

# Milestone2

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## Maternal Risk

### Check for potential data quality issues

```
maternal_risk <- read_csv("MaternalHealthRiskDataSet.csv")

## Rows: 1014 Columns: 7
## -- Column specification -----
## Delimiter: ","
## chr (1): RiskLevel
## dbl (6): Age, SystolicBP, DiastolicBP, BS, BodyTemp, HeartRate
##
## i Use `spec()` to retrieve the full column specification for this data.
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.
# No missing values
sum(maternal_risk %>%
  is.na())

## [1] 0
```

Variable definitions from Kaggle: - Age: Age in years when a woman is pregnant. - SystolicBP: Upper value of Blood Pressure in mmHg, another significant attribute during pregnancy. - DiastolicBP: Lower value of Blood Pressure in mmHg, another significant attribute during pregnancy. - BS: Blood glucose levels is in terms of a molar concentration, mmol/L. - HeartRate: A normal resting heart rate in beats per minute. - Risk Level: Predicted Risk Intensity Level during pregnancy considering the previous attribute.

```
summary(maternal_risk)
```

##	Age	SystolicBP	DiastolicBP	BS
##	Min. :10.00	Min. : 70.0	Min. : 49.00	Min. : 6.000
##	1st Qu.:19.00	1st Qu.:100.0	1st Qu.: 65.00	1st Qu.: 6.900
##	Median :26.00	Median :120.0	Median : 80.00	Median : 7.500
##	Mean :29.87	Mean :113.2	Mean : 76.46	Mean : 8.726
##	3rd Qu.:39.00	3rd Qu.:120.0	3rd Qu.: 90.00	3rd Qu.: 8.000
##	Max. :70.00	Max. :160.0	Max. :100.00	Max. :19.000
##	BodyTemp	HeartRate	RiskLevel	
##	Min. : 98.00	Min. : 7.0	Length:1014	
##	1st Qu.: 98.00	1st Qu.:70.0	Class :character	
##	Median : 98.00	Median :76.0	Mode :character	
##	Mean : 98.67	Mean :74.3		
##	3rd Qu.: 98.00	3rd Qu.:80.0		
##	Max. :103.00	Max. :90.0		

HeartRate seems to have at least one outlier, as the minimum is only 7 but the majority of the points are in the 70-80 bpm range.

### Check for duplicate entries

```
maternal_risk %>%  
  distinct()
```

```
## # A tibble: 452 x 7  
##       Age SystolicBP DiastolicBP      BS BodyTemp HeartRate RiskLevel  
##   <dbl>      <dbl>      <dbl> <dbl>    <dbl>    <dbl> <chr>  
## 1    25         130         80  15      98      86 high risk  
## 2    35         140         90  13      98      70 high risk  
## 3    29          90         70   8     100     80 high risk  
## 4    30         140         85   7      98      70 high risk  
## 5    35         120         60  6.1     98      76 low risk  
## 6    23         140         80  7.01    98      70 high risk  
## 7    23         130         70  7.01    98      78 mid risk  
## 8    35          85         60  11     102     86 high risk  
## 9    32         120         90  6.9     98      70 mid risk  
## 10   42         130         80  18      98      70 high risk  
## # ... with 442 more rows
```

There appear to be many duplicates, as `distinct()` reduced the number of rows from 1014 to just 452. It does seem that the duplicate entries still represent different people and they just happened to have the same measurements, so these entries should still be included as we proceed with our analysis.

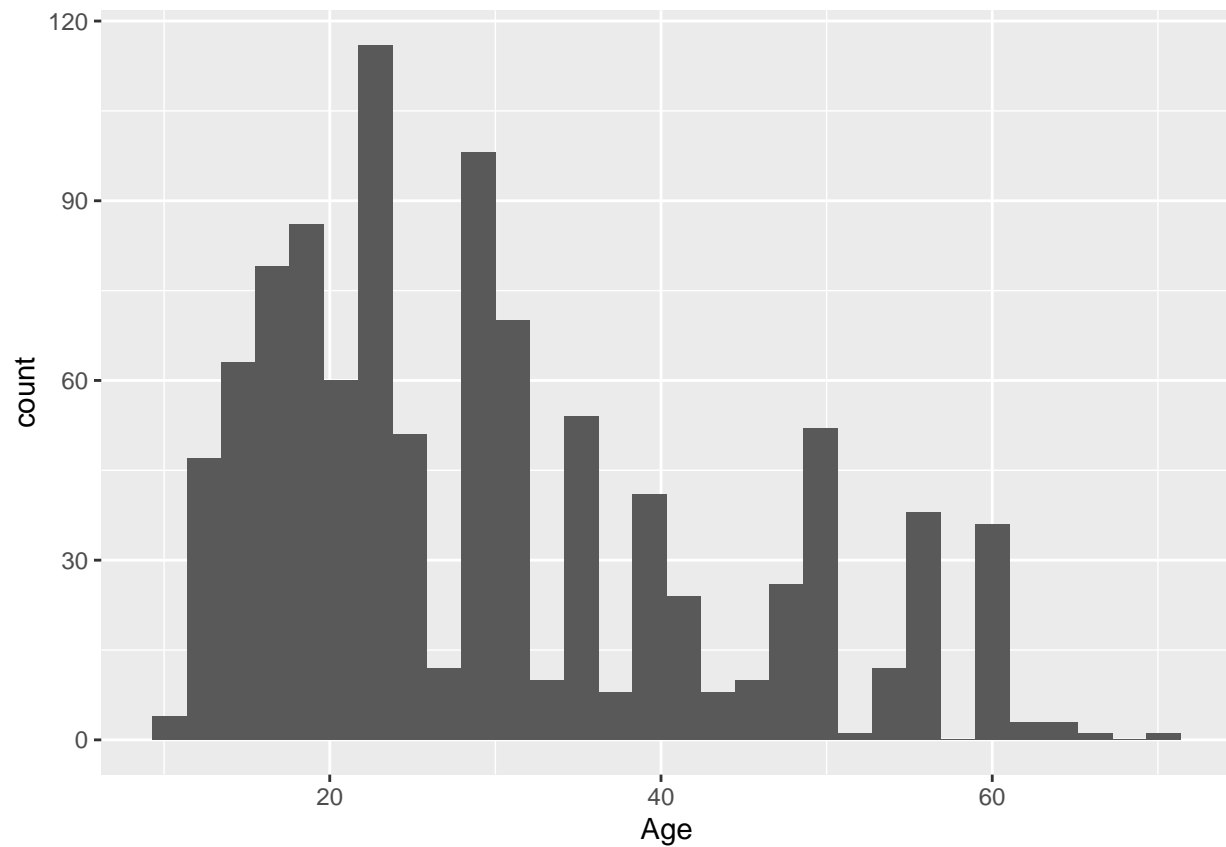
### Build intuition about individual variables

```
histogram <- function(colname) {  
  maternal_risk %>%  
    ggplot() +  
    geom_histogram(aes(x = .data[[colname]]))  
}
```

#### Age

```
histogram("Age")
```

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```

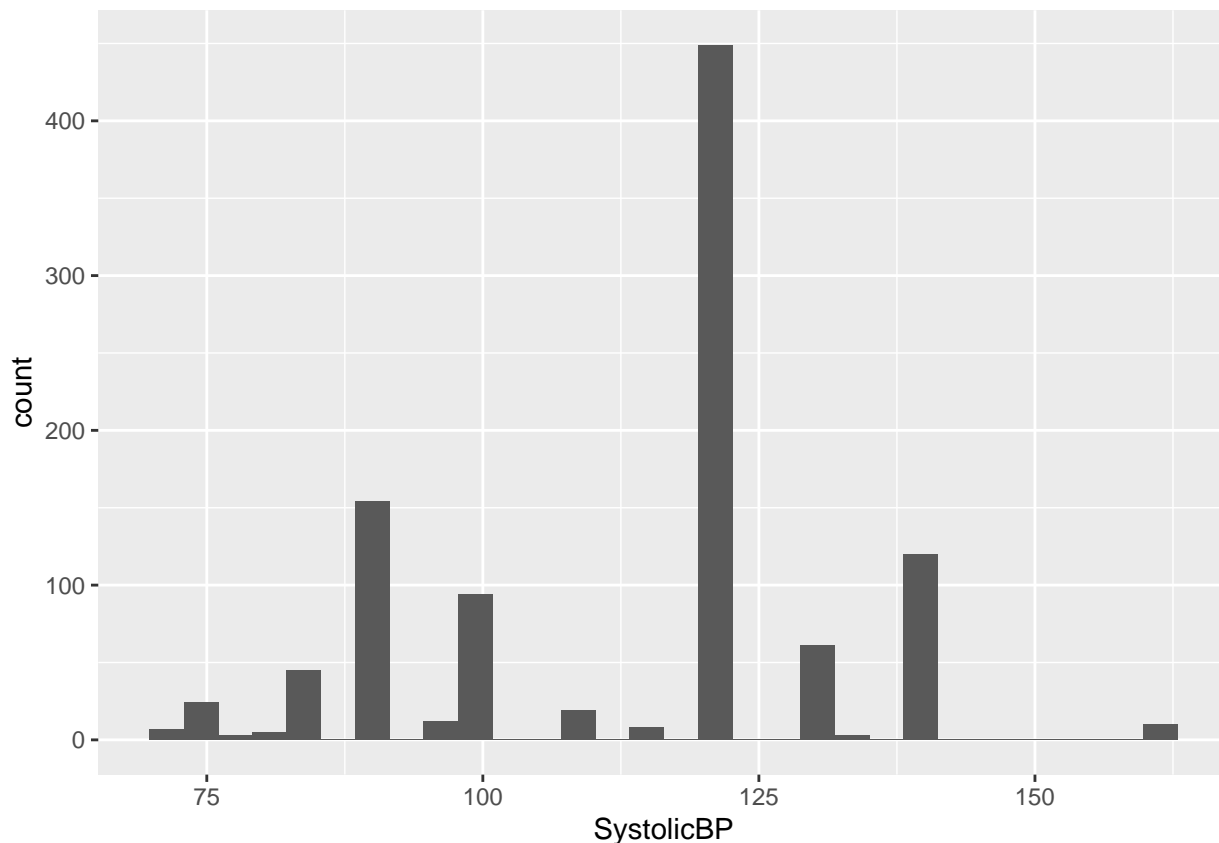


Age ranges from 10 to 70, with the majority between 15 and 25, which is expected.

### SystolicBP

```
histogram("SystolicBP")
```

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```



Small number of entries with SystolicBP above 150. What are their risk levels?

```
maternal_risk %>%
  filter(SystolicBP > 150)
```

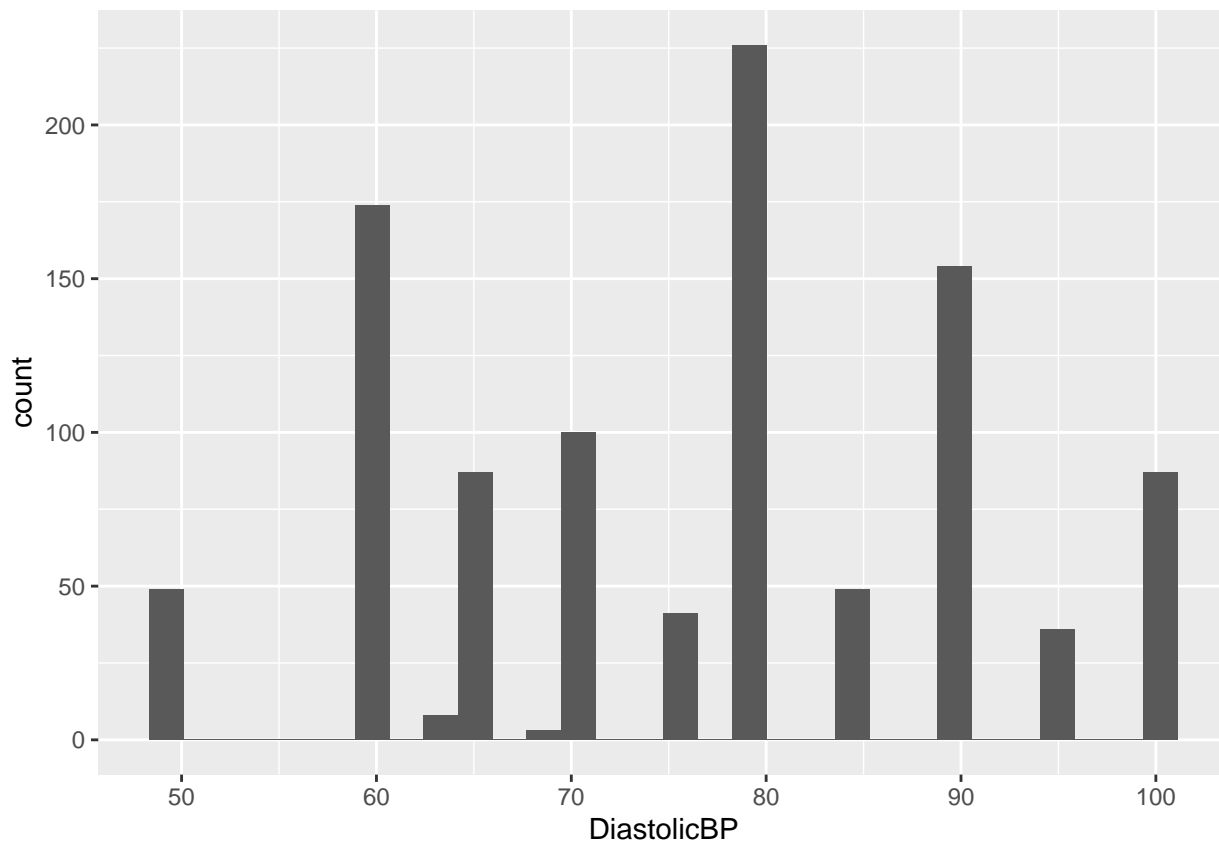
```
## # A tibble: 10 x 7
##   Age SystolicBP DiastolicBP BS BodyTemp HeartRate RiskLevel
##   <dbl>      <dbl>      <dbl> <dbl>    <dbl>      <dbl> <chr>
## 1  40        160        100    19      98        77 high risk
## 2  40        160        100    19      98        77 high risk
## 3  40        160        100    19      98        77 high risk
## 4  40        160        100    19      98        77 high risk
## 5  40        160        100    19      98        77 high risk
## 6  40        160        100    19      98        77 high risk
## 7  40        160        100    19      98        77 high risk
## 8  40        160        100    19      98        77 high risk
## 9  40        160        100    19      98        77 high risk
## 10 40        160        100    19      98        77 high risk
```

All ten rows are identical and classify each as high risk, which is expected given that these entries are outliers.

## DiastolicBP

```
histogram("DiastolicBP")
```

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```



Fairly evenly distributed. Check under 50 group.

```
maternal_risk %>%
  filter(DiastolicBP < 50)
```

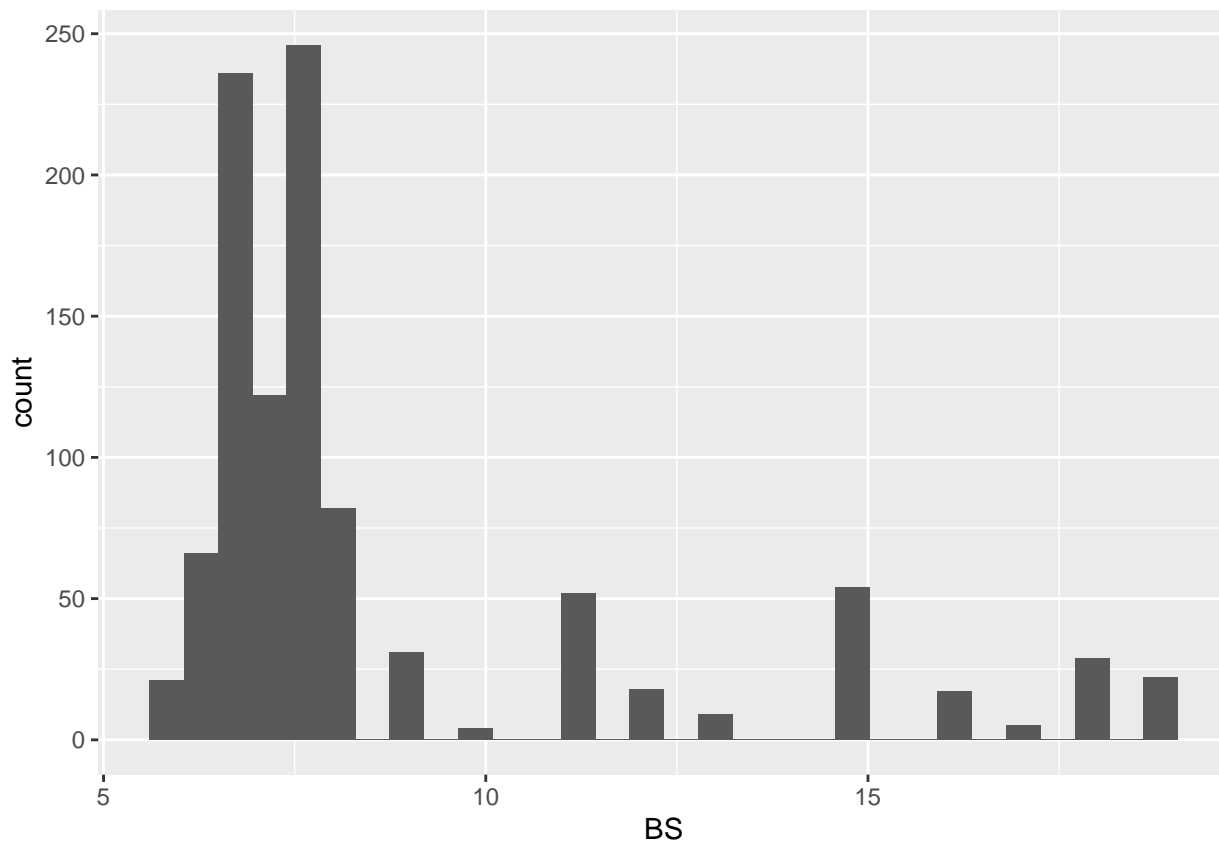
```
## # A tibble: 25 x 7
##   Age SystolicBP DiastolicBP   BS BodyTemp HeartRate RiskLevel
##   <dbl>      <dbl>      <dbl> <dbl>   <dbl>      <dbl> <chr>
## 1    15         76         49  7.5     98         77 low risk
## 2    15         76         49  6.4     98         77 low risk
## 3    15         76         49  7.5     98         77 low risk
## 4    15         76         49  6.4     98         77 low risk
## 5    15         75         49  7.7     98         77 low risk
## 6    15         76         49  7.8     98         77 low risk
## 7    15         76         49  6.8     98         77 low risk
## 8    15         76         49  6.8     98         77 low risk
## 9    15         76         49  7.9     98         77 low risk
## 10   15         76         49  7.9     98         77 low risk
## # ... with 15 more rows
```

25 entries with Diastolic blood pressure less than 50 (all 49). All are classified as low risk, so upon initial exploration, these low DiastolicBP outliers don't seem to increase risk level.

## Blood Sugar

```
histogram("BS")
```

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```



The majority are under 10, but there are many entries spread between 10 and 20. Do these have higher risk level?

```
maternal_risk %>%
  filter(BS > 10) %>%
  group_by(RiskLevel) %>%
  summarise(count = n())
```

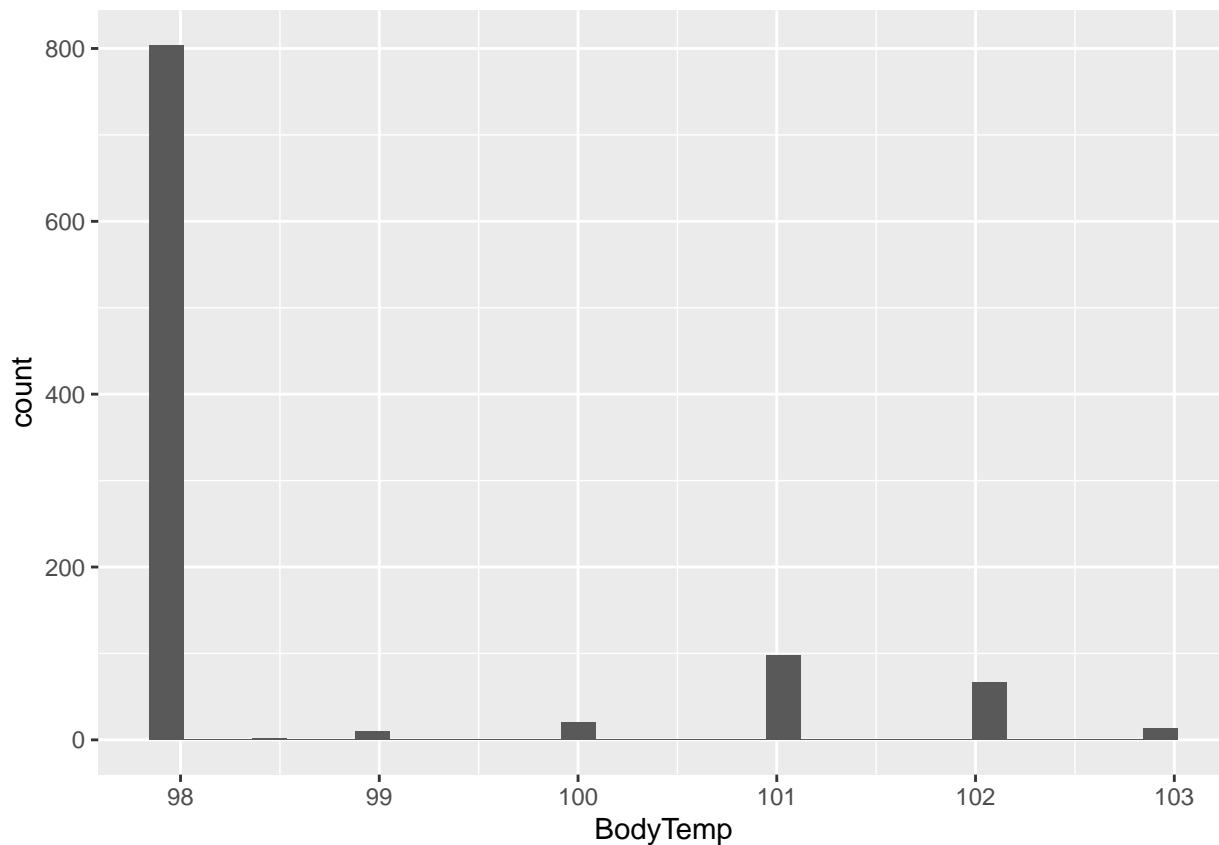
```
## # A tibble: 3 x 2
##   RiskLevel count
##   <chr>      <int>
## 1 high risk   170
## 2 low risk     4
## 3 mid risk    32
```

Yes, the majority of the entries with a blood sugar level greater than 10 are classified as high risk. This is a factor we should explore further, as this seems to indicate that high values of BP correlate with higher risk level.

## Body Temperature

```
histogram("BodyTemp")
```

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```



The majority are in the normal temperature range (~98-99).

```
maternal_risk %>%
  filter(BodyTemp > 100) %>%
  group_by(RiskLevel) %>%
  summarise(count = n())
```

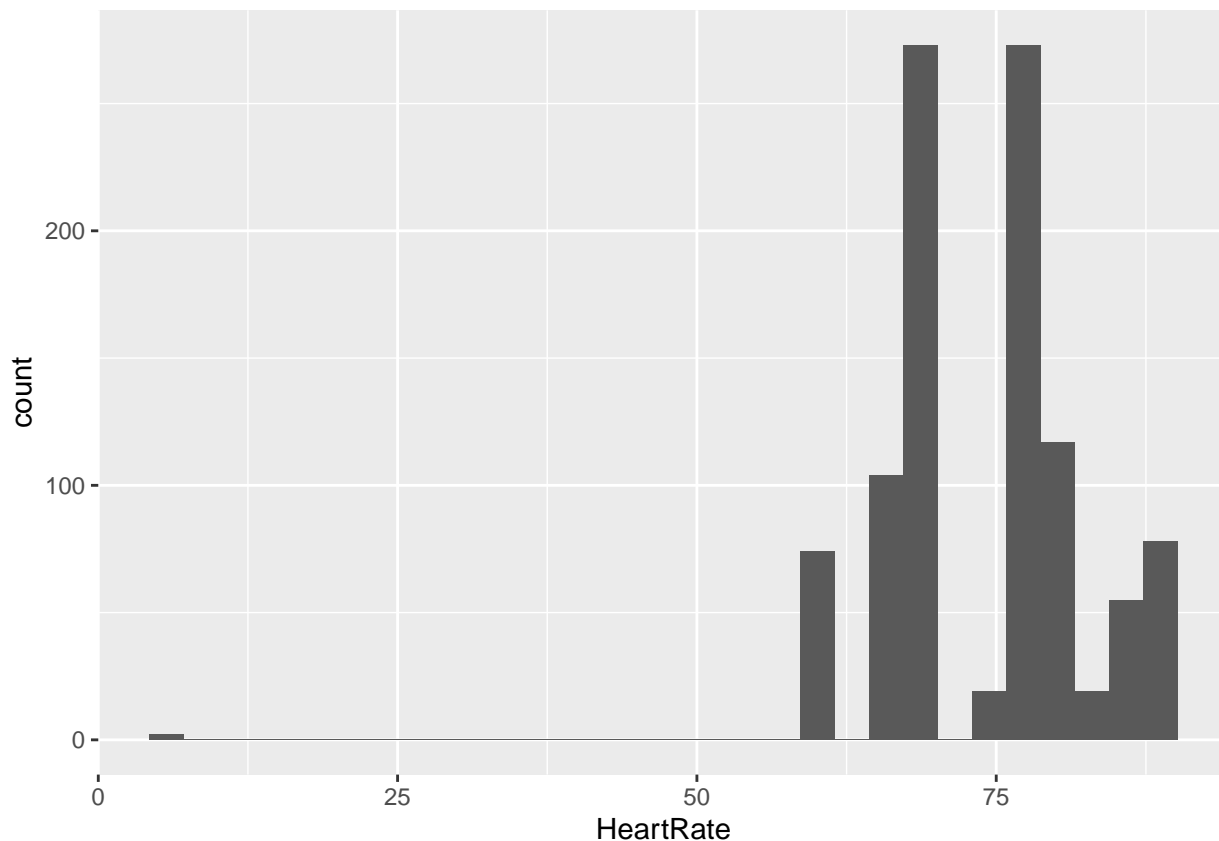
```
## # A tibble: 3 x 2
##   RiskLevel count
##   <chr>      <int>
## 1 high risk    65
## 2 low risk     37
## 3 mid risk     75
```

There doesn't seem to be a super strong correlation, but there are more mid to high risk entries than low risk in the group of higher body temperatures.

## HeartRate

```
histogram("HeartRate")
```

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```



There is a major outlier, with heart rate less than 10.

```
maternal_risk %>%
  filter(HeartRate < 25)
```

```
## # A tibble: 2 x 7
##   Age SystolicBP DiastolicBP   BS BodyTemp HeartRate RiskLevel
##   <dbl>      <dbl>      <dbl> <dbl>   <dbl>    <dbl> <chr>
## 1    16         120         75   7.9     98         7 low risk
## 2    16         120         75   7.9     98         7 low risk
```

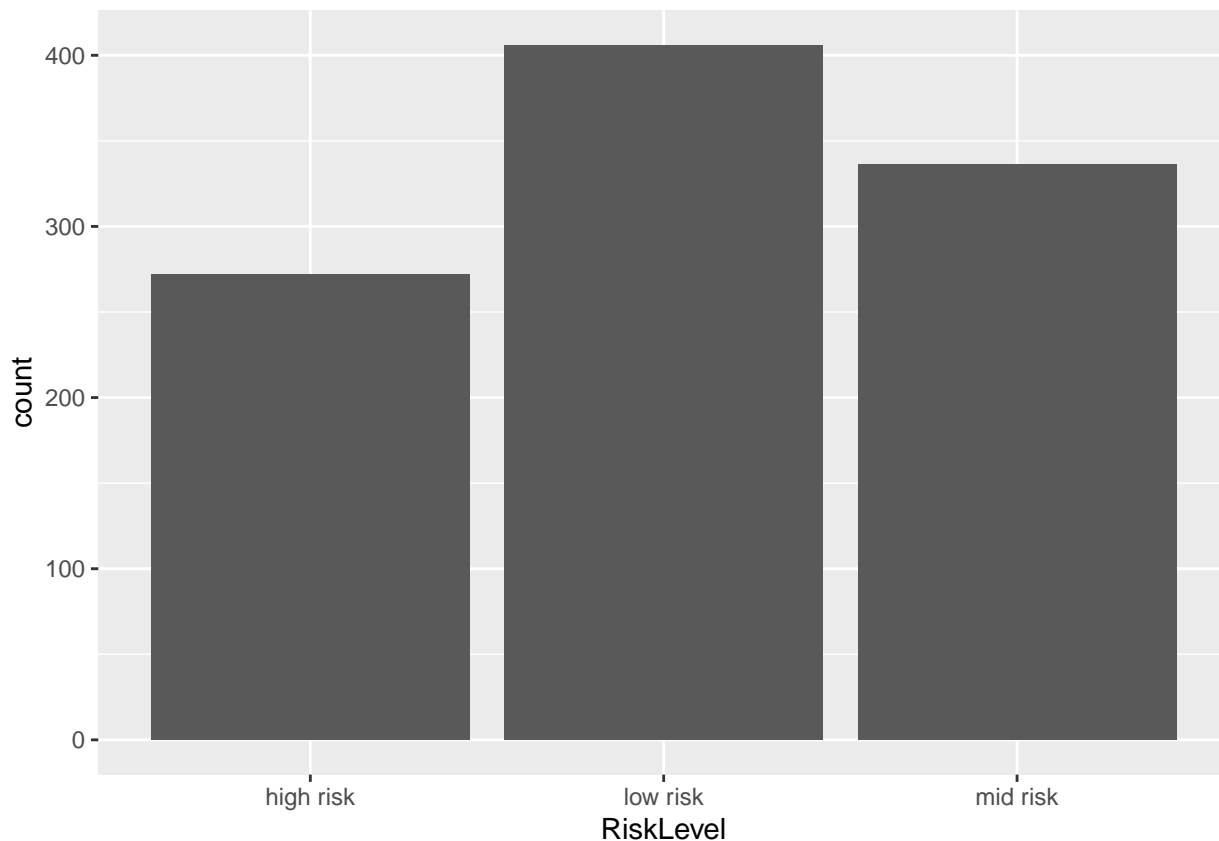
HeartRate may have been entered incorrectly? This person is classified as low risk but has a heart rate of just 7, while the majority of heart rates in this dataset are around 70 bpm.

### Risk Level

```
# Plot discrete histogram (RiskLevel)
maternal_risk %>%
  ggplot() +
  geom_histogram(aes(x = RiskLevel), stat = "count")
```

```
## Warning: Ignoring unknown parameters: binwidth, bins, pad
```





Risk level mainly low but fairly evenly dispersed.

```
maternal_risk %>%  
  count(RiskLevel)
```

```
## # A tibble: 3 x 2  
##   RiskLevel     n  
##   <chr>      <int>  
## 1 high risk   272  
## 2 low risk    406  
## 3 mid risk   336
```

The order of the levels isn't intuitive - modify this in next step.

## Prepare processed data

RiskLevel in order that doesn't make sense for plotting. Update factor to have levels increase from low to high.

```
risks <- (c("low risk", "mid risk", "high risk"))  
  
maternal_risk2 <- maternal_risk %>%  
  mutate(RiskLevel = factor(RiskLevel, levels = risks))
```

# Global Mortality

## Check for potential data quality issues

```
global <- read_csv("global_mortality.csv")

## Rows: 6156 Columns: 35
## -- Column specification -----
## Delimiter: ","
## chr (2): country, country_code
## dbl (33): year, Cardiovascular diseases (%), Cancers (%), Respiratory diseases...
##
## i Use `spec()` to retrieve the full column specification for this data.
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.
# no missing values in data we care about
sum(global %>%
  select(c(`Neonatal deaths (%)`, `Maternal deaths (%)`)) %>%
  is.na())
```

```
## [1] 0
```

Key Variables: - Maternal deaths (%): Percentage of all deaths in a given country & year where the cause was linked to maternal/childbirth complications - Neonatal deaths (%): Percentage of all deaths in a given country & year where death occurred within the first 28 days of life

```
summary(global[c(1, 10, 25)])
```

##	country	Neonatal deaths (%)	Maternal deaths (%)
##	Length:6156	Min. : 0.04071	Min. :0.00188
##	Class :character	1st Qu.: 0.68559	1st Qu.:0.03230
##	Mode :character	Median : 3.89183	Median :0.23696
##		Mean : 4.56666	Mean :0.58591
##		3rd Qu.: 7.74003	3rd Qu.:1.00166
##		Max. :17.80683	Max. :3.41435

## Check for duplicate entries

```
# same size as original dataset, no duplicates
nrow(global %>% distinct())
```

```
## [1] 6156
```

```
length(global %>% distinct())
```

```
## [1] 35
```

## Prepare processed data

```
# keep most recent year
global <- global %>%
  filter(year == 2016)
```

## Build intuition about individual variables

```
global <- read_csv("global_mortality.csv")
```

```

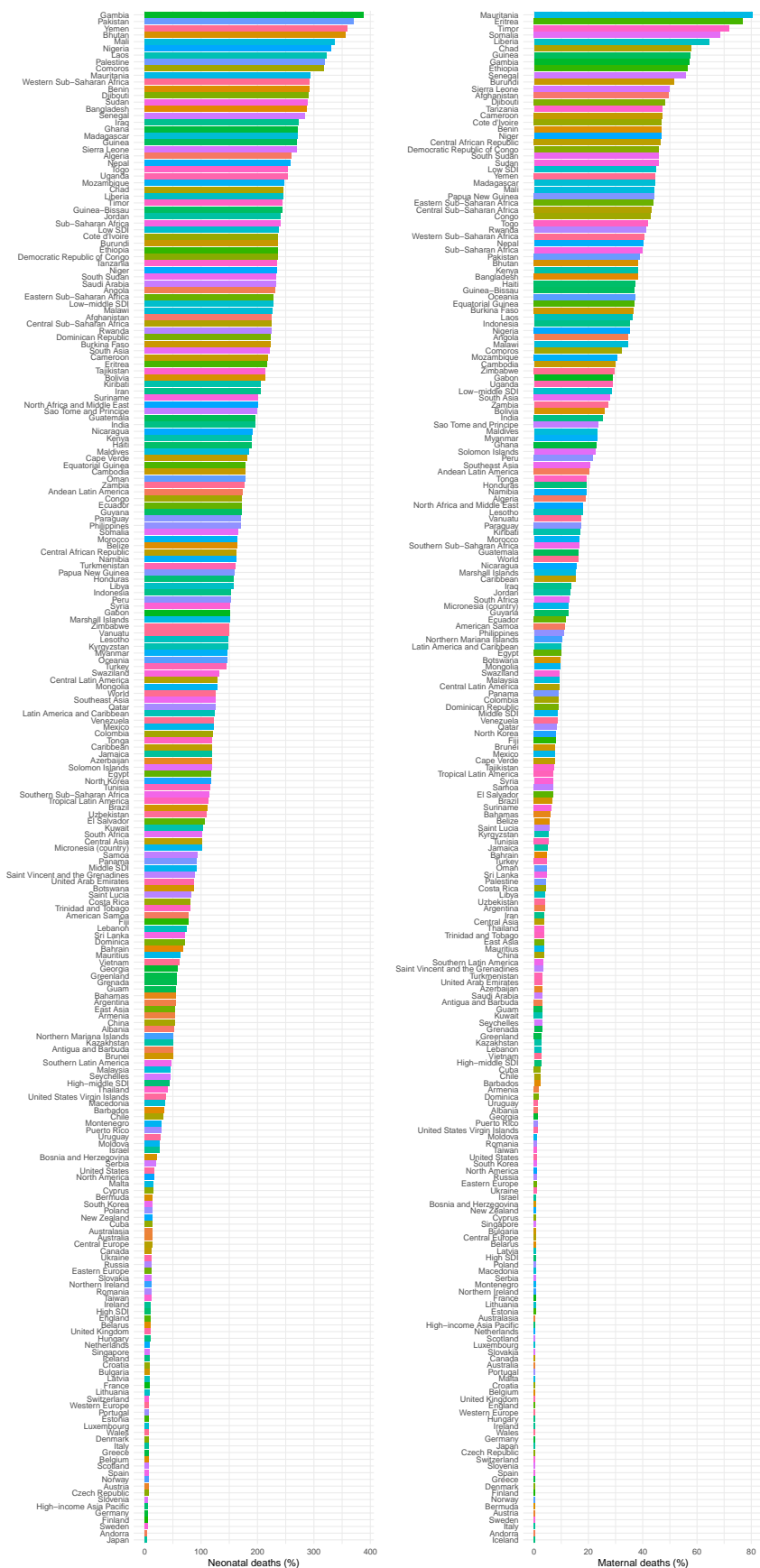
## Rows: 6156 Columns: 35
## -- Column specification -----
## Delimiter: ","
## chr (2): country, country_code
## dbl (33): year, Cardiovascular diseases (%), Cancers (%), Respiratory diseases...
##
## i Use `spec()` to retrieve the full column specification for this data.
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.

countries1 <- ggplot(global) +
  geom_col(aes(`Neonatal deaths (%)`, reorder(country, `Neonatal deaths (%)`), fill = country)) +
  ylab(label = "") +
  theme(axis.text = element_text(size = 2)) +
  theme_minimal() +
  guides(fill="none")

countries2 <- ggplot(global) +
  geom_col(aes(`Maternal deaths (%)`, reorder(country, `Maternal deaths (%)`), fill = country)) +
  ylab(label = "") +
  theme(axis.text = element_text(size = 2)) +
  theme_minimal() +
  guides(fill="none")

countries1 + countries2

```



## Interventions & Maternal Outcomes

### Check for potential data quality issues

The following .csv files were hand-generated from the cited paper (more details in the write-up).

```
interventions <- read_csv("interventions.csv")

## Rows: 3 Columns: 14
## -- Column specification -----
## Delimiter: ","
## chr (2): setting, caregiver
## dbl (12): external_tocometer, fetal_scalp_electrode, amniotomy, oxytocin, ni...
##
## i Use `spec()` to retrieve the full column specification for this data.
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.
outcomes <- read_csv("maternal_outcomes.csv")

## Rows: 3 Columns: 20
## -- Column specification -----
## Delimiter: ","
## chr (2): setting, caregiver
## dbl (18): prolapsed_cord, uterine_rupture, postpartum_hemorrhage, blood_tran...
##
## i Use `spec()` to retrieve the full column specification for this data.
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.
# no missing values
sum(interventions %>%
  is.na())

## [1] 0

sum(outcomes %>%
  is.na())

## [1] 0
```

“interventions” Variables:

Number of pregnant people who had...

- external\_tocometer: external electronic fetal monitoring
- fetal\_scalp\_electrode: internal electronic fetal monitoring
- amniotomy: purposeful puncturing of the amniotic sac
- oxytocin: drug used to induce labor, brand name Pitocin
- nitrous\_oxide: “laughing gas”
- epidural: numbing prodeduce for the lower part of the body
- narcotic: pain-relieving drugs
- spontaneous\_vaginal: vaginal birth without forceps or other mechanical interventions
- assisted\_vaginal: vaginal birth with mechanical assistance
- cesarean: birth via cesarean-section surgery
- episiotomy: cutting of the perineum for vaginal birth

total: total number in each birth setting/caregiver (same as below)

“outcomes” variables:

Number of pregnant people who experienced...

- prolapsed\_cord: a prolapsed umbilical cord
- uterine\_rupture: a rupture of the uterus
- postpartum\_hemorrhage: hemorrhaging after giving birth
- blood\_transfusion: requiring a blood transfusion
- obstetric\_shock: shock, usually caused by excessive blood loss
- death
- placenta\_removal: manual removal of the placenta (rather than natural expulsion)
- uterine\_prolapse: a prolapsed uterus
- pyrexia: fever
- uti: urinary tract infection
- puerpural\_fever: prolonged fever immediately after giving birth
- wound\_infection: infection at the site of a wound
- no\_tear: no tearing from birth
- first\_second\_degree: first or second degree tears
- third\_fourth\_degree: third or fourth degree tears
- unknown\_degree: tearing of unknown degree
- cervical\_tear: tearing of the cervix

total: total number in each birth setting/caregiver (same as above)

```
summary(interventions)
```

```
##      setting      caregiver      external_tocometer fetal_scalp_electrode
## Length:3      Length:3      Min.   : 389      Min.   : 60.0
## Class :character Class :character 1st Qu.:1180      1st Qu.:153.5
## Mode  :character Mode  :character Median :1970      Median :247.0
##                                     Mean  :2174      Mean  :285.0
##                                     3rd Qu.:3067      3rd Qu.:397.5
##                                     Max.   :4164      Max.   :548.0
##      amniotomy      oxytocin      nitrous_oxide      epidural
## Min.   : 560      Min.   :172.0      Min.   : 199      Min.   : 224.0
## 1st Qu.:1039      1st Qu.:387.5      1st Qu.: 882      1st Qu.: 562.5
## Median :1518      Median :603.0      Median :1565      Median : 901.0
## Mean   :1397      Mean   :585.3      Mean   :1550      Mean   : 870.7
## 3rd Qu.:1815      3rd Qu.:792.0      3rd Qu.:2226      3rd Qu.:1194.0
## Max.   :2112      Max.   :981.0      Max.   :2887      Max.   :1487.0
##      narcotic      spontaneous_vaginal assisted_vaginal      cesarean
## Min.   : 122.0      Min.   :2605      Min.   : 86.0      Min.   :208.0
## 1st Qu.: 417.5      1st Qu.:3258      1st Qu.:215.0      1st Qu.:353.0
## Median : 713.0      Median :3910      Median :344.0      Median :498.0
## Mean   : 904.0      Mean   :3507      Mean   :388.7      Mean   :431.3
## 3rd Qu.:1295.0      3rd Qu.:3958      3rd Qu.:540.0      3rd Qu.:543.0
## Max.   :1877.0      Max.   :4007      Max.   :736.0      Max.   :588.0
```

```
## episiotomy          total
## Min.   : 84.0   Min.   :2899
## 1st Qu.:186.5   1st Qu.:3826
## Median :289.0   Median :4752
## Mean    :391.0   Mean    :4327
## 3rd Qu.:544.5   3rd Qu.:5042
## Max.    :800.0   Max.    :5331
```

```
summary(outcomes)
```

```
##      setting          caregiver      prolapsed_cord  uterine_rupture
## Length:3              Length:3          Min.   :2.000   Min.   :0.0000
## Class :character      Class :character  1st Qu.:4.000   1st Qu.:0.0000
## Mode  :character      Mode  :character  Median :6.000   Median :0.0000
##                                     Mean    :5.667   Mean    :0.6667
##                                     3rd Qu.:7.500   3rd Qu.:1.0000
##                                     Max.    :9.000   Max.    :2.0000
## postpartum_hemorrhage blood_transfusion obstetric_shock    death
## Min.   :110.0   Min.   : 2.0   Min.   :1   Min.   :0
## 1st Qu.:197.5   1st Qu.: 6.0   1st Qu.:1   1st Qu.:0
## Median :285.0   Median :10.0   Median :1   Median :0
## Mean    :250.7   Mean    : 9.0   Mean    :1   Mean    :0
## 3rd Qu.:321.0   3rd Qu.:12.5   3rd Qu.:1   3rd Qu.:0
## Max.    :357.0   Max.    :15.0   Max.    :1   Max.    :0
## placenta_removal  uterine_prolapse    pyrexia          uti
## Min.   :28.00   Min.   :1.000   Min.   : 19.00   Min.   :0.0
## 1st Qu.:56.50   1st Qu.:1.000   1st Qu.: 43.50   1st Qu.:0.5
## Median :85.00   Median :1.000   Median : 68.00   Median :1.0
## Mean    :67.67   Mean    :1.333   Mean    : 80.33   Mean    :2.0
## 3rd Qu.:87.50   3rd Qu.:1.500   3rd Qu.:111.00   3rd Qu.:3.0
## Max.    :90.00   Max.    :2.000   Max.    :154.00   Max.    :5.0
## puerperal_fever wound_infection    no_tear          first_second_degree
## Min.   :1.0   Min.   : 0.0   Min.   :1578   Min.   :1262
## 1st Qu.:2.5   1st Qu.: 5.5   1st Qu.:1884   1st Qu.:1824
## Median :4.0   Median :11.0   Median :2189   Median :2387
## Mean    :4.0   Mean    : 9.0   Mean    :2019   Mean    :2162
## 3rd Qu.:5.5   3rd Qu.:13.5   3rd Qu.:2240   3rd Qu.:2612
## Max.    :7.0   Max.    :16.0   Max.    :2291   Max.    :2836
## third_fourth_degree unknown_degree  cervical_tear      total
## Min.   : 34.0   Min.   :21.00   Min.   :2.000   Min.   :2899
## 1st Qu.: 85.5   1st Qu.:23.00   1st Qu.:3.000   1st Qu.:3826
## Median :137.0   Median :25.00   Median :4.000   Median :4752
## Mean    :118.0   Mean    :28.33   Mean    :3.667   Mean    :4327
## 3rd Qu.:160.0   3rd Qu.:32.00   3rd Qu.:4.500   3rd Qu.:5042
## Max.    :183.0   Max.    :39.00   Max.    :5.000   Max.    :5331
```

Some variables have a very low sample size/count, which is not as useful for visualization.

### Check for duplicate entries

As these two .csv's were made by hand (and are very small overall), we verify there are no duplicate entries.

## Prepare processed data

```
interventions_ratios <- interventions %>%
  mutate_at(vars(-c(setting, caregiver, total)), funs(./total)) %>%
  unite("birth_plan", c(setting, caregiver), sep = ", ")

## Warning: `funs()` was deprecated in dplyr 0.8.0.
## Please use a list of either functions or lambdas:
##
##   # Simple named list:
##   list(mean = mean, median = median)
##
##   # Auto named with `tibble::lst()`:
##   tibble::lst(mean, median)
##
##   # Using lambdas
##   list(~ mean(., trim = .2), ~ median(., na.rm = TRUE))
## This warning is displayed once every 8 hours.
## Call `lifecycle::last_lifecycle_warnings()` to see where this warning was generated.

outcomes_ratios <- outcomes %>%
  mutate_at(vars(-c(setting, caregiver, total)), funs(./total)) %>%
  unite("birth_plan", c(setting, caregiver), sep = ", ")
```

## Build intuition about individual variables

```
i1 <- ggplot(interventions_ratios) +
  geom_col(aes(birth_plan, cesarean, fill = birth_plan)) +
  theme(axis.text = element_text(size = 1)) +
  theme_minimal() +
  guides(fill="none") +
  xlab("")

i2 <- ggplot(interventions_ratios) +
  geom_col(aes(birth_plan, oxytocin, fill = birth_plan)) +
  theme(axis.text = element_text(size = 1)) +
  theme_minimal() +
  guides(fill="none") +
  xlab("")

i3 <- ggplot(interventions_ratios) +
  geom_col(aes(birth_plan, epidural, fill = birth_plan)) +
  theme(axis.text = element_text(size = 1)) +
  theme_minimal() +
  guides(fill="none") +
  xlab("")

i4 <- ggplot(interventions_ratios) +
  geom_col(aes(birth_plan, episiotomy, fill = birth_plan)) +
  theme(axis.text = element_text(size = 1)) +
  theme_minimal() +
  guides(fill="none") +
  xlab("")
```



$$(i1 + i2) / (i3 + i4)$$

