

# Baby\_faces\_data\_analysis

```
library(tidyverse)
```

```
-- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
v dplyr      1.1.4      v readr      2.1.5
v forcats    1.0.0      v stringr    1.5.1
v ggplot2    3.4.4      v tibble     3.2.1
v lubridate  1.9.3      v tidyr      1.3.1
v purrr      1.0.2
```

```
-- Conflicts ----- tidyverse_conflicts() --
```

```
x dplyr::filter() masks stats::filter()
```

```
x dplyr::lag()     masks stats::lag()
```

```
i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become
```

```
library(dplyr)
```

```
data <- read.csv("~/Documents/2024spring/summer_surf/ICIS_data_analysis/data/BabyFaces2018_
```

```
str(data)
```

```
'data.frame': 2847 obs. of 676 variables:
```

```
$ CHILDID      : int  10004113 10006315 10015113 10016219 10020319 10029114 10030...
$ D1_ID        : int  338833 581936 311838 906834 370839 969833 909835 331933 502...
$ CENTERID     : int  962339 538237 722230 186232 -3 507330 355432 751430 503246 ...
$ CPID         : int  12142114 77262013 38072112 92636118 38404012 30007110 33899...
$ CLASSID      : int  562937 545718 951045 358842 567916 729945 194653 331876 726...
$ PI1_DATA     : num  1 1 1 1 1 1 1 1 1 1 ...
$ PI1_DATE     : chr   "18-Apr" "18-May" "18-Apr" "18-May" ...
$ PR1_DATA     : int  1 0 1 1 0 1 1 1 1 0 ...
$ PR1_DATE     : chr   "18-Mar" " " "18-Apr" "18-May" ...
```

```

$ SR1_DATA          : num  1 1 1 1 0 1 1 1 1 1 ...
$ SR1_DATE          : chr  "18-May" "18-May" "18-May" "18-May" ...
$ T_HV_FLAG         : int  1 2 1 1 2 1 1 1 1 1 ...
$ ISPREGNANT        : int  0 0 0 0 0 0 0 0 0 0 ...
$ CHILDMONTHS       : int  19 31 28 31 25 34 29 25 9 22 ...
$ CHILDSEX          : int  1 2 1 2 1 2 2 1 2 1 ...
$ STRAT             : chr  "M" "M" "M" "M" ...
$ Productive.Vocabulary : int  15 0 88 100 0 71 73 44 8 0 ...
$ Comprehension.Vocabular: int  66 0 99 100 0 83 95 77 4 0 ...
$ PI_WT             : num  62.7 64.3 71.8 39 108 ...
$ PR_WT             : num  50.2 NA 107.7 64.9 NA ...
$ SCR_WT            : num  50.2 56.8 35.9 49.6 NA ...
$ CHILDNC_WT        : num  50.2 56.8 35.9 39 108 ...
$ PI_CLASS_WT       : num  62.7 NA 71.8 39 NA ...
$ SCRPR_CLASS_WT    : num  50.2 NA 35.9 49.6 NA ...
$ PR_CLASS_WT       : num  50.2 NA 107.7 64.9 NA ...
$ SCR_CLASS_WT      : num  50.2 NA 35.9 49.6 NA ...
$ PI_PR_SCR_WT      : num  62.7 64.3 71.8 49.6 NA ...
$ PI_SCR_WT         : num  62.7 64.3 71.8 49.6 NA ...
$ PI_PR_WT          : num  62.7 NA 107.7 66.1 NA ...
$ PIPRSCRCLASSTEACHER_WT : num  62.7 NA 71.8 49.6 NA ...
$ PIPRSCR_HV_WT     : num  NA 64.3 NA NA NA ...
$ PI1SERVTYPE       : int  1 2 1 3 2 1 1 1 1 2 ...
$ PI1G04REV         : int  5 -3 5 4 -3 5 5 5 5 -3 ...
$ PI1R_ID           : int  1 3 1 1 1 1 1 1 3 1 ...
$ PI1P1_ID          : int  3 1 -3 -3 3 -3 6 -3 -3 -3 ...
$ PI1P2_ID          : int  -3 -3 -3 -3 -3 -3 -3 -3 -3 -3 ...
$ PI1R_AGE          : int  3 6 5 6 6 3 4 4 5 5 ...
$ PI1M_AGE          : int  3 6 5 6 6 3 4 4 5 5 ...
$ PI1F_AGE          : int  -9 6 -9 -9 -9 -9 -9 -9 5 -9 ...
$ PI1M_AGE_FSTBC    : int  2 4 3 3 6 1 3 3 4 1 ...
$ PI1R_RACE         : int  2 6 3 1 2 3 1 3 3 3 ...
$ PI1M_RACE         : int  2 6 3 1 2 3 1 3 -9 3 ...
$ PI1F_RACE         : int  2 6 -9 -9 2 -9 -9 -9 3 -9 ...
$ PI1R_YRS_USC      : int  -3 -3 -3 -3 -3 -3 -3 1 3 3 ...
$ PI1M_YRS_USC      : int  -3 3 -3 -3 -3 -3 -3 1 -3 3 ...
$ PI1F_YRS_USC      : int  -3 -3 -3 -3 -3 -3 -3 -3 3 -3 ...
$ PI1R_EDUC         : int  2 5 5 2 2 6 2 2 5 2 ...
$ PI1M_EDUC         : int  2 4 5 2 2 6 2 2 5 2 ...
$ PI1F_EDUC         : int  2 5 3 -8 7 1 1 -8 5 1 ...
$ PI1R_EMPTY        : int  1 1 2 1 1 5 5 1 1 1 ...
$ PI1M_EMPTY        : int  1 5 2 1 1 5 5 1 -9 1 ...
$ PI1F_EMPTY        : int  5 1 -9 -9 1 -9 -9 -9 1 -9 ...

```

\$ PI1C_BIRTHWEIGHTC	: int	1 2 1 1 2 1 2 1 1 1 ...
\$ PI1C_RACE	: int	2 6 3 1 2 3 1 -9 3 3 ...
\$ PI1C_RACE4	: int	2 4 3 1 2 3 1 -9 3 3 ...
\$ PI1HH_BIO	: int	1 1 2 2 1 2 2 2 3 2 ...
\$ PI1C_LNG_SPKTO	: int	1 1 3 1 1 3 1 2 5 3 ...
\$ PI1C_LNG_SPKBY	: int	1 1 1 1 1 3 1 2 -3 -3 ...
\$ PI1C_LNG_SPKTO_SP	: int	-3 -3 2 -3 -3 2 -3 1 -3 2 ...
\$ PI1C_LNG_SPKTO_OTH	: int	-3 -3 -3 -3 -3 -3 -3 -3 2 -3 ...
\$ PI1P_IMGRNT	: int	0 1 0 0 0 0 0 1 1 1 ...
\$ PI1M_PREG	: int	0 0 0 0 0 0 0 0 0 0 ...
\$ PI1R_MALE	: int	0 1 0 0 0 0 0 0 1 0 ...
\$ PI1P1_MALE	: int	1 0 -3 -3 1 -3 1 -3 -3 -3 ...
\$ PI1P2_MALE	: int	-3 -3 -3 -3 -3 -3 -3 -3 -3 -3 ...
\$ PI1R_AGE	: int	20 36 33 38 41 21 26 26 31 33 ...
\$ PI1M_AGE	: int	20 38 33 38 41 21 26 26 32 33 ...
\$ PI1F_AGE	: int	-9 36 -9 -9 -9 -9 -9 -9 31 -9 ...
\$ PI1M_AGE_FCB	: int	19 35 31 36 39 18 23 24 31 31 ...
\$ PI1F_AGE_FCB	: int	-9 33 -9 -9 -9 -9 -9 -9 31 -9 ...
\$ PI1M_AGE_FSTB	: int	18 25 22 22 36 17 23 23 25 17 ...
\$ PI1R_BORNINUS	: int	1 1 1 1 1 1 1 0 0 0 ...
\$ PI1M_BORNINUS	: int	1 0 1 1 1 1 1 0 -9 0 ...
\$ PI1F_BORNINUS	: int	1 1 -9 -9 1 -9 -9 -9 0 -9 ...
\$ PI1M_YRS_US	: int	-3 17 -3 -3 -3 -3 -3 5 -3 32 ...
\$ PI1F_YRS_US	: int	-3 -3 -3 -3 -3 -3 -3 -3 22 -3 ...
\$ PI1M_CLASS	: int	0 0 1 0 1 1 1 1 -9 0 ...
\$ PI1F_CLASS	: int	-9 0 -9 -9 -9 -9 -9 -9 0 -9 ...
\$ PI1M_EMPTY12	: int	-3 0 -3 -3 -3 1 1 -3 -9 -3 ...
\$ PI1F_EMPTY12	: int	0 -3 -9 -9 -3 -9 -9 -9 -3 -9 ...
\$ PI1M_JOB_TR	: int	0 0 1 0 0 0 0 0 -9 0 ...
\$ PI1F_JOB_TR	: int	-9 0 -9 -9 -9 -9 -9 -9 0 -9 ...
\$ PI1C_MALE	: int	1 0 1 0 1 0 0 1 0 1 ...
\$ PI1CAGE_MTH	: num	19.9 31.5 29.2 31.6 26 34.5 29.8 26.1 9.8 25.2 ...
\$ PI1CAGE_MTHC	: int	2 3 3 3 3 3 3 3 1 3 ...
\$ PI1C_WEEKSEARLY	: int	2 2 -3 -3 4 -3 5 -3 -3 -3 ...
\$ PI1CBPREM	: int	0 0 0 0 1 0 1 0 0 0 ...
\$ PI1C_BIRTHWEIGHT	: num	7.81 4.31 8.81 -9 5.31 ...
\$ PI1C_USBORN	: int	1 1 1 1 1 1 1 1 1 1 ...
\$ PI1HH_NEMPTY	: int	1 1 1 1 2 0 1 1 1 1 ...
\$ PI1HH_COMP	: int	3 4 4 4 4 3 3 5 4 6 ...
\$ PI1HH_KIDS	: int	1 2 3 3 2 1 2 2 3 5 ...
\$ PI1HH_ADULTS	: int	2 2 1 1 2 2 1 3 1 1 ...
\$ PI1P1_NUMPAR	: int	2 2 1 1 2 1 2 1 1 1 ...
\$ PI1HH_PRNT	: int	2 2 1 1 2 1 1 1 1 1 ...

```
$ PI1MF_MAR          : int  0 1 1 0 0 0 0 0 0 0 ...
$ PI1F_LV_IN         : int  1 1 0 0 1 0 0 0 1 0 ...
$ PI1HH_IG           : int  0 0 0 0 0 0 0 0 0 0 ...
$ PI1HHSTRUC         : int  2 1 4 4 2 4 3 4 5 4 ...
[list output truncated]
```

```
data_cleaned<-data|>
  filter(PI1CESDRT >= 0 )|>
  filter(PR1BITSPR >=0)|>
  filter(PR1BITSCR>= 0)
```

## problem behaviors

```
model1 <- lm(PR1BITSPR~PI1CESDRT, data = data_cleaned)
summary(model1)
```

Call:

```
lm(formula = PR1BITSPR ~ PI1CESDRT, data = data_cleaned)
```

Residuals:

Min	1Q	Median	3Q	Max
-14.287	-4.857	-1.027	3.322	45.096

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	10.02746	0.18175	55.172	< 2e-16 ***
PI1CESDRT	0.16919	0.02106	8.034	1.67e-15 ***

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 6.59 on 1834 degrees of freedom

Multiple R-squared: 0.034, Adjusted R-squared: 0.03347

F-statistic: 64.54 on 1 and 1834 DF, p-value: 1.674e-15

The model's slope coefficient is 0.16919, and the p-value is 1.67e-15. Since this p-value is smaller than the significance level of 0.05, it indicates that there is significant relationship between parental depression and the problem behaviors in children.

## social competence

```
model3 <- lm(PR1BITSCR~PI1CESDRT, data = data_cleaned)
summary(model3)
```

Call:

```
lm(formula = PR1BITSCR ~ PI1CESDRT, data = data_cleaned)
```

Residuals:

	Min	1Q	Median	3Q	Max
	-13.8396	-1.8051	0.4711	2.2390	4.9199

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	17.83963	0.08639	206.507	< 2e-16 ***
PI1CESDRT	-0.03453	0.01001	-3.449	0.000575 ***

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 3.132 on 1834 degrees of freedom

Multiple R-squared: 0.006445, Adjusted R-squared: 0.005903

F-statistic: 11.9 on 1 and 1834 DF, p-value: 0.0005751

The model's slope coefficient is -0.035, and the p-value is 0.000575. Since this p-value is smaller than the significance level of 0.05, it indicates that there is significant relationship between parental depression and the social competence in children.

## language skills

### comprehensive vocabulary

```
model4 <- lm(Comprehension.Vocabular~PI1CESDRT, data = data_cleaned)
summary(model4)
```

Call:

```
lm(formula = Comprehension.Vocabular ~ PI1CESDRT, data = data_cleaned)
```

Residuals:

Min	1Q	Median	3Q	Max
-61.68	-23.68	10.38	32.32	40.78

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	61.67879	0.98690	62.50	<2e-16 ***
PI1CESDRT	-0.06066	0.11435	-0.53	0.596

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 35.78 on 1834 degrees of freedom

Multiple R-squared: 0.0001534, Adjusted R-squared: -0.0003918

F-statistic: 0.2814 on 1 and 1834 DF, p-value: 0.5959

The model's slope coefficient is -0.2151, and the p-value is 0.0522. Since this p-value is greater than the significance level of 0.05, it indicates that there is no significant relationship between parental depression and the comprehensive vocabulary development in children.

## productive vocabulary

```
model5 <- lm(Productive.Vocabulary~PI1CESDRT, data = data_cleaned)
summary(model5)
```

Call:

```
lm(formula = Productive.Vocabulary ~ PI1CESDRT, data = data_cleaned)
```

Residuals:

Min	1Q	Median	3Q	Max
-43.015	-30.152	-6.368	28.271	64.160

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	43.0154	0.9068	47.435	<2e-16 ***
PI1CESDRT	-0.1434	0.1051	-1.365	0.172

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 32.88 on 1834 degrees of freedom  
Multiple R-squared: 0.001015, Adjusted R-squared: 0.0004702  
F-statistic: 1.863 on 1 and 1834 DF, p-value: 0.1724

The model's slope coefficient is -0.2084, and the p-value is 0.0267. Since this p-value is smaller than the significance level of 0.05, it indicates that there is significant relationship between parental depression and the productive vocabulary development in children.