















Data quality

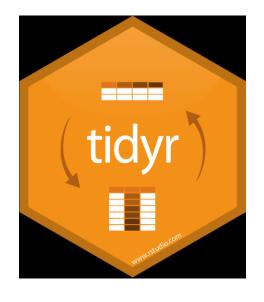


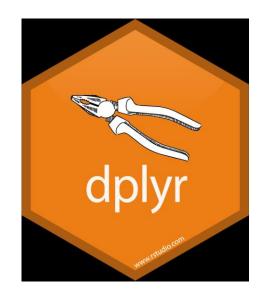


Data quality

Planning for data preparation













DATA

Only 3% of Companies' Data Meets Basic Quality Standards

by Tadhg Nagle, Thomas C. Redman, and David Sammon

September 11, 2017

















2020 Global data management research

Our 2020 Global data management benchmark report launches February 18th. Sign up to access our exclusive industry insights and learn how you can gain control over your data.

Our new research will help you unlock the true power of your organization's data, helping you transform your business. We spoke with more than 1,000 global professionals to uncover today's top challenges in leveraging trusted data and tips for how you can drive a data-driven culture.





Data asset that isn't necessarily fit for the purpose or has high degree of inaccuracy



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78% of organizations suffer from data debt.

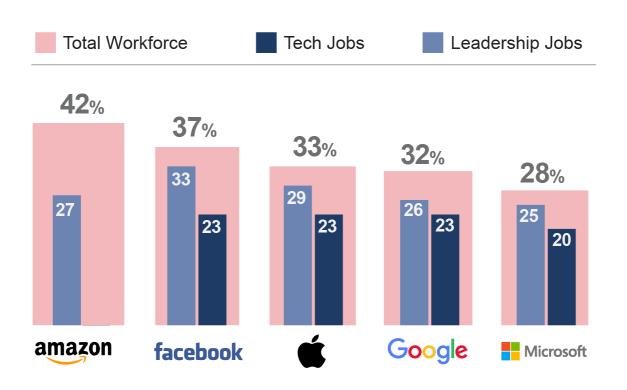


Data Debt Hurts

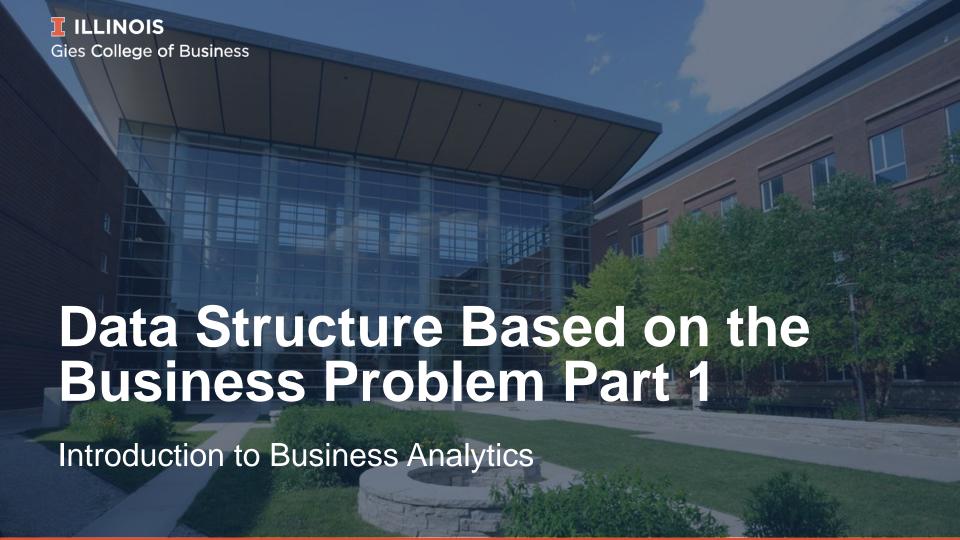
- Lack of trust on insights from the data
- Poor ROI on the data and tech investments
- Barrier to being a data-driven company

Proportion of Female in Tech Industry













The real data



The data that goes for modeling









Business Analytics



Managerial Domain of Action/Decision

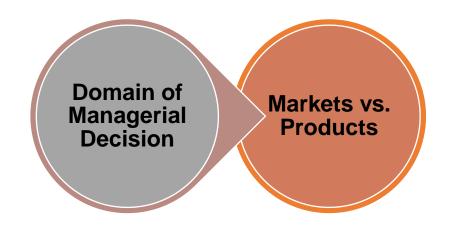


Analysis



Data







Market Focus





Which markets to focus more on?



Analysis

Compare markets for profitability potential



Data Structured

For market level analysis



Product Focus







Problem Focus

Which products to focus more on?

Analysis

Compare products for profitability potential

Data Structured

For product level analysis



Data prepared for Market level decision

Data prepared for Product level decision

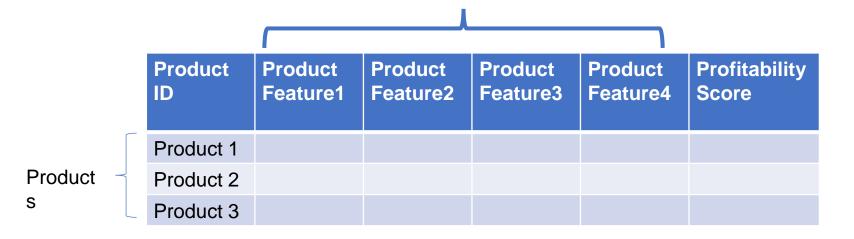


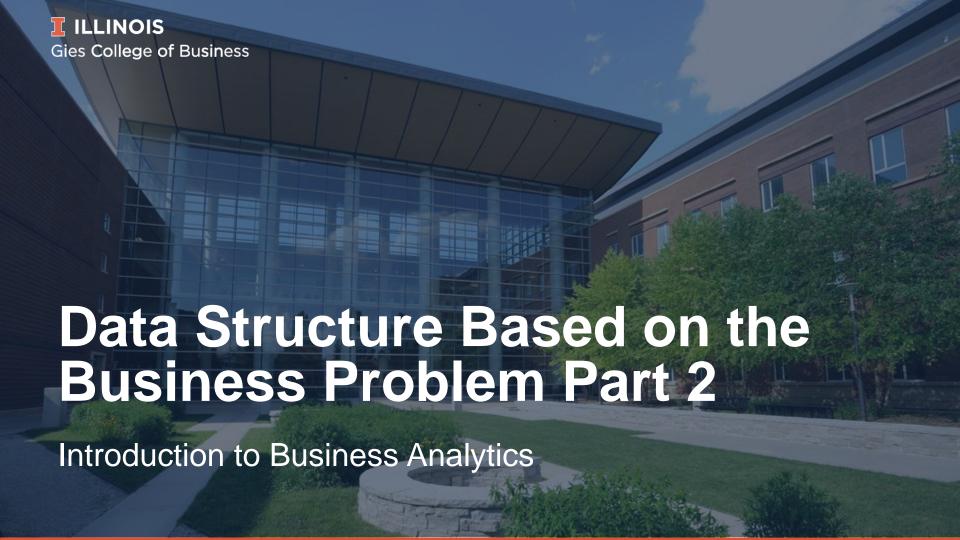
Characteristics of each row/Markets

			*				
	Market ID	Market Feature1	Market Feature2	Market Feature3	Market Feature4	Profitability Score	
Market s	Market1						
	Market2						
	Market3						



Characteristics of each row/Product







Levels within the Product Domain

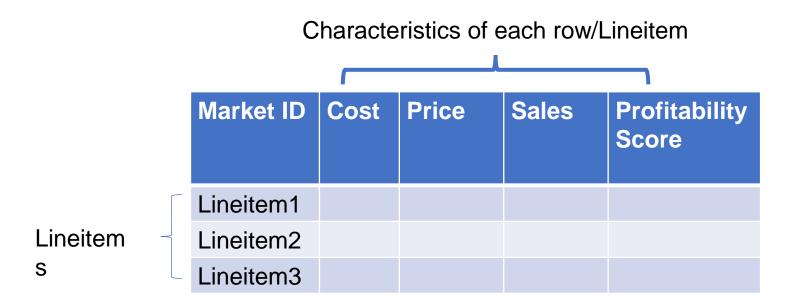
Department

Category

Lineitem



Data Structured for Lineitem Level Analysis



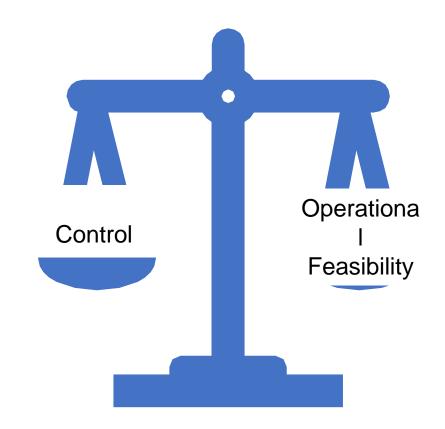


Levels within the Domain of Time













Two useful functions in R

■%in%

Pipe %>%



• Filter "ities" data where *Operation Type is SALE* and then select columns Cost and Price

Code using %>%

df_piped <- df %>% filter(Operation_Type == "SALE") %>% select(Cost, Price)

I

Code without Using %>%

```
Two steps:
                Intermediate object
a) df1 <- filter(df, OperationType = "SALE")
b) df2 <- select(df1, Cost, Price)
         Passed as data
           argument
```



Code without using %>%

df2 <- select(filter(df, OperationType = "SALE"), Cost, Price)

Piped code

df_piped <- df %>% filter(OperationType == "SALE") %>%
select(Cost, Price)



• The output returned by %>% is always a dataframe

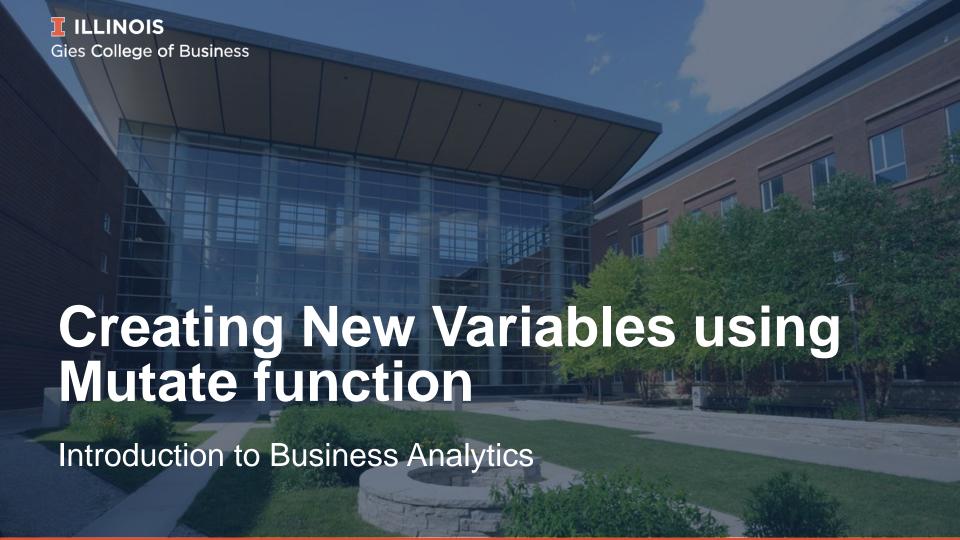
 The output returned by %>% is always a valid input (data argument) to any tidyverse function such as filter, select, ggplot, join, etc.

) [

%in%



%>%



)[

• Mutate - To create a new column

Distinct - To extract the unique value





Dplyr Functions for Aggregating Data

1. summaries

2. group by



Aggregating Data

Customer Level Insight

Which are the best customers ?



Scoring each customer



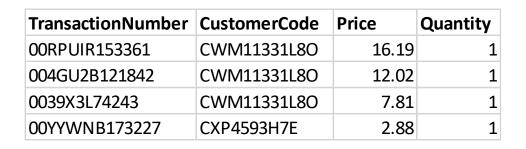
Aggregate d at the customer level





Transaction Data Aggregated at the Customer Level

Transaction Level Data



Customer Level Data

CustomerCode	Avg_Price	Avg_Quantity
CWM11331L8O	12.84133333	1
CXP4593H7E	2.88	1



Summary Measures

Mean

• Sum

Median



Dplyr Functions for Aggregating Data

1. summarise

2. group_by





Identifying and Handling Missing Values in R

Missing values are saved as "na"

is.na()

Returns list of TRUE and FALSE



Handling Missing Values

- Mean imputation
- Median imputation (when there are outliers)
- Model-based approach

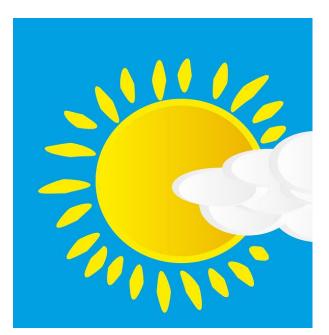
Dropping the rows with missing values



Joining Datasets Using Dplyr Package



Add More Context to Our Analysis







https://pixabay.com/vectors/cloud-weather-forecast-weather-sky-346710/https://pixabay.com/photos/football-playing-field-corner-1419954/

https://pixabay.com/photos/media-social-media-apps-998990/



Levels of the Datasets to Join Must Match

Date Level

‡	date	PRCP [‡]	snow [‡]	TMAX [‡]	TMIN [‡]
2138	2017-04-03	0.0000000	0.0000000	60.08	33.08
2137	2017-04-02	0.0000000	0.0000000	57.02	32.00
2136	2017-04-01	0.1181102	0.0000000	57.02	33.98
2135	2017-03-31	4.2125984	0.0000000	62.06	33.98
2134	2017-03-30	0.0000000	0.0000000	64.94	46.94
2133	2017-03-29	0.0000000	0.0000000	68.00	42.98
2132	2017-03-28	0.0000000	0.0000000	69.08	39.92
2131	2017-03-27	0.0000000	0.0000000	64.94	30.92
2130	2017-03-26	1.4960630	0.0000000	57.92	39.92

Transaction Level – many rows for one date

÷	Time	OperationType [‡]	BarCode
392387	2017-04-03T17:08:00Z	SALE	*
392388	2017-04-03T17:08:00Z	SALE	*
392390	2017-04-03T16:52:00Z	SALE	*
392396	2017-04-03T16:49:00Z	SALE	*
392402	2017-04-03T16:46:00Z	SALE	*
392401	2017-04-03T16:46:00Z	SALE	*
392406	2017-04-03T16:26:00Z	SALE	*
392410	2017-04-03T15:36:00Z	SALE	*
392416	2017-04-03T14:47:00Z	SALE	*

Needs to be aggregate at the "date" level



Joining Datasets

Left Right

df_ities_date df_weather



Datasets Are Merged/Joined Based on a Common Column

Common column – primary key

- Unique values
- Non-missing values



Four Types of Join

- Left Join
- Right Join
- Full Join
- Inner Join





Two Formats for Representing Data

Long Format

Wide Format

Long Format

LineItem	CashierName [‡]	Quantity [‡]	CustomerSatisfaction [‡]
Aubergine and Chickpea Vindaloo	Katherine Roth	1.362319	0.990235803
Aubergine and Chickpea Vindaloo	Rachael Price	1.305556	0.990235803
Aubergine and Chickpea Vindaloo	Trinidad Johnson	1.000000	0.990235803
Aubergine and Chickpea Vindaloo	Vincent Ball	1.506024	0.990235803
Beef and Apple Burgers	Katherine Roth	1.011364	0.325210194
Beef and Apple Burgers	Rachael Price	1.000000	0.325210194
Beef and Apple Burgers	Trinidad Johnson	1.000000	0.325210194
Beef and Apple Burgers	Vincent Ball	1.015432	0.325210194
Beef and Apple Burgers - Illini Bhangra	Katherine Roth	1.335723	0.527040931
Beef and Apple Burgers - Illini Bhangra	Rachael Price	1.259542	0.527040931
Beef and Apple Burgers - Illini Bhangra	Trinidad Johnson	1.229141	0.527040931
Beef and Apple Burgers - Illini Bhangra	Vincent Ball	15.000000	0.527040931



Wide Format

LineItem	Quantity_Katherine [‡] Roth	Quantity_Rachael Price	Quantity_Trinidad	Quantity_Vincent \$\frac{\pi}{2}\$ Ball	CustomerSatisfaction
Aubergine and Chickpea Vindaloo	1.362319	1.305556	1.000000	1.506024	0.990235803
Beef and Apple Burgers	1.011364	1.000000	1.000000	1.015432	0.325210194
Beef and Apple Burgers - Illini Bhangra	1.335723	1.259542	1.229141	15.000000	0.527040931



Long → Wide

• The number of rows decreases

The number of columns increases



Wide → Long

- A few columns are dropped
- The number of rows increases





Str_*verb/ noun*

- Replace
- Detect
- Extract
- Length





Length of LineItem Name

- Correlation(LineItem_length, Quantity) = -0.09
- Correlation(LineItem_length, Price) = 0.037
- Correlation(LineItem_length, Cost) = 0.045

