let us prepare the data:

```
mydata <-
read.csv("https://ximarketing.github.io/class/Kickstarter-
Project.csv", fileEncoding = "UTF-8-BOM")
mydata$LogFunding = log(mydata$FundingRaised + 1)</pre>
```

We want to investigate the relationship between the total funding, the creators' experience and the number of products offered. Let us run a regression with an interaction term:

```
1 result = lm(LogFunding ~ Created * NumberOfProducts, data =
   mydata)
2 summary(result)
```

```
Estimate Std. Error t value Pr(>|t|) (Intercept) 3.039097 0.071653 42.414 < 2e-16 *** Created 0.240761 0.042393 5.679 1.41e-08 *** NumberOfProducts 0.443064 0.008182 54.148 < 2e-16 *** Created:NumberOfProducts -0.012090 0.005019 -2.409 0.016 *
```

 $\operatorname{LogFunding} = 3.04 + 0.24 \times \operatorname{Created} + 0.44 \times \# \text{ of Products} - 0.012 \times \operatorname{Created} \times \# \text{ of Products}$ 

What do you learn from the results?

Let's explore something more interesting...

We already know that in a crowdfunding project, putting your face in front of the camera makes the project more successful.

However, does it make a difference whether this is a female face or a male face? What's your intuition?

```
1 subdata = subset(mydata, IsVideoAvailable == 1)
      2 result = lm(LogFunding ~ factor(Gender) * Human, data =
        subdata)
      3 summary(result)
                       Estimate Std. Error t value Pr(>|t|)
                                                               ***
                         5.1768
                                     0.3778 \quad 13.704 \quad < 2e-16
(Intercept)
factor(Gender)M
                         0.4710
                                     0.4003 1.177 0.23939
                                               4.734 2.26e-06
                                                               ***
factor(Gender)U
                         1.9207
                                     0.4057
                                                               ***
                                     0.4137 5.672 1.48e-08
                         2.3467
Human
                                                               **
factor(Gender)M:Human
                        -1.1873
                                     0.4411 - 2.692 0.00713
factor(Gender)U:Human
                        -0.5688
                                     0.4453
                                              -1.277 0.20153
```

# Group Data Project II: HK Property Valuation

In this project, we want to understand the HK real estate market. We have collaborated with Centaline (中原地產), one of the largest property agencies in Hong Kong, to get the property transaction data in Hong Kong.



#### ACRC

Asia Case Research Centre 亞洲案例研究中心

XI LI KELVIN S.K. WONG CHURONG WANG

# VALUATION OF HONG KONG RESIDENTIAL PROPERTY

Kelvin Wong is Professor of Real Estate at the University of Hong Kong. Churong Wang is currently my PhD student. Churong Wang (王楚绒)

Ph.D. Candidate in Marketing, HKU M.Sc in Business Analytics, HKU B.A. in Architecture, HKU

She ranked No. 19 in National College Entrance Examination (高考) in Yunnan Province in 2016

```
Sign up here: https://acrc.internad.hk/
Adding the course content here:
https://acrc.internad.hk/enrol/1000025503
Loading the data:
```

```
1 df =
  read.csv('/dataset/Centaline/Centaline_train.csv',header=
    TRUE)
```

Transaction\_price: The transaction price of the property (in Hong Kong dollars). You may want to take the log transformation of this variable.

Transaction\_year: The year in which the transaction takes place (e.g., 2020).

Transaction\_month: The month in which the transaction takes place (e.g., 10 for October). When using this variable, you may want to take it as a fixed effect.

Location and Estate: The location and estate for each property. Please do not use them in your data analysis.

HMA: It stands for "Housing Market Area", a term used to describe the area at which the property is located (e.g., Pok Fu Lam).

Developer: The developer of the property (e.g., Hang\_Lung\_Group for 恆隆集團). If the developer is a small developer not included in the dataset, then the value is "Other".

Gross\_size: 建築面積 in Chinese. It is measured in square foot. If data is unavailable for a property, then its Gross\_size = -1.

Saleable\_size:使用面積 in Chinese. It is measured in square foot. If data is unavailable for a property, then its Saleable\_size = -1.

No\_of\_rooms: The number of rooms in the property. 0 means studio; -1 means data is not available.

Floor: The floor of the property (10 for 10th floor).

Region: The region of the property; it takes values Hong Kong, Kowloon and New Territories.

Primary\_school:小學學區 in Chinese. Primary school Net divides Hong Kong's primary schools into 36 zones

Secondary\_school: 中學學區 in Chinese. Secondary schools use a zoning system based on the 18 districts in Hong Kong.

Age\_of\_property: The age of the property in years; -1 means the property is not built yet (-1 對應樓花).

Uncompleted: Whether the construction is completed. 0 means completed and 1 means under construction.

MTR\_station: The name of the nearest MTR station. -1 means property is distant from all MTR stations.

Close\_to\_MTR: Whether the property is close an MTR station. 1 means close to and 0 means far from MTR stations.

Shopping\_Mall, Swimming\_Pool, Sport\_facility, Club, Garden, Sauna\_Shower, Playground, Cinema, Bar\_karaoke, Study\_Room, Ballroom: These are all binary variables. 1 means the amenity is available while 0 means there are no such amenities.

District: One of Hong Kong's 18 districts.

Median\_income: The median income in the HMA.

Median\_age: The median age of residents in the HMA.

Population: The total population of the HMA.

Unit: Number of property units in the HMA.

## Sample Codes (run on DAP)

```
1 library(stargazer)
2 mydata = read.csv('/dataset/Centaline/Centaline_train.csv',header=TRUE)
3 head(mydata)
4 mydata$LogPrice = log(mydata$Transaction_price)
5 result <- lm(LogPrice ~ Age_of_property * Close_to_MTR, data = mydata)
6 summary(result)
7 stargazer(result, title = "regression output", align = TRUE, out = "regression.html", type = "html")</pre>
```

## What should we do in this project?

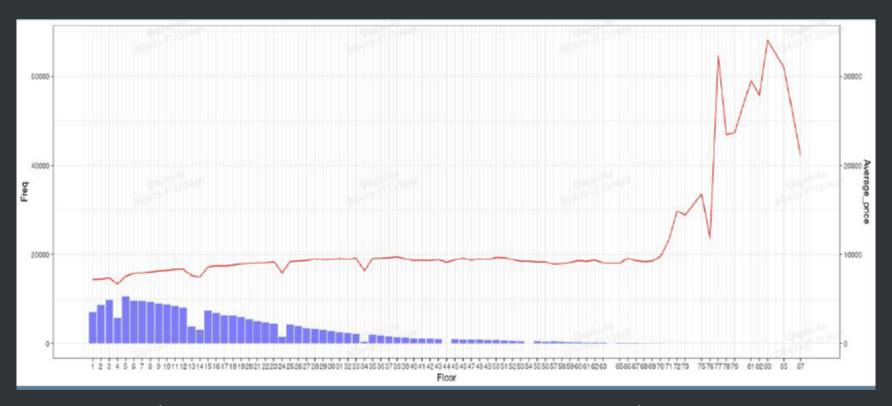
Each group should only ask one big research question in your project. Quality beats quantity. Choose the right data analysis methods and come up with a good answer to your questions, with implications for sellers, buyers, developers, property agencies and the government.

## What should we do in this project?

You need to include at least one interaction term or a square term in your analysis.

## A full-score example (last year)

## Floor Numbers and Housing Price



There is a significant drop in prices when floor is 13 or ends with 4. But not for 18.

## Special Numbers and Housing Price

The higher the floor, the higher the unit price.

However, the marginal effect of floor on unit price is decreasing with the floor level.

#### Submission

To save your time, you only need to submit a few pages of slides (no more than 12 pages for main text + no more than 6 slides for appendix) to Moodle covering your research question(s), data analysis (e.g., regression equations), findings, and implications.

No reports/presentations are needed!

Deadline: Jan 11, 2024

12:30 for Class A, 17:00 for Class B, and 21:30 for Class C

# Individual Project

# Why do we have an individual project?

As a professor, I hate all kinds of individual assignment; it means I need to grade 200+ copies myself...

But I have no other choice. For a group project, we always have free-riders or students who do not analyze the data.

## Individual Project

I understand that not everyone is going to become an analyst in the future. So it is fine that you don't work on data. That's why we have two options for each of you.

Option 1: Collect and analyze your own dataset

Option 2: Find and discuss an innovative data strategy

This is similar to your group projects; the difference is that you need to collect your own dataset now. Let me give you a few sources for data.

Kaggle, a platform with all types of datasets. [Video]

Amazon Reviews, good for text mining.

Yelp Reviews

Stanford Social Network Data

Harvard Datasets

Don't be too aggressive! Try some small datasets that you are able to handle. We do have a deadline for this.

You can also scrape data yourself (this may be a bit risky because you may face a lot of practical issues; e.g., the website is a dynamic website, your IP address is blocked)

You can also hire a scraper on Taobao to help you (cost usually a few hundred RMB). But make sure you can finish the task on time.



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¥300.00 1000+人付款 上海

69

If you still have no data and don't have time, let me know.

Discuss the innovative data strategy of a company (like writing a case study). You need to search for resources online. For instance, you can discuss

- How does Uber use data to set prices?
- How does Freshippo (盒马) use data to optimize its retail operations?

#### Deliverables

If you choose option 1, you need to submit up to 20 pages slides explaining your data sources, question, analysis, findings, and implications.

If you choose option 2, you need to submit up to 5 pages report (in PDF format, not including title page) discussing the company's big-data marketing strategy.

# What is a good project?

Option 1: Find a good data, ask an interesting question and obtained useful, especially surprising findings.

Option 2: Find some innovative uses of data that were not well-known before (e.g., Uber collects your phone battery data; Target collects satellite images).

#### Deadline

Two weeks from today, that is,

Deadline: Jan 16, 2024

12:30 for Class A, 17:00 for Class B, and 21:30 for Class C