

EHR Data Quality Control and Analysis

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```
# Set path to data directory
data_path <- "C:/Users/User/Downloads/test_data/data/dest/"

# Load main datasets
```

```
patients <- read_csv(paste0(data_path, "patients.csv"), show_col_types = FALSE)
conditions <- read_csv(paste0(data_path, "conditions.csv"), show_col_types = FALSE)
observations <- read_csv(paste0(data_path, "observations.csv"), show_col_types = FALSE)
medications <- read_csv(paste0(data_path, "medications.csv"), show_col_types = FALSE)
```

```
## Warning: One or more parsing issues, call `problems()` on your data frame for details,
## e.g.:
##   dat <- vroom(...)
##   problems(dat)
```

```
encounters <- read_csv(paste0(data_path, "encounters.csv"), show_col_types = FALSE)

# Load dictionaries
dict_snomed <- read_csv(paste0(data_path, "dictionary_snomed.csv"), show_col_types = FALSE)
dict_rxnorm <- read_csv(paste0(data_path, "dictionary_rxnorm.csv"), show_col_types = FALSE)
dict_loinc <- read_csv(paste0(data_path, "dictionary_loinc.csv"), show_col_types = FALSE)
```

Patients Quality Control (QC)

```
# Convert date columns
patients$BIRTHDATE <- as.Date(patients$BIRTHDATE)
patients$DEATHDATE <- as.Date(patients$DEATHDATE)

# Calculate age
patients$AGE <- as.numeric(floor(interval(patients$BIRTHDATE, Sys.Date()) / years(1)))

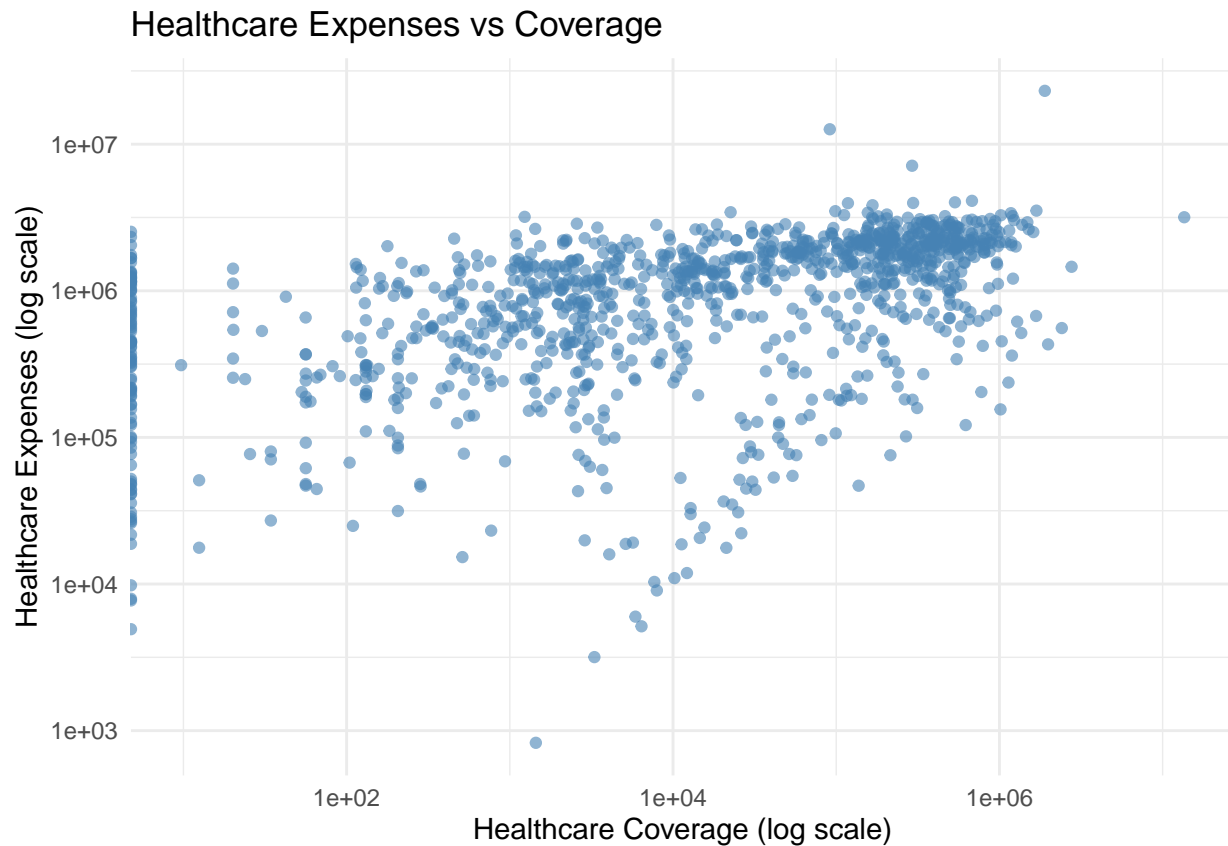
# HEALTHCARE_EXPENSES vs HEALTHCARE_COVERAGE Scatter plot with log scales
ggplot(patients, aes(x = HEALTHCARE_COVERAGE, y = HEALTHCARE_EXPENSES)) +
  geom_point(alpha = 0.6, color = "steelblue") +
  scale_x_log10() +
  scale_y_log10() +
  labs(
    title = "Healthcare Expenses vs Coverage",
```

```

x = "Healthcare Coverage (log scale)",
y = "Healthcare Expenses (log scale)"
) +
theme_minimal()

```

```
## Warning in scale_x_log10(): log-10 transformation introduced infinite values.
```



```

# Filter out invalid data
patients <- patients %>%
  mutate(
    # Logical flags for invalid healthcare expense/coverage
    FLAG_INVALID_EXPENSE = HEALTHCARE_EXPENSES < 500 | HEALTHCARE_EXPENSES > 2e7,
    FLAG_INVALID_COVERAGE = HEALTHCARE_COVERAGE > HEALTHCARE_EXPENSES
  ) %>%
  filter(
    AGE >= 0 & AGE <= 100,
    is.na(DEATHDATE) | BIRTHDATE <= DEATHDATE,
    !(LAT == 0 & LON == 0),
    LAT >= -90 & LAT <= 90,
    LON >= -180 & LON <= 180,
    !(HEALTHCARE_EXPENSES == 0 & HEALTHCARE_COVERAGE > 1000),
    GENDER %in% c("M", "F", "", NA),
    !((PREFIX == "Mr." & GENDER == "F") |
      (PREFIX %in% c("Mrs.", "Ms.", "Miss") & GENDER == "M"))
  )

```

```

)

# Remove US state + UK city mismatches
us_states <- state.abb
uk_cities <- c("London", "Manchester", "Leeds", "Birmingham", "Glasgow", "Liverpool")
patients <- patients[!(patients$CITY %in% uk_cities & patients$STATE %in% us_states), ]

# Fix race coding
patients$RACE[patients$RACE == "XJniDSe"] <- "other (possibly miscoded)"

# ZIP formatting
patients$ZIP <- sprintf("%05d", patients$ZIP)

# The number of patients and unique patients after QC
print(dim(patients))

```

```
## [1] 930 28
```

```
print(n_distinct(patients$Id))
```

```
## [1] 930
```

Encounters QC

```

# Convert date columns
encounters$START <- as.POSIXct(encounters$START)
encounters$STOP <- as.POSIXct(encounters$STOP)
# Filter out invalid data
encounters <- encounters %>%
  filter(is.na(STOP) | STOP >= START) %>%
  filter(BASE_ENCOUNTER_COST >= 0, TOTAL_CLAIM_COST >= 0, PAYER_COVERAGE >= 0) %>%
  filter(PATIENT %in% patients$Id)

```

Conditions QC

```

# Convert date columns
conditions$START <- as.Date(conditions$START)
conditions$STOP <- as.Date(conditions$STOP)

# Filter out invalid data
conditions <- conditions %>%
  filter(is.na(STOP) | START <= STOP) %>%
  filter(PATIENT %in% patients$Id) %>%
  filter(ENCOUNTER %in% encounters$Id) %>%
  filter(CODE %in% dict_snomed$CODE)

conditions <- left_join(conditions, dict_snomed, by = "CODE")

```

```
## Warning in left_join(conditions, dict_snomed, by = "CODE"): Detected an unexpected many-to-many relationship
## i Row 304 of `x` matches multiple rows in `y`.
## i Row 15 of `y` matches multiple rows in `x`.
## i If a many-to-many relationship is expected, set `relationship =
##   "many-to-many"` to silence this warning.
```

```
# Add 0/1 flags for socioeconomic factors based on DESCRIPTION content
conditions <- conditions %>%
  mutate(
    factor_education = as.integer(grepl("education|high school|primary school|higher education", DESCRIPTION, ignore.case = TRUE)),
    factor_employment = as.integer(grepl("employment|unemployed|labor force", DESCRIPTION, ignore.case = TRUE)),
    factor_stress = as.integer(grepl("stress|anxiety|panic", DESCRIPTION, ignore.case = TRUE)),
    factor_refugee = as.integer(grepl("refugee", DESCRIPTION, ignore.case = TRUE)),
    factor_criminal = as.integer(grepl("criminal", DESCRIPTION, ignore.case = TRUE)),
    factor_violence = as.integer(grepl("violence|abuse|victim", DESCRIPTION, ignore.case = TRUE)),
    factor_transport = as.integer(grepl("transport", DESCRIPTION, ignore.case = TRUE)),
    factor_drugs_alcohol = as.integer(grepl("alcohol|misuses drugs", DESCRIPTION, ignore.case = TRUE)),
    factor_social_isolation = as.integer(grepl("social|isolation|limited contact", DESCRIPTION, ignore.case = TRUE)),
    factor_military = as.integer(grepl("armed forces|military", DESCRIPTION, ignore.case = TRUE)),
    factor_housing = as.integer(grepl("housing", DESCRIPTION, ignore.case = TRUE))
  )
```

Observations QC

```
# Convert date columns
observations$DATE <- as.POSIXct(observations$DATE)
observations$CODE <- as.character(observations$CODE)

dict_loinc$CODE <- as.character(dict_loinc$CODE)

# Filter out invalid data
observations <- observations %>%
  left_join(dict_loinc, by = "CODE") %>%
  filter(!is.na(DESCRIPTION), !is.na(VALUE), VALUE != "", !is.na(TYPE), TYPE != "")
```

```
## Warning in left_join(., dict_loinc, by = "CODE"): Detected an unexpected many-to-many relationship between
## i Row 10 of `x` matches multiple rows in `y`.
## i Row 1 of `y` matches multiple rows in `x`.
## i If a many-to-many relationship is expected, set `relationship =
##   "many-to-many"` to silence this warning.
```

```
observations$VALUE_NUMERIC <- suppressWarnings(as.numeric(observations$VALUE))
observations <- observations[observations$TYPE != "numeric" | !is.na(observations$VALUE_NUMERIC), ]

# Weight-kg mismatch
mismatch <- grepl("weight", tolower(observations$DESCRIPTION)) & !grepl("kg", tolower(observations$UNIT))
observations <- observations[!mismatch, ]

observations <- observations %>%
  filter(PATIENT %in% patients$Id, ENCOUNTER %in% encounters$Id)
```

```

# Join observations with patients to get DEATHDATE
observations <- observations %>%
  left_join(patients %>% select(Id, DEATHDATE), by = c("PATIENT" = "Id"))

# Keep only observations that occurred before or on the DEATHDATE, or if DEATHDATE is missing
observations <- observations %>%
  filter(is.na(DEATHDATE) | DATE <= DEATHDATE)

```

Medications QC

```

# Convert date columns
medications$START <- as.POSIXct(medications$START)
medications$STOP <- as.POSIXct(medications$STOP)

medications <- medications %>%
  filter(is.na(STOP) | START <= STOP, PATIENT %in% patients$Id, ENCOUNTER %in% encounters$Id) %>%
  filter(BASE_COST >= 0, PAYER_COVERAGE >= 0, TOTALCOST >= 0)

medications <- medications %>%
  left_join(dict_rxnrm, by = "CODE") %>%
  filter(!is.na(DESCRIPTION)) %>%
  filter(BASE_COST <= TOTALCOST) %>%
  filter(abs((BASE_COST + PAYER_COVERAGE) - TOTALCOST) <= 1)

```

```

## Warning in left_join(., dict_rxnrm, by = "CODE"): Detected an unexpected many-to-many relationship
## i Row 51 of `x` matches multiple rows in `y`.
## i Row 9 of `y` matches multiple rows in `x`.
## i If a many-to-many relationship is expected, set `relationship =
##   "many-to-many"` to silence this warning.

```

```

# Join medications with patients to get DEATHDATE
medications <- medications %>%
  left_join(patients %>% select(Id, DEATHDATE), by = c("PATIENT" = "Id"))

# Filter out medications that start after the patient's death
medications <- medications %>%
  filter(is.na(DEATHDATE) | START <= DEATHDATE)

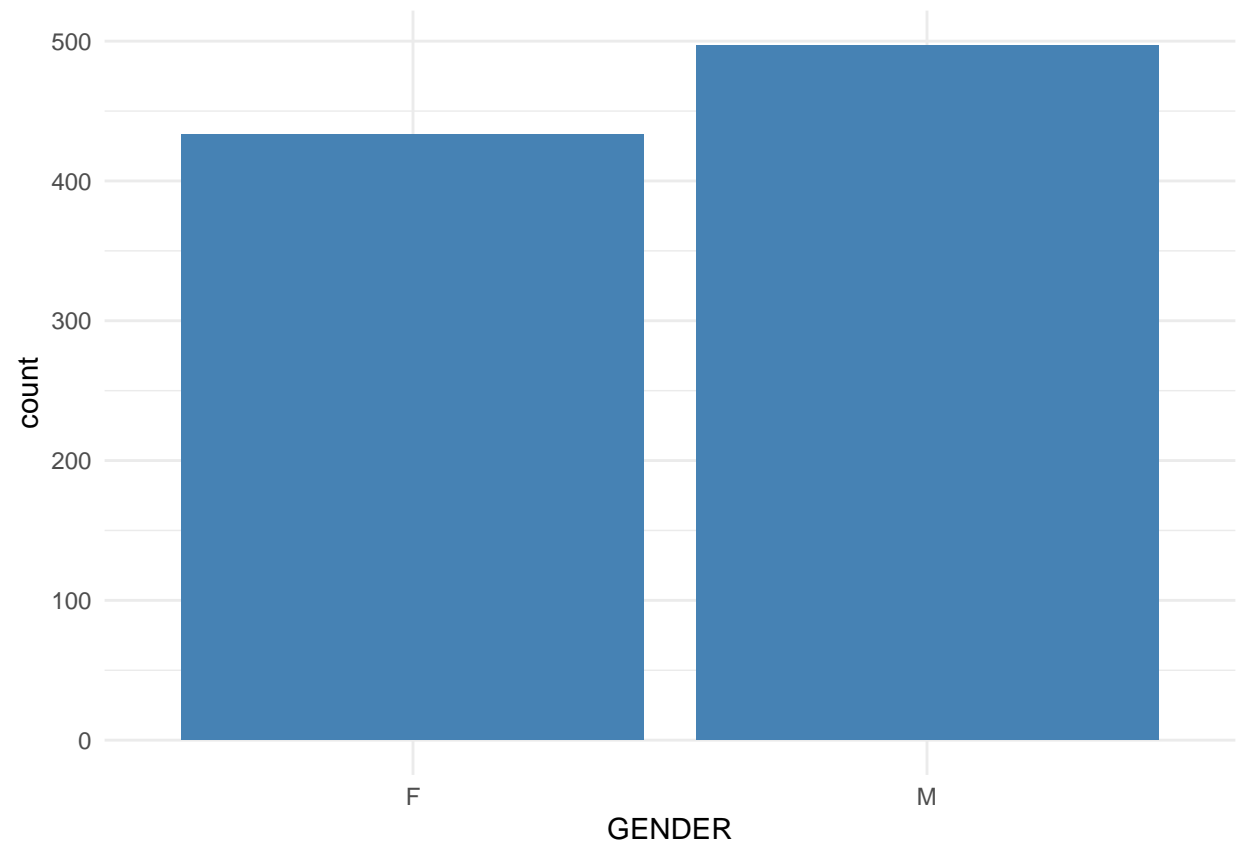
```

Descriptive Summaries and Visualizations

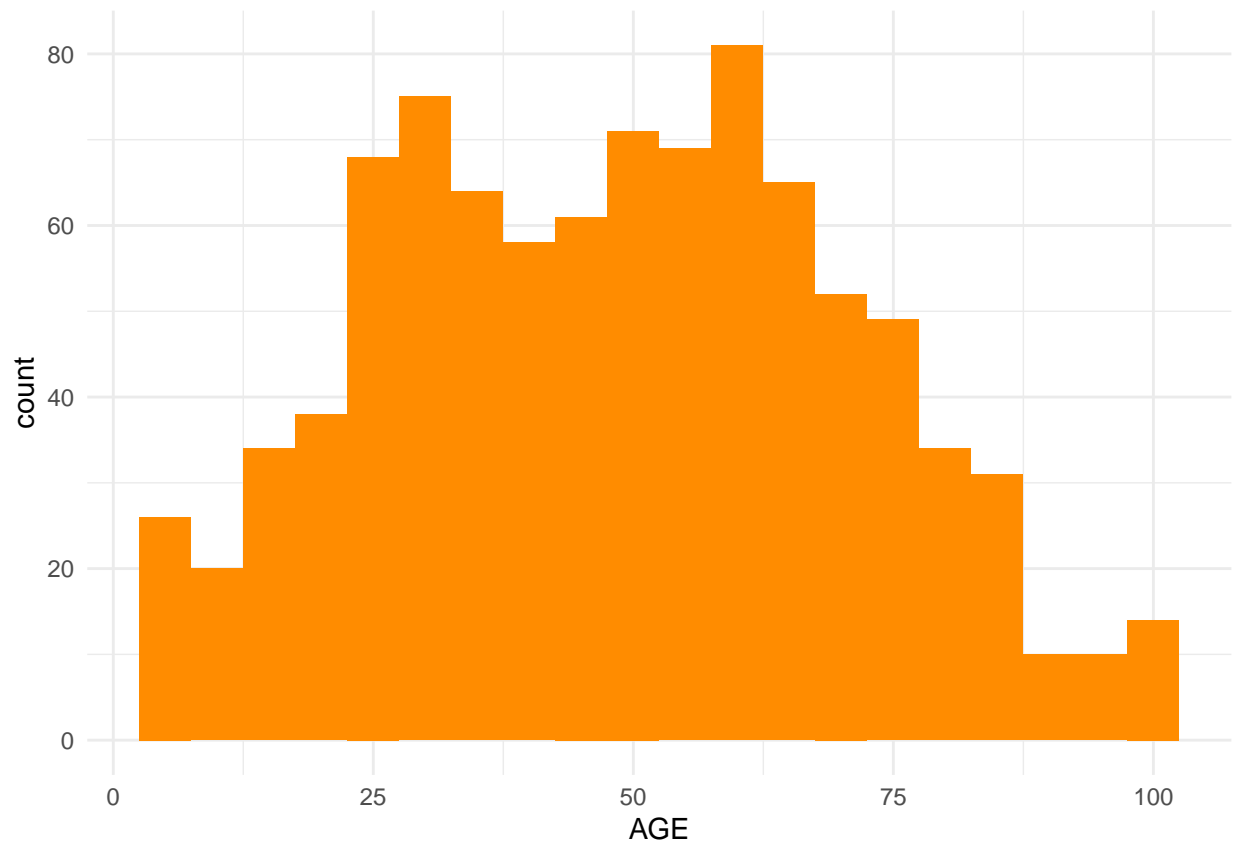
```

# Gender distribution
ggplot(patients, aes(x = GENDER)) + geom_bar(fill = "steelblue") + theme_minimal()

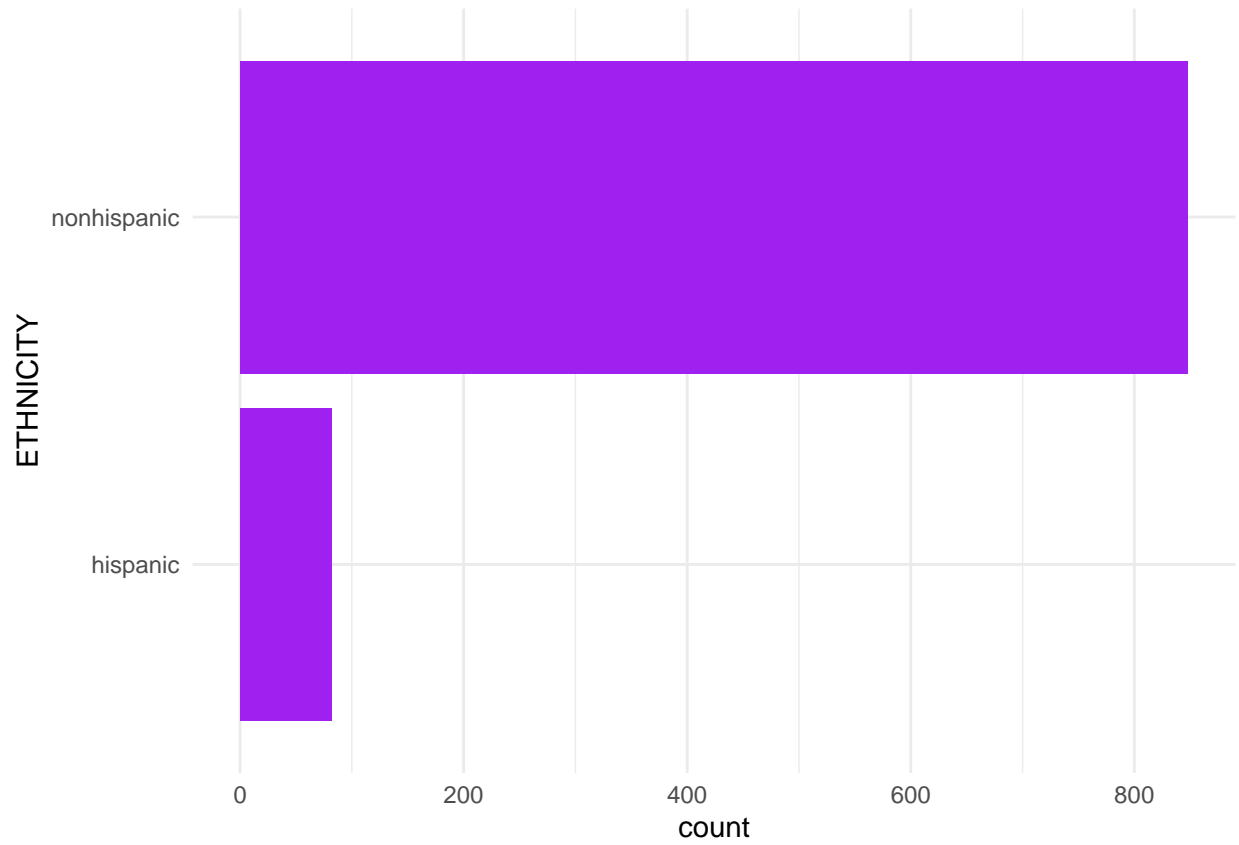
```



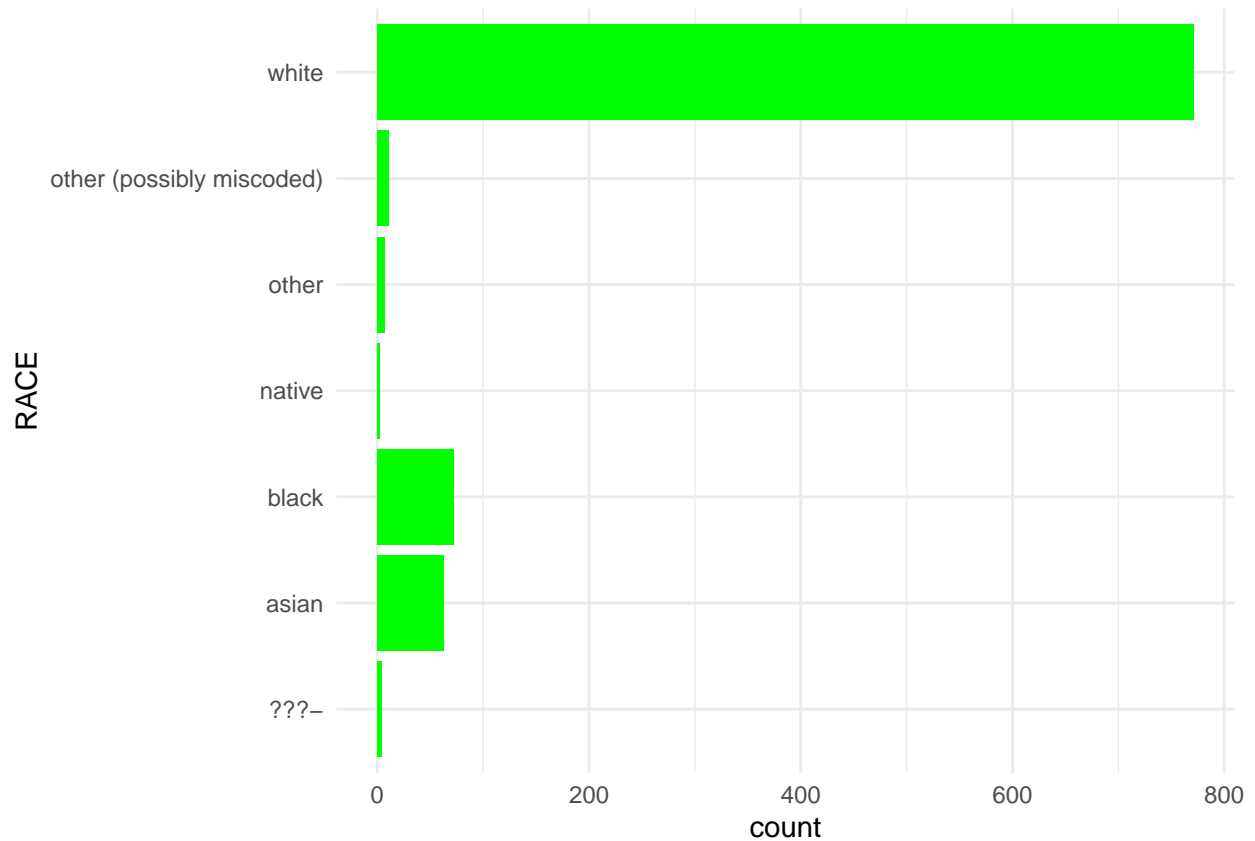
```
# Age distribution  
ggplot(patients, aes(x = AGE)) + geom_histogram(binwidth = 5, fill = "darkorange") + theme_minimal()
```



```
# Ethnicity and Race  
ggplot(patients, aes(x = ETHNICITY)) + geom_bar(fill = "purple") + coord_flip() + theme_minimal()
```



```
ggplot(patients, aes(x = RACE)) + geom_bar(fill = "green") + coord_flip() + theme_minimal()
```

Top Medications, Conditions, and Labs

```
top_meds <- medications %>%
  group_by(PATIENT, DESCRIPTION) %>% summarise(n = n()) %>%
  group_by(DESCRIPTION) %>% summarise(total = sum(n)) %>% arrange(desc(total)) %>% head(10)
```

`summarise()` has grouped output by 'PATIENT'. You can override using the
`.groups` argument.

```
print(top_meds)
```

```
## # A tibble: 10 x 2
##   DESCRIPTION                                total
##   <chr>                                     <int>
## 1 Hydrochlorothiazide 25 MG Oral Tablet      5944
## 2 Simvastatin 10 MG Oral Tablet              1639
## 3 1 ML Epoetin Alfa 4000 UNT/ML Injection [Epogen] 1015
## 4 insulin human isophane 70 UNT/ML / Regular Insulin Human 30 UNT/ML I~ 992
## 5 lisinopril 10 MG Oral Tablet               915
## 6 24 HR Metformin hydrochloride 500 MG Extended Release Oral Tablet 510
## 7 Acetaminophen 325 MG Oral Tablet           480
## 8 Simvastatin 20 MG Oral Tablet              455
## 9 Amlodipine 5 MG Oral Tablet                415
## 10 amLODIPine 2.5 MG Oral Tablet             406
```

```

conditions_summary <- conditions %>%
  group_by(DESCRIPTION) %>% summarise(n = n()) %>% arrange(desc(n))
top_conditions <- head(conditions_summary, 10)
least_conditions <- tail(conditions_summary, 10)

print(top_conditions)

```

```

## # A tibble: 10 x 2
##   DESCRIPTION          n
##   <chr>              <int>
## 1 Full-time employment (finding) 11051
## 2 Stress (finding)              4130
## 3 Part-time employment (finding) 2047
## 4 Social isolation (finding)      1067
## 5 Limited social contact (finding) 1030
## 6 Viral sinusitis (disorder)      1025
## 7 Not in labor force (finding)     900
## 8 Victim of intimate partner abuse (finding) 695
## 9 Acute viral pharyngitis (disorder) 554
## 10 Normal pregnancy             482

```

```

print(least_conditions)

```

```

## # A tibble: 10 x 2
##   DESCRIPTION          n
##   <chr>              <int>
## 1 Infection caused by Pseudomonas aeruginosa 1
## 2 Injury of heart (disorder)                1
## 3 Injury of kidney (disorder)               1
## 4 Macular edema and retinopathy due to type 2 diabetes mellitus (disorde~ 1
## 5 Major depression disorder                 1
## 6 Male Infertility                         1
## 7 Microalbuminuria due to type 2 diabetes mellitus (disorder)             1
## 8 Pyelonephritis                          1
## 9 Spina bifida occulta (disorder)           1
## 10 Tear of meniscus of knee                 1

```

```

top_labs <- observations %>%
  group_by(DESCRIPTION) %>% summarise(n = n()) %>% arrange(desc(n)) %>% head(10)
print(top_labs)

```

```

## # A tibble: 10 x 2
##   DESCRIPTION          n
##   <chr>              <int>
## 1 Diastolic Blood Pressure 11386
## 2 Systolic Blood Pressure 11386
## 3 Pain severity - 0-10 verbal numeric rating [Score] - Reported 11372
## 4 Body Weight             10114
## 5 Heart rate              9948
## 6 Respiratory rate        9948
## 7 Body Height             9648
## 8 Tobacco smoking status NHIS 9616

```

```
## 9 Body Mass Index
## 10 Housing status
```

9109
7584

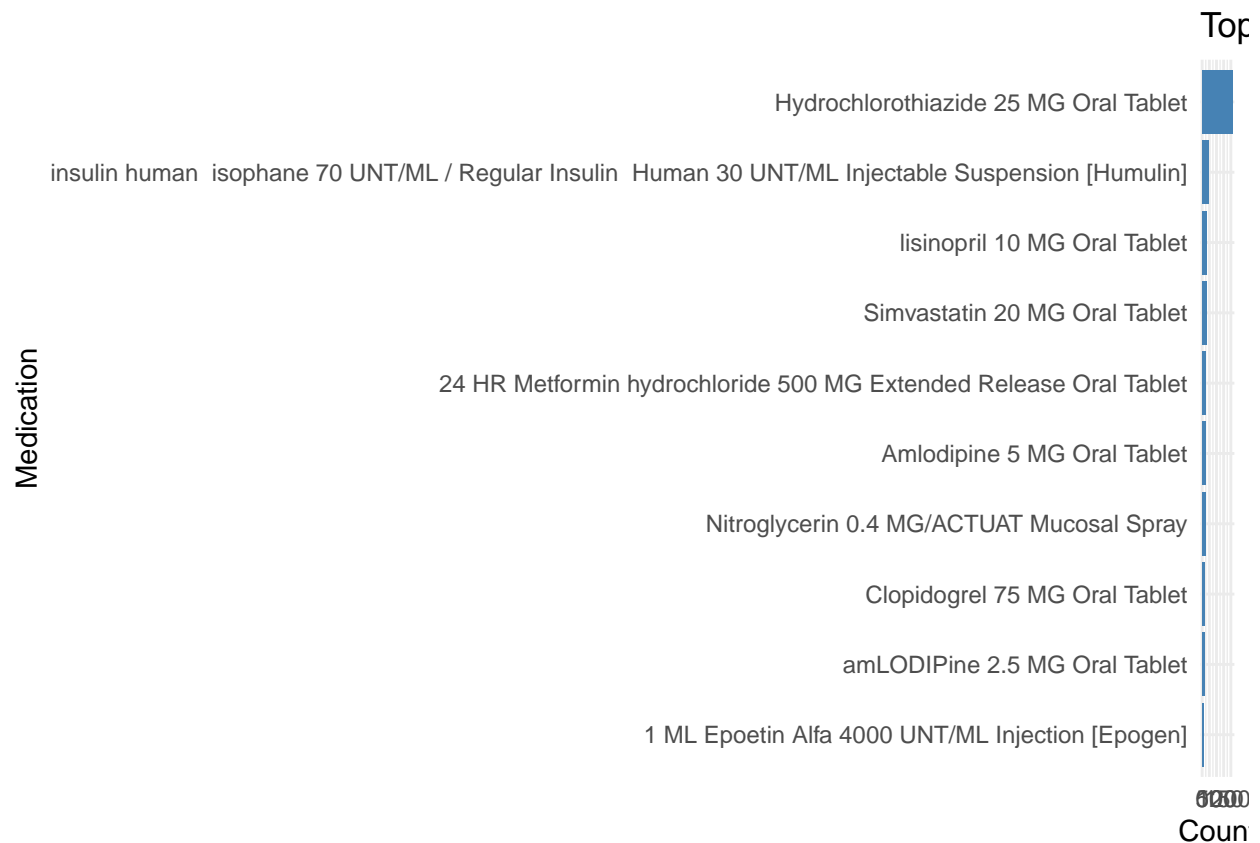
Medications by Socioeconomic Group

Medications for Unemployed Patients

```
# Step 1: Find encounters where unemployment is recorded
unemployed_encounters <- conditions %>%
  filter(grepl("unemployed|not in labor force", DESCRIPTION, ignore.case = TRUE)) %>%
  distinct(PATIENT, ENCOUNTER)

# Step 2: Join to medications
medications_unemployed <- medications %>%
  semi_join(unemployed_encounters, by = c("PATIENT", "ENCOUNTER"))

# Step 3: Plot top medications
medications_unemployed %>%
  count(DESCRIPTION, sort = TRUE) %>%
  slice_max(n, n = 10) %>%
  ggplot(aes(x = reorder(DESCRIPTION, n), y = n)) +
  geom_col(fill = "steelblue") +
  coord_flip() +
  labs(
    title = "Top Medications Among Unemployed Patients",
    x = "Medication", y = "Count"
  ) +
  theme_minimal()
```



Conditions with Highest Out-of-Pocket Costs

```
cond_cost <- merge(conditions, encounters[, c("Id", "TOTAL_CLAIM_COST", "PAYER_COVERAGE")],
                    by.x = "ENCOUNTER", by.y = "Id") %>%
  group_by(DESCRIPTION) %>%
  summarise(
    avg_claim_cost = mean(TOTAL_CLAIM_COST, na.rm = TRUE),
    avg_patient_pay = mean(TOTAL_CLAIM_COST - PAYER_COVERAGE, na.rm = TRUE)) %>%
  arrange(desc(avg_patient_pay))
```

Reasons for Visit

```
reasons <- encounters %>%
  filter(!is.na(REASONCODE)) %>%
  count(REASONCODE, sort = TRUE) %>%
  left_join(dict_snomed, by = c("REASONCODE" = "CODE")) %>%
  rename(ReasonDescription = DESCRIPTION)
print(head(reasons, 20))
```

```
## # A tibble: 20 x 3
##   REASONCODE      n ReasonDescription
```

```
##          <dbl> <int> <chr>
## 1    72892002  3585 Normal pregnancy
## 2    55822004  1640 Hyperlipidemia
## 3    88805009  1536 Chronic congestive heart failure (disorder)
## 4    444814009 1210 Viral sinusitis (disorder)
## 5     10509002   633 Acute bronchitis (disorder)
## 6    195662009  632 Acute viral pharyngitis (disorder)
## 7    254837009  380 Malignant neoplasm of breast (disorder)
## 8    271737000  284 Anemia (disorder)
## 9    192127007  280 Child attention deficit disorder
## 10   59621000   242 Hypertension
## 11   75498004   213 Acute bacterial sinusitis (disorder)
## 12   195967001  193 Asthma
## 13   36971009   176 Sinusitis (disorder)
## 14   55680006   145 Drug overdose
## 15   43878008   131 Streptococcal sore throat (disorder)
## 16   65363002   122 Otitis media
## 17   233678006  115 Childhood asthma
## 18   74400008   106 Appendicitis
## 19   82423001    99 Chronic pain
## 20   196416002   89 Impacted molars
```

Summaries

```
summary(patients)
```

```
##          Id          BIRTHDATE          DEATHDATE
## Length:930      Min.   :1924-12-04      Min.   :1945-10-20
## Class :character 1st Qu.:1960-03-30      1st Qu.:1997-01-05
## Mode  :character Median :1975-11-20      Median :2008-07-02
##                  Mean  :1976-06-17      Mean  :2005-04-22
##                  3rd Qu.:1994-07-28      3rd Qu.:2015-05-22
##                  Max.   :2021-09-05      Max.   :2021-09-29
##                  NA's   :833
##          SSN          DRIVERS          PASSPORT          PREFIX
## Length:930      Length:930      Length:930      Length:930
## Class :character Class :character Class :character Class :character
## Mode  :character Mode  :character Mode  :character Mode  :character
##
##
##
##          FIRST          LAST          SUFFIX          MAIDEN
## Length:930      Length:930      Length:930      Length:930
## Class :character Class :character Class :character Class :character
## Mode  :character Mode  :character Mode  :character Mode  :character
##
##
##
##          MARITAL          RACE          ETHNICITY          GENDER
## Length:930      Length:930      Length:930      Length:930
```

```

## Class :character   Class :character   Class :character   Class :character
## Mode  :character   Mode  :character   Mode  :character   Mode  :character
##
##
##
## BIRTHPLACE          ADDRESS          CITY          STATE
## Length:930          Length:930          Length:930          Length:930
## Class :character     Class :character     Class :character     Class :character
## Mode  :character     Mode  :character     Mode  :character     Mode  :character
##
##
##
## COUNTY              ZIP              LAT              LON
## Length:930          Length:930          Min.   :41.50      Min.   : -73.45
## Class :character     Class :character     1st Qu.:42.10      1st Qu.: -71.43
## Mode  :character     Mode  :character     Median :42.31      Median : -71.13
##                                     Mean  :42.25      Mean  : -71.30
##                                     3rd Qu.:42.45      3rd Qu.: -70.99
##                                     Max.   :42.89      Max.   : -69.98
##
## HEALTHCARE_EXPENSES HEALTHCARE_COVERAGE AGE          FLAG_INVALID_EXPENSE
## Min.   :    826      Min.   :    0      Min.   :   3.00      Mode :logical
## 1st Qu.: 589426      1st Qu.: 1675      1st Qu.: 30.25      FALSE:930
## Median :1248438      Median : 22745      Median : 49.00
## Mean   :1305014      Mean   :158385      Mean   : 48.58
## 3rd Qu.:1914155      3rd Qu.:214946      3rd Qu.: 65.00
## Max.   :12641789      Max.   :2411214      Max.   :100.00
##
## FLAG_INVALID_COVERAGE
## Mode :logical
## FALSE:900
## TRUE :30
##
##
##
##

```

```
summary(encounters)
```

```

##      Id              START              STOP
## Length:49339      Min.   :1925-10-01 21:07:04      Min.   :1925-10-01 21:22:04
## Class :character   1st Qu.:2001-10-31 20:33:41      1st Qu.:2001-11-01 06:28:47
## Mode  :character   Median :2013-07-16 01:52:59      Median :2013-07-16 02:35:10
##                                     Mean  :2007-08-27 23:09:14      Mean  :2007-08-28 04:39:02
##                                     3rd Qu.:2017-09-04 05:48:47      3rd Qu.:2017-09-04 06:03:47
##                                     Max.   :2021-11-19 16:50:22      Max.   :2021-11-19 17:05:22
##
## PATIENT              ORGANIZATION          PROVIDER          PAYER
## Length:49339          Length:49339          Length:49339          Length:49339
## Class :character       Class :character       Class :character       Class :character
## Mode  :character       Mode  :character       Mode  :character       Mode  :character
##

```

```
##
##
##
## ENCOUNTERCLASS          CODE          BASE_ENCOUNTER_COST TOTAL_CLAIM_COST
## Length:49339           Min.    : 1505002      Min.    : 77.49          Min.    : 0.0
## Class :character       1st Qu.:162673000      1st Qu.: 77.49          1st Qu.: 129.2
## Mode  :character       Median :185347001      Median :129.16          Median : 786.3
##                               Mean  :260883600      Mean  :113.68          Mean  : 3976.6
##                               3rd Qu.:390906007      3rd Qu.:129.16          3rd Qu.: 1615.7
##                               Max.   :702927004      Max.   :129.16          Max.   :873646.2
##
## PAYER_COVERAGE          REASONCODE
## Min.    : 0.00          Min.    :1.734e+06
## 1st Qu.: 0.00          1st Qu.:6.256e+07
## Median : 0.00          Median :7.289e+07
## Mean    : 866.70        Mean    :5.840e+12
## 3rd Qu.: 20.14         3rd Qu.:1.957e+08
## Max.    :227851.81      Max.    :1.094e+16
##                               NA's    :36206
```

[summary](#)(medications)

```
##          START                      STOP                      PATIENT
## Min.    :1931-03-09 03:45:19      Min.    :1931-03-27 03:45:19      Length:16967
## 1st Qu.:1997-11-03 03:45:08      1st Qu.:1998-02-07 07:37:02      Class :character
## Median :2008-12-06 23:41:21      Median :2009-01-20 11:58:49      Mode  :character
## Mean    :2005-06-23 23:21:23      Mean    :2005-07-30 17:20:11
## 3rd Qu.:2016-06-08 03:55:18      3rd Qu.:2016-06-19 21:56:31
## Max.    :2021-11-18 14:01:22      Max.    :2021-11-18 14:01:22
##                               NA's    :494
##          PAYER          ENCOUNTER          CODE          BASE_COST
## Length:16967           Length:16967           Min.    : 106258      Min.    : 0.01
## Class :character       Class :character       1st Qu.: 310798      1st Qu.: 0.01
## Mode  :character       Mode  :character       Median : 310798      Median : 0.02
##                               Mean    : 427546      Mean    : 383.79
##                               3rd Qu.: 314231      3rd Qu.: 60.56
##                               Max.    :2123111      Max.    :6994.54
##
## PAYER_COVERAGE          DISPENSES          TOTALCOST          REASONCODE
## Min.    :0.0000000      Min.    : 1.00          Min.    : 0.01          Min.    : 10509002
## 1st Qu.:0.0000000      1st Qu.: 1.00          1st Qu.: 0.12          1st Qu.: 55822004
## Median :0.0000000      Median : 1.00          Median : 0.56          Median : 59621000
## Mean    :0.0001532      Mean    : 12.09         Mean    : 383.88         Mean    :111906137
## 3rd Qu.:0.0000000      3rd Qu.: 12.00         3rd Qu.: 60.56         3rd Qu.: 59621000
## Max.    :0.6200000      Max.    :45000.00       Max.    :6994.54         Max.    :706870000
##                               NA's    :2852
## DESCRIPTION          DEATHDATE
## Length:16967          Min.    :1945-10-20
## Class :character       1st Qu.:2008-11-15
## Mode  :character       Median :2008-11-15
##                               Mean    :2009-07-03
##                               3rd Qu.:2012-10-12
##                               Max.    :2021-09-29
##                               NA's    :12753
```

```
summary(conditions)
```

```
##          START          STOP          PATIENT
## Min.      :1930-08-05   Min.      :1930-10-21   Length:31334
## 1st Qu.:1996-04-30     1st Qu.:2000-05-09   Class :character
## Median :2011-07-23     Median :2013-06-20   Mode  :character
## Mean    :2005-01-29     Mean    :2007-07-07
## 3rd Qu.:2017-02-08     3rd Qu.:2018-03-07
## Max.    :2021-11-15     Max.    :2021-11-18
##                               NA's      :7014
## ENCOUNTER          CODE          DESCRIPTION          factor_education
## Length:31334      Min.      :1.734e+06   Length:31334      Min.      :0.00000
## Class :character   1st Qu.:9.130e+07   Class :character   1st Qu.:0.00000
## Mode  :character   Median :1.609e+08   Mode  :character   Median :0.00000
##                               Mean    :5.463e+13   Mean    :0.02582
##                               3rd Qu.:2.376e+08   3rd Qu.:0.00000
##                               Max.    :1.094e+16   Max.    :1.00000
##
## factor_employment factor_stress   factor_refugee   factor_criminal
## Min.      :0.0000   Min.      :0.0000   Min.      :0.0000   Min.      :0.000000
## 1st Qu.:0.0000   1st Qu.:0.0000   1st Qu.:0.0000   1st Qu.:0.000000
## Median :0.0000   Median :0.0000   Median :0.0000   Median :0.000000
## Mean    :0.4488   Mean    :0.1356   Mean    :0.0015   Mean    :0.005234
## 3rd Qu.:1.0000   3rd Qu.:0.0000   3rd Qu.:0.0000   3rd Qu.:0.000000
## Max.    :1.0000   Max.    :1.0000   Max.    :1.0000   Max.    :1.000000
##
## factor_violence   factor_transport   factor_drugs_alcohol
## Min.      :0.00000   Min.      :0.000000   Min.      :0.000000
## 1st Qu.:0.00000   1st Qu.:0.000000   1st Qu.:0.000000
## Median :0.00000   Median :0.000000   Median :0.000000
## Mean    :0.03763   Mean    :0.006159   Mean    :0.007245
## 3rd Qu.:0.00000   3rd Qu.:0.000000   3rd Qu.:0.000000
## Max.    :1.00000   Max.    :1.000000   Max.    :1.000000
##
## factor_social_isolation factor_military   factor_housing
## Min.      :0.00000   Min.      :0.000000   Min.      :0.000000
## 1st Qu.:0.00000   1st Qu.:0.000000   1st Qu.:0.000000
## Median :0.00000   Median :0.000000   Median :0.000000
## Mean    :0.06692   Mean    :0.001213   Mean    :0.004564
## 3rd Qu.:0.00000   3rd Qu.:0.000000   3rd Qu.:0.000000
## Max.    :1.00000   Max.    :1.000000   Max.    :1.000000
##
```

```
summary(observations)
```

```
##          DATE          PATIENT          ENCOUNTER
## Min.      :1935-11-30 10:04:43   Length:438591   Length:438591
## 1st Qu.:2013-07-18 12:07:29   Class :character   Class :character
## Median :2016-07-31 18:42:32   Mode  :character   Mode  :character
## Mean    :2015-01-30 02:48:47
## 3rd Qu.:2019-07-23 08:20:50
## Max.    :2021-11-18 16:26:22
```



```
##
##   CATEGORY          CODE          VALUE          UNITS
## Length:438591      Length:438591      Length:438591      Length:438591
## Class :character    Class :character    Class :character    Class :character
## Mode  :character    Mode  :character    Mode  :character    Mode  :character
##
##
##
##   TYPE          DESCRIPTION          VALUE_NUMERIC          DEATHDATE
## Length:438591      Length:438591      Min.   :   -3.7      Min.   :1945-10-20
## Class :character    Class :character    1st Qu.:    5.3      1st Qu.:2005-11-09
## Mode  :character    Mode  :character    Median :   29.2      Median :2011-07-10
##                                     Mean  :  3086.7      Mean  :2009-04-11
##                                     3rd Qu.:   92.0      3rd Qu.:2017-04-06
##                                     Max.   :957744.0      Max.   :2021-09-29
##                                     NA's   :168780      NA's   :356492
```

Exploring and comparing the distribution of: systolic and diastolic blood pressure and BMI measurements in patients with diagnosed hypertension

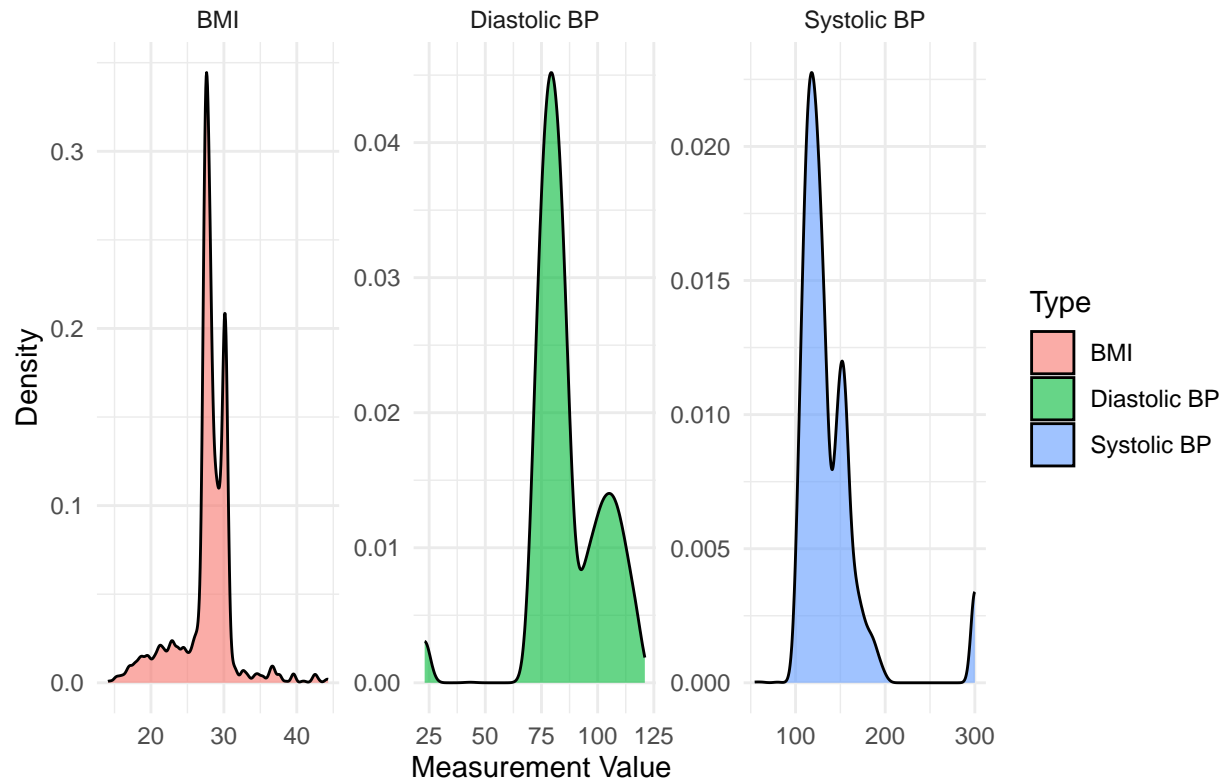
```
# 1. Get hypertensive patients
hypertensive_patients <- conditions %>%
  filter(grepl("hypertension", DESCRIPTION, ignore.case = TRUE)) %>%
  distinct(PATIENT)

# 2. Extract their blood pressure and BMI observations
bp_bmi_data <- observations %>%
  semi_join(hypertensive_patients, by = "PATIENT") %>%
  filter(CODE %in% c("55284-4", "8480-6", "8462-4", "39156-5")) %>% # systolic, diastolic, BMI
  mutate(
    Type = case_when(
      CODE == "8480-6" ~ "Systolic BP",
      CODE == "8462-4" ~ "Diastolic BP",
      CODE == "39156-5" ~ "BMI",
      CODE == "55284-4" ~ "Blood Pressure Panel",
      TRUE ~ "Other"
    ),
    VALUE = as.numeric(VALUE)
  ) %>%
  filter(!is.na(VALUE)) # remove any non-numeric or NA values

# 3. Plot distributions
bp_bmi_data %>%
  filter(Type %in% c("Systolic BP", "Diastolic BP", "BMI")) %>%
  ggplot(aes(x = VALUE, fill = Type)) +
  geom_density(alpha = 0.6) +
  facet_wrap(~Type, scales = "free") +
  labs(
    title = "Distribution of Systolic/Diastolic BP and BMI in Hypertensive Patients",
    x = "Measurement Value",
    y = "Density"
```

```
) +  
theme_minimal()
```

Distribution of Systolic/Diastolic BP and BMI in Hypertensive Patients



The crude, and adjusted (to the UK population as much as possible) prevalence of hypertension

```
# Step 1: Identify hypertension codes from SNOMED dictionary  
htn_codes <- dict_snomed %>%  
  filter(grepl("hypertension", tolower(DESCRIPTION))) %>%  
  pull(CODE)  
  
# Step 2: Mark hypertensive patients  
htn_patients <- conditions %>%  
  filter(CODE %in% htn_codes) %>%  
  distinct(PATIENT)  
  
# Step 3: Mark all patients as hypertensive or not  
patients <- patients %>%  
  mutate(HYPERTENSION = ifelse(Id %in% htn_patients$PATIENT, 1, 0))  
  
# Step 4: Crude prevalence  
crude_prev <- mean(patients$HYPERTENSION) * 100  
cat("Crude Prevalence of Hypertension: ", round(crude_prev, 2), "%\n")
```

```
## Crude Prevalence of Hypertension: 28.17 %
```

```
library(dplyr)
library(lubridate)
library(ggplot2)
library(tidyr)

# Step 1: Age and age group
patients <- patients %>%
  mutate(BIRTHDATE = as.Date(BIRTHDATE),
         AGE = as.numeric(floor(interval(BIRTHDATE, Sys.Date()) / years(1))),
         age_group = case_when(
           AGE < 45 ~ "16-44",
           AGE < 65 ~ "45-64",
           TRUE ~ "65+"
         ))

# Step 2: UK HSE 2021 reference prevalence (from Table 12)
hse_prevalence <- tribble(
  ~age_group, ~GENDER, ~hse_prev,
  "16-44", "M", 0.11,
  "45-64", "M", 0.39,
  "65+", "M", 0.59,
  "16-44", "F", 0.08,
  "45-64", "F", 0.32,
  "65+", "F", 0.60
)

# Step 3: Synthea cohort's prevalence by age group and gender
cohort_prev <- patients %>%
  filter(GENDER %in% c("M", "F")) %>%
  group_by(age_group, GENDER) %>%
  summarise(cohort_prev = mean(HYPERTENSION), n = n(), .groups = "drop")

# Step 4: Join for comparison
comparison <- left_join(cohort_prev, hse_prevalence, by = c("age_group", "GENDER"))

# View comparison table
print(comparison)
```

```
## # A tibble: 6 x 5
##   age_group GENDER cohort_prev     n hse_prev
##   <chr>     <chr>      <dbl> <int>   <dbl>
## 1 16-44     F          0.304  158    0.08
## 2 16-44     M          0.166  247    0.11
## 3 45-64     F          0.340  159    0.32
## 4 45-64     M          0.349  126    0.39
## 5 65+      F          0.233  116    0.6
## 6 65+      M          0.387  124    0.59
```

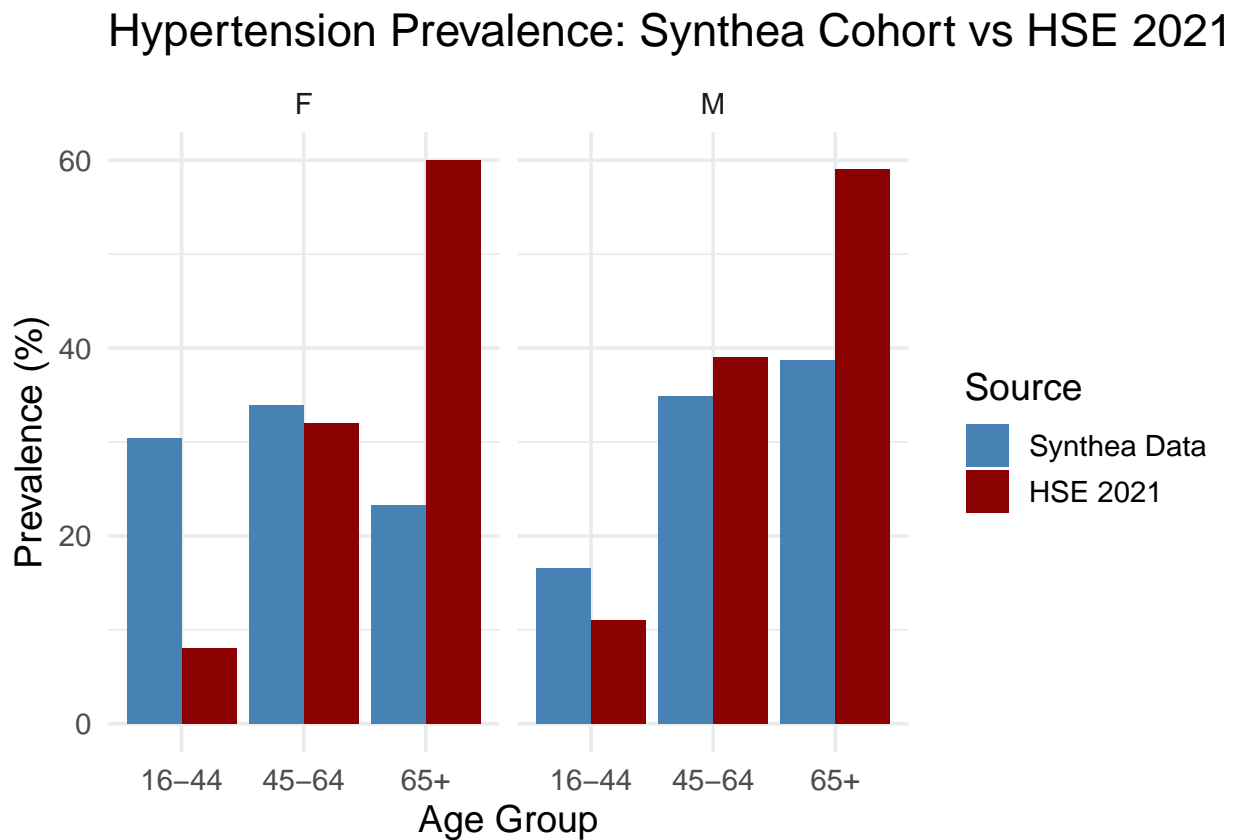
```
# Step 5: Reshape for plotting
plot_data <- comparison %>%
  pivot_longer(cols = c("cohort_prev", "hse_prev"),
```

```

names_to = "source", values_to = "prevalence")

# Step 6: Plot
ggplot(plot_data, aes(x = age_group, y = prevalence * 100, fill = source)) +
  geom_bar(stat = "identity", position = "dodge") +
  facet_wrap(~GENDER) +
  scale_fill_manual(values = c("cohort_prev" = "steelblue", "hse_prev" = "darkred"),
                    labels = c("Synthea Data", "HSE 2021")) +
  labs(title = "Hypertension Prevalence: Synthea Cohort vs HSE 2021",
       x = "Age Group", y = "Prevalence (%)", fill = "Source") +
  theme_minimal(base_size = 14)

```



Further Analysis: Stratified Summary and Plots

```

bp_bmi_summary <- bp_bmi_data %>%
  group_by(Type) %>%
  summarise(
    count = n(),
    mean = round(mean(VALUE_NUMERIC, na.rm = TRUE), 1),
    median = round(median(VALUE_NUMERIC, na.rm = TRUE), 1),
    sd = round(sd(VALUE_NUMERIC, na.rm = TRUE), 1),
    .groups = 'drop'
  )

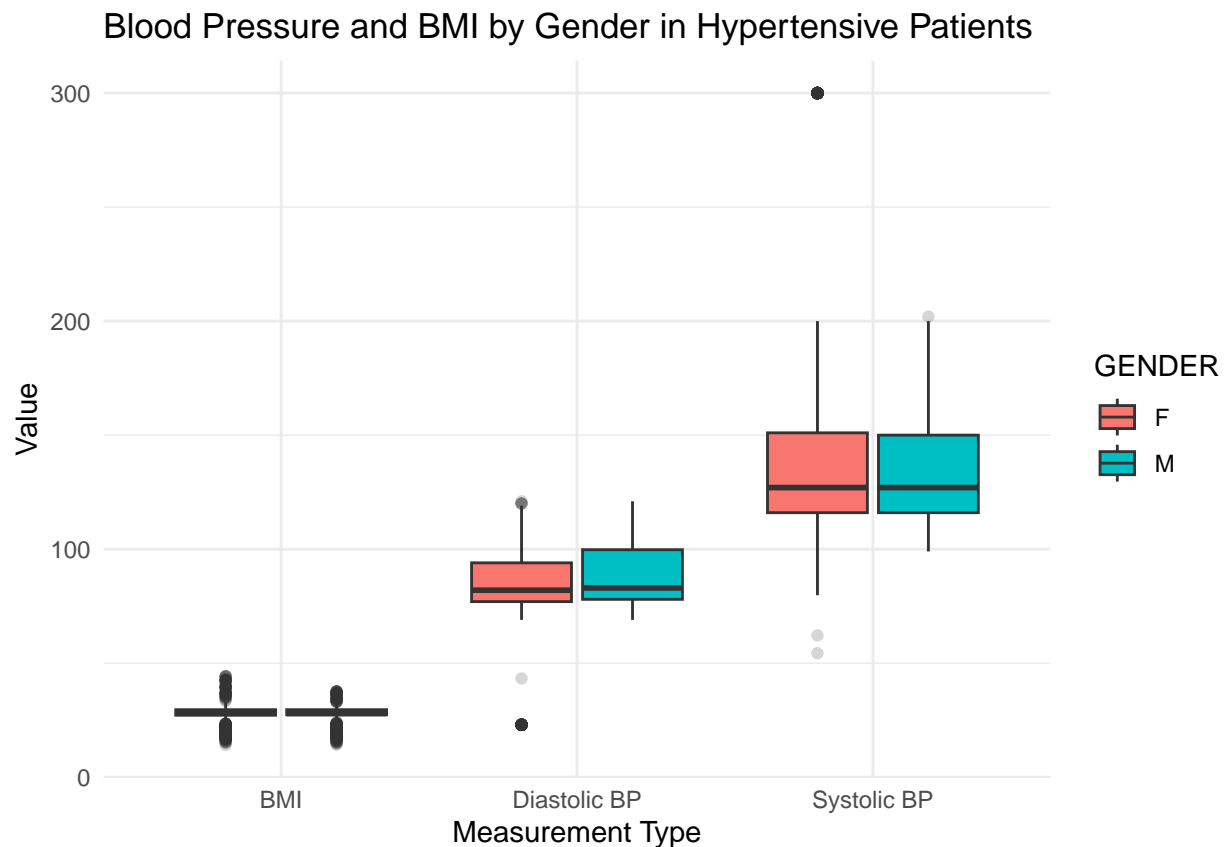
```

```
bp_bmi_summary
```

```
## # A tibble: 3 x 5
##   Type          count mean median   sd
##   <chr>         <int> <dbl> <dbl> <dbl>
## 1 BMI           3355  27.7  27.9   3.6
## 2 Diastolic BP  4026  86.1   83   15.7
## 3 Systolic BP  4026 138.   127   37.8
```

```
bp_bmi_data <- bp_bmi_data %>%
  left_join(patients %>% select(Id, GENDER), by = c("PATIENT" = "Id"))

ggplot(bp_bmi_data, aes(x = Type, y = VALUE_NUMERIC, fill = GENDER)) +
  geom_boxplot(outlier.alpha = 0.2) +
  labs(
    title = "Blood Pressure and BMI by Gender in Hypertensive Patients",
    x = "Measurement Type", y = "Value"
  ) +
  theme_minimal()
```



```
bp_bmi_data <- bp_bmi_data %>%
  left_join(patients %>% select(Id, AGE), by = c("PATIENT" = "Id")) %>%
  mutate(AgeGroup = cut(
    AGE,
```

```

breaks = c(0, 30, 50, 70, 120),
labels = c(" ≤30", "31-50", "51-70", "70+")
))

ggplot(bp_bmi_data, aes(x = Type, y = VALUE_NUMERIC, fill = AgeGroup)) +
  geom_boxplot(outlier.alpha = 0.2) +
  labs(
    title = "BP and BMI by Age Group in Hypertensive Patients",
    x = "Measurement Type", y = "Value"
  ) +
  theme_minimal()

```

