

# Manipulate survey data

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This file imports, transforms, and then exports the data that I created for the *Assessment Institute* meeting in Indianapolis in October 2024.

## 0.1 Setup & import libraries

Standard library import steps. Enables all that is to come.

```
library(tidyverse)
library(skimr)
```

## 0.2 Import data

Data in CSV format with headers.

```
survey <- read_csv("data/retention_survey_history.csv")
```

### 0.3 Examine data

Show us a bit of information about the imported data.

`glimpse(survey)`

Rows: 33,524

Columns: 24

```

$ Year <dbl> 2012, 2012, 2012, 2012, 2012, 2012, 2012, 2012, 2012, 2012, 2012, 2012, 2012
$ ID <chr> "xuojqdfdozvu", "vvwkinqvnibo", "ibyjcmiopiqa", "lsqamawy~"
$ NPS <dbl> 8, 8, 8, 8, 8, 8, 8, 4, 4, 7, 6, 6, 6, 4, 7, 4, 6, 7, 4, ~
$ Field <chr> "Undecl", "SocSci", "CompSci", "Other", "SocSci", "Undecl~
$ ClassLevel <chr> "Sr", "Sr", "Sr", "Sr", "Sr", "Sr", "Sr", "Fresh", "Fresh~
$ Status <chr> "Full-time", "Full-time", "Full-time", "Full-time", "Full~
$ Gender <chr> "Male", "Female", "Male", "Female", "Other", "Female", "M~
$ BirthYear <dbl> 1990, 1999, 1999, 1989, 1993, 1988, 1996, 1991, 1990, 198~
$ FinPL <chr> "No", "No", "No", "No", "Yes", "Yes", "No", "No", "No", "No", ~
$ FinSch <chr> "No", "No", "No", "Yes", "Yes", "Yes", "Yes", "No", "No", "No", ~
$ FinGov <chr> "No", "Yes", "Yes", NA, "No", "No", "No", "No", "No", "No", "No~
$ FinSelf <chr> "No", "Yes", "No", "No", "Yes", "Yes", "No", "Yes", "No", "No", ~
$ FinPar <chr> "Yes", "No", NA, "Yes", "No", "No", "No", "No", "No", "No", "Ye~
$ FinOther <chr> "No", "No~
$ TooDifficult <chr> "Disagree", "Strongly Disagree", "Disagree", "Agree", NA, ~
$ NotRelevant <chr> "Strongly Disagree", "Disagree", NA, "Strongly Disagree", ~
$ PoorTeaching <chr> "Agree", "Agree", "Agree", "Disagree", "Agree", "Strongly~
$ UnsuppFac <chr> NA, "Neutral", "Neutral", "Neutral", NA, "Neutral", "Stro~
$ Grades <chr> NA, "Disagree", "Strongly Disagree", NA, NA, "Agree", NA, ~
$ Sched <chr> "Strongly Agree", "Strongly Disagree", "Strongly Disagree~
$ ClassTooBig <chr> "Neutral", NA, "Strongly Disagree", "Strongly Disagree", ~
$ BadAdvising <chr> "Disagree", "Disagree", NA, "Disagree", "Strongly Disagre~
$ FinAid <chr> "Strongly Agree", "Strongly Agree", "Strongly Agree", "Ag~
$ OverallValue <chr> "Strongly Agree", "Strongly Agree", "Neutral", "Strongly ~

```

names(survey)

```
[1] "Year"           "ID"            "NPS"          "Field"         "ClassLevel"  
[6] "Status"        "Gender"        "BirthYear"     "FinPL"        "FinSch"  
[11] "FinGov"        "FinSelf"       "FinPar"        "FinOther"     "TooDifficult"  
[16] "NotRelevant"  "PoorTeaching" "UnsuppFac"    "Grades"       "Sched"  
[21] "ClassTooBig"   "BadAdvising"  "FinAid"        "OverallValue"
```

```
skim(survey)
```

Table 1: Data summary

Name	survey
Number of rows	33524
Number of columns	24
Column type frequency:	
character	21
numeric	3
Group variables	None

### Variable type: character

skim_variable	n_missing	complete_rate	min	max	empty	n_unique	whitespace
ID	0	1.00	12	12	0	33524	0
Field	12	1.00	2	9	0	15	0
ClassLevel	0	1.00	2	5	0	5	0
Status	627	0.98	5	9	0	3	0
Gender	875	0.97	4	6	0	3	0
FinPL	1674	0.95	2	3	0	2	0
FinSch	1615	0.95	2	3	0	2	0
FinGov	1586	0.95	2	3	0	2	0
FinSelf	1622	0.95	2	3	0	2	0
FinPar	1699	0.95	2	3	0	2	0
FinOther	1699	0.95	2	3	0	2	0
TooDifficult	6645	0.80	5	17	0	5	0
NotRelevant	6622	0.80	5	17	0	5	0
PoorTeaching	6819	0.80	5	17	0	5	0
UnsuppFac	6676	0.80	5	17	0	5	0
Grades	6637	0.80	5	17	0	5	0
Sched	6684	0.80	5	17	0	5	0
ClassTooBig	6634	0.80	5	17	0	5	0
BadAdvising	6735	0.80	5	17	0	5	0
FinAid	6676	0.80	5	17	0	5	0
OverallValue	6631	0.80	5	17	0	5	0

### Variable type: numeric

skim_variable	n_missing	complete_rate	mean	sd	p0	p25	p50	p75	p100	hist
Year	0	1	2014.50	1.71	2012	2013	2014	2016	2017	
NPS	0	1	5.95	1.52	4	4	6	7	8	

skim_variable	n_missing	complete_rate	mean	sd	p0	p25	p50	p75	p100	hist
BirthYear	0	1	1994.01	3.76	1988	1991	1994	1997	2000	

## 0.4 Fix NA values

Just in case any columns represent NA values in a variety of ways, we can standardize the representation before we continue to simplify any logic later in the process.

### 0.4.1 Create standardized representation of NA

Actually do the transformation.

```
survey <-  
  survey |>  
    mutate(across(c(Field:Gender,  
                  FinPL:OverallValue),  
                 ~ case_when(is.na(.x) ~ NA,  
                             .x == "" ~ NA,  
                             .x == "NA" ~ NA,  
                             .x == "--" ~ NA,  
                             .default = .x)))
```

### 0.4.2 Ensure that it works

Just what it says — ensure that it works. This counts the number of appearances in every single column of the *non-standard* ways in which NA values might have been represented. We should see nothing but zero (0) values in the table below.

```
survey |>  
  summarize(across(everything(),  
                  ~sum(.x %in% c("NA", "--", "")))) |>  
  print(width = Inf)
```

# A tibble: 1 x 24  
 Year ID NPS Field ClassLevel Status Gender BirthYear FinPL FinSch FinGov  
 <int>  
1 0 0 0 0 0 0 0 0 0 0 0 0  
 FinSelf FinPar FinOther TooDifficult NotRelevant PoorTeaching UnsuppFac Grades  
 <int> <int> <int> <int> <int> <int> <int> <int>  
1 0 0 0 0 0 0 0 0  
 Sched ClassTooBig BadAdvising FinAid OverallValue  
 <int> <int> <int> <int>  
1 0 0 0 0 0

## 0.5 Factors

### 0.5.1 Define factors

Define factors for as many fields as we can. Specify its order when we can so that later analyses are better organized.

```
fin_cols <- c("FinPL", "FinSch", "FinGov",
              "FinSelf", "FinPar", "FinOther")
exp_cols <- c("TooDifficult", "NotRelevant", "PoorTeaching",
              "UnsuppFac", "Grades", "Sched", "ClassTooBig",
              "BadAdvising", "FinAid", "OverallValue")
fin_levels <- c("Yes", "No")
exp_levels <- c("Strongly Disagree", "Disagree", "Neutral", "Agree",
                 "Strongly Agree")
class_levels <- c("Fresh", "Soph", "Jr", "Sr")

survey <-
  survey |>
    mutate(across(all_of(exp_cols),
                 ~factor(.x,
                           levels = exp_levels,
                           ordered = TRUE)))

survey <-
  survey |>
    mutate(across(all_of(fin_cols),
                 ~factor(.x,
                           levels = fin_levels,
                           ordered = TRUE)))

survey <-
  survey |>
    mutate(ClassLevel = factor(ClassLevel,
                                levels = class_levels,
                                ordered = TRUE))

survey <-
  survey |>
    mutate(Status = factor(Status,
                           levels = c("Full-time", "Part-time",
                                      "Other")))

survey <-
  survey |>
    mutate(Gender = factor(Gender,
                           levels = c("Female", "Male", "Other")))
```

We want to handle `Field` of study somewhat differently. It has `NA` values in the column, but we already have `Undecl` and `Other` as values. We would rather include the `NA` values in the analysis,

but we want to be able to see if the results for that value differ from the results of these other two values. Let's make `Unknown` an accepted value for `Field`.

```
poss_fields <- c("LifeSci", "PhysSci", "PubHealth", "Nurs", "OthHealth",
                 "PubAdm", "SocSci", "ArtsHum",
                 "CompSci", "Eng", "Bus", "Ed", "ArchUP",
                 "Other", "Undecl", "Unknown")

survey$Field <-
  survey$Field |>
  replace_na("Unknown")
survey <-
  survey |>
  mutate(Field = factor(Field,
                        levels = poss_fields))
```

### 0.5.2 Validate the factor creation process

Let's show the `structure` of the table. This will allow us to validate that all of our `factor` declarations above worked.

```
str(survey)
```

```
tibble [33,524 x 24] (S3: tbl_df/tbl/data.frame)
$ Year      : num [1:33524] 2012 2012 2012 2012 2012 ...
$ ID        : chr [1:33524] "xuojqdfdozvu" "vvwkinqvnibo" "ibyjcmiopiqli" "lsqamawyancj" ...
$ NPS       : num [1:33524] 8 8 8 8 8 8 4 4 7 ...
$ Field     : Factor w/ 16 levels "LifeSci","PhysSci",...: 15 7 9 14 7 15 15 10 7 7 ...
$ ClassLevel: Ord.factor w/ 4 levels "Fresh"<"Soph"<...: 4 4 4 4 4 4 1 1 3 ...
$ Status    : Factor w/ 3 levels "Full-time","Part-time",...: 1 1 1 1 1 2 2 1 1 1 ...
$ Gender    : Factor w/ 3 levels "Female","Male",...: 2 1 2 1 3 1 2 1 1 2 ...
$ BirthYear : num [1:33524] 1990 1999 1999 1989 1993 ...
$ FinPL     : Ord.factor w/ 2 levels "Yes"<"No": 2 2 2 2 1 1 2 2 2 ...
$ FinSch    : Ord.factor w/ 2 levels "Yes"<"No": 2 2 2 1 1 1 2 2 1 ...
$ FinGov    : Ord.factor w/ 2 levels "Yes"<"No": 2 1 1 NA 2 2 2 2 2 ...
$ FinSelf   : Ord.factor w/ 2 levels "Yes"<"No": 2 1 2 2 1 1 2 1 2 1 ...
$ FinPar    : Ord.factor w/ 2 levels "Yes"<"No": 1 2 NA 1 2 2 2 2 2 1 ...
$ FinOther  : Ord.factor w/ 2 levels "Yes"<"No": 2 2 2 2 2 2 2 2 2 2 ...
$ TooDifficult: Ord.factor w/ 5 levels "Strongly Disagree"<...: 2 1 2 4 NA NA 3 1 5 1 ...
$ NotRelevant: Ord.factor w/ 5 levels "Strongly Disagree"<...: 1 2 NA 1 3 NA 5 NA NA 4 ...
$ PoorTeaching: Ord.factor w/ 5 levels "Strongly Disagree"<...: 4 4 4 2 4 1 4 1 2 5 ...
$ UnsuppFac  : Ord.factor w/ 5 levels "Strongly Disagree"<...: NA 3 3 3 NA 3 1 5 3 NA ...
$ Grades     : Ord.factor w/ 5 levels "Strongly Disagree"<...: NA 2 1 NA NA 4 NA 2 NA 2 ...
$ Sched      : Ord.factor w/ 5 levels "Strongly Disagree"<...: 5 1 1 4 1 5 4 4 4 5 ...
$ ClassTooBig: Ord.factor w/ 5 levels "Strongly Disagree"<...: 3 NA 1 1 5 NA 5 4 NA 2 ...
$ BadAdvising: Ord.factor w/ 5 levels "Strongly Disagree"<...: 2 2 NA 2 1 2 1 NA 2 1 ...
```

```
$ FinAid      : Ord.factor w/ 5 levels "Strongly Disagree"<...: 5 5 5 4 NA NA 5 5 1 NA ...
$ OverallValue: Ord.factor w/ 5 levels "Strongly Disagree"<...: 5 5 3 5 2 5 5 4 NA 2 ...
```

This is another way to validate that it worked. This will show 6 distinct values for each column in the table. It will pad with NA values if it has less than six. The ordered values should be listed in order; the others are essentially random and nothing should be read into it.

```
get_distinct_and_pad <- function(column, numvals = 6) {
  distinct_values <- tibble(value = column) |>
    distinct(value) |>
    arrange(value) |>
    slice_head(n = numvals) |>
    pull(value)

  # Pad with NA if there are less than n values
  length(distinct_values) <- numvals
  return(distinct_values)
}

survey |>
  map(get_distinct_and_pad) |>
  as.data.frame()
```

	Year	ID	NPS	Field	ClassLevel	Status	Gender	BirthYear	FinPL
1	2012	aaapjamvmsgd	4	LifeSci	Fresh	Full-time	Female	1988	Yes
2	2013	aacgcpuqbhue	6	PhysSci	Soph	Part-time	Male	1989	No
3	2014	aacgctmukgti	7	PubHealth	Jr	Other	Other	1990	<NA>
4	2015	aacukkmcqpdr	8	Nurs	Sr	<NA>	<NA>	1991	<NA>
5	2016	aacuutwvloft	NA	OthHealth	<NA>	<NA>	<NA>	1992	<NA>
6	2017	aadtesmehzsd	NA	PubAdm	<NA>	<NA>	<NA>	1993	<NA>
							TooDifficult		NotRelevant
1	Yes	Yes	Yes	Yes	Yes	Strongly Disagree	Strongly Disagree		
2	No	No	No	No	No	Disagree		Disagree	
3	<NA>	<NA>	<NA>	<NA>	<NA>	Neutral		Neutral	
4	<NA>	<NA>	<NA>	<NA>	<NA>	Agree		Agree	
5	<NA>	<NA>	<NA>	<NA>	<NA>	Strongly Agree	Strongly Agree	Agree	
6	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>		<NA>	
		PoorTeaching		UnsuppFac		Grades		Sched	
1	Strongly Disagree								
2	Disagree		Disagree		Disagree		Disagree	Disagree	
3	Neutral		Neutral		Neutral		Neutral	Neutral	
4	Agree		Agree		Agree		Agree	Agree	
5	Strongly Agree		Strongly Agree		Strongly Agree	Strongly Agree	Strongly Agree	Agree	
6	<NA>		<NA>		<NA>	<NA>		<NA>	
	ClassTooBig		BadAdvising			FinAid		OverallValue	

1	Strongly Disagree					
2	Disagree	Disagree	Disagree	Disagree	Disagree	Disagree
3	Neutral	Neutral	Neutral	Neutral	Neutral	Neutral
4	Agree	Agree	Agree	Agree	Agree	Agree
5	Strongly Agree					
6	<NA>	<NA>	<NA>	<NA>	<NA>	<NA>

## 0.6 Convert “wide” to “long” data

R has to have data in “long” format for analysis. The `pivot_longer` function is the tool that allows us to make the transformation from “wide” to “long”.

```
survey <-
  survey |>
  pivot_longer(
    names_to = "Question",
    cols = c(TooDifficult:OverallValue),
    values_to = "Response"
  ) |>
  arrange(Year, ID, Question)
```

## 0.7 Prepare numerical data

When analyzing survey data, it can be helpful to have both a string representation and an integer representation of response data. This statement creates the new `NumResp` column that will hold the numeric response data while retaining the string `Response` column.

```
survey <-
  survey |>
  mutate(NumResp = case_when(
    Response == "Strongly Disagree" ~ 1,
    Response == "Disagree" ~ 2,
    Response == "Neutral" ~ 3,
    Response == "Agree" ~ 4,
    Response == "Strongly Agree" ~ 5,
    .default = NA))
```

## 0.8 Look at data again

Now that we have made all of these transformations, let’s simply print out information about the `survey` table in its new form for documentation.

```
glimpse(survey)
```

Rows: 335,240

Columns: 17

```
$ Year <dbl> 2012, 2012, 2012, 2012, 2012, 2012, 2012, 2012, 2012, 2012, 2012, ~
$ ID <chr> "aacuutwvloft", "aacuutwvloft", "aacuutwvloft", "aacuutwvlof~
$ NPS <dbl> 7, 7, 7, 7, 7, 7, 7, 7, 7, 7, 7, 7, 7, 7, 7, 7, 7, 7, 7, 7, ~
$ Field <fct> Undecl, Undecl, Undecl, Undecl, Undecl, Undecl, Undecl, Undecl, ~
$ ClassLevel <ord> Jr, ~
$ Status <fct> Full-time, Full-time, Full-time, Full-time, Full-time, Full-~
$ Gender <fct> Female, Female, Female, Female, Female, Female, Female, Fem~
$ BirthYear <dbl> 1995, 1995, 1995, 1995, 1995, 1995, 1995, 1995, 1995, 1995, ~
$ FinPL <ord> No, ~
$ FinSch <ord> Yes, ~
$ FinGov <ord> No, ~
$ FinSelf <ord> Yes, ~
$ FinPar <ord> NA, No, No, No, No, No, ~
$ FinOther <ord> No, ~
$ Question <chr> "BadAdvising", "ClassTooBig", "FinAid", "Grades", "NotRelev~
$ Response <ord> NA, Neutral, Agree, NA, NA, Strongly Agree, Strongly Agree, ~
$ NumResp <dbl> NA, 3, 4, NA, NA, 5, 5, 2, 2, 2, 1, 3, 4, 1, NA, 5, NA, 3, ~
```

```
skim(survey)
```

Table 4: Data summary

Name	survey
Number of rows	335240
Number of columns	17
<hr/>	
Column type frequency:	
character	2
factor	11
numeric	4
<hr/>	
Group variables	None

## Variable type: character

skim_variable	n_missing	complete_rate	min	max	empty	n_unique	whitespace
ID	0	1	12	12	0	33524	0
Question	0	1	5	12	0	10	0

## Variable type: factor

	skim_variablen_missing	complete_rate	ordered	n_unique	top_counts
Field	0	1.00	FALSE	16	Soc: 71580, Lif: 63180, Und: 41150, Eng: 36790
ClassLevel	5510	0.98	TRUE	4	Fre: 109270, Sop: 83110, Jr: 71340, Sr: 66010
Status	6270	0.98	FALSE	3	Ful: 251240, Par: 61730, Oth: 16000
Gender	8750	0.97	FALSE	3	Fem: 159880, Mal: 137130, Oth: 29480
FinPL	16740	0.95	TRUE	2	No: 222110, Yes: 96390
FinSch	16150	0.95	TRUE	2	No: 160240, Yes: 158850
FinGov	15860	0.95	TRUE	2	No: 286960, Yes: 32420
FinSelf	16220	0.95	TRUE	2	No: 190670, Yes: 128350
FinPar	16990	0.95	TRUE	2	No: 255370, Yes: 62880
FinOther	16990	0.95	TRUE	2	No: 301970, Yes: 16280
Response	66759	0.80	TRUE	5	Str: 62765, Agr: 61927, Neu: 55456, Dis: 48919

## Variable type: numeric

skim_variable	n_missing	complete_rate	mean	sd	p0	p25	p50	p75	p100	hist
Year	0	1.0	2014.50	1.71	2012	2013	2014	2016	2017	
NPS	0	1.0	5.95	1.52	4	4	6	7	8	
BirthYear	0	1.0	1994.01	3.76	1988	1991	1994	1997	2000	
NumResp	66759	0.8	3.22	1.37	1	2	3	4	5	

## 0.9 Summarize by question

Now that we have the new NumResp column, we can calculate numerical data on the survey responses.

```
survey |>
  group_by(Question) |>
  summarize(Median = median(NumResp, na.rm = TRUE),
             Avg = mean(NumResp, na.rm = TRUE))

# A tibble: 10 x 3
  Question    Median     Avg
  <chr>        <dbl> <dbl>
1 BadAdvising    2    2.33
2 ClassTooBig    2    2.52
3 FinAid         4    3.83
4 Grades          3    3.02
5 NotRelevant    3    2.73
```

```
6 OverallValue      4  4.11
7 PoorTeaching     4  3.42
8 Sched            4  4.00
9 TooDifficult     3  2.99
10 UnsuppFac       3  3.28
```

## 0.10 Remove columns that we do not need

This table is quite large. We can get rid of the ID field if nothing else since we will not be doing any analysis related to the ID of the survey.

```
survey <-
survey |>
  select(Year, NPS, Field, ClassLevel, Status, Gender, BirthYear,
         FinPL, FinSch, FinGov, FinSelf, FinPar, FinOther,
         Question, Response, NumResp)
```

## 0.11 Export data

Export the data but don't include the row number.

```
write.csv(survey, "data/survey-output.csv",
          row.names=FALSE)
```