# Supplement 3

# Number of pain sites

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# 1 Import and check data

```
# Import
data <- read_rds('data-cleaned/data-pain-sites.rds')
demo <- read_rds('data-cleaned/data-demographics.rds')
# Check</pre>
```

```
## Pain sites
dim(data)
## [1] 596 21
names (data)
    [1] "ID"
                             "Head"
                                                  "Throat"
##
   [4] "Shoulder"
                             "Arms"
                                                  "Elbows"
   [7] "Wrists.Hands"
                             "Chest"
                                                  "Upper_back"
## [10] "Lower back"
                             "Abdomen"
                                                  "Cervical_spine"
                             "Lumbosacral_spine" "Groin"
## [13] "Thoracic_spine"
## [16] "Hips"
                             "Legs"
                                                  "Knees"
## [19] "Ankles.Feet"
                             "Buttocks"
                                                  "Site"
glimpse(data)
## Rows: 596
## Columns: 21
## $ ID
                        <chr> "RPB73", "RPB74", "RPB75", "RPB76", "RPB77", "RPB...
## $ Head
                        <chr> "No", "No", "No", "Yes", "Yes", "No", "No", "No", ...
                        <chr> "No", "No", "No", "No", "No", "No", "No", "No", "...
## $ Throat
## $ Shoulder
                        <chr> "No", "No", "No", "No", "No", "No", "No", "No", "...
                       <chr> "No", "No", "No", "No", "No", "No", "No", "No", "No", "...
## $ Arms
                        <chr> "No", "No", "No", "No", "No", "No", "No", "No", "No", "...
## $ Elbows
                        <chr> "No", "No", "No", "No", "No", "No", "No", "No", "No", "...
## $ Wrists.Hands
                       <chr> "No", "No", "No", "Yes", "No", "No", "No", "No", ...
## $ Chest
                        <chr> "No", "No", "No", "No", "No", "No", "No", "No", "...
## $ Upper_back
                        <chr> "No", "No", "No", "No", "No", "No", "No", "No", "No", "...
## $ Lower_back
                        <chr> "No", "No", "Yes", "Yes", "No", "No", "Yes", "No"...
## $ Abdomen
                       <chr> "No", "No", "No", "No", "No", "No", "No", "No", "No", "...
## $ Cervical spine
## $ Thoracic_spine
                       <chr> "No", "No", "No", "No", "No", "No", "No", "No", "...
## $ Lumbosacral spine <chr> "No", "No", "No", "No", "No", "No", "No", "No", "No", "...
## $ Groin
                       <chr> "No", "No", "No", "No", "No", "No", "No", "No", "...
                       <chr> "No", "No", "No", "No", "No", "No", "No", "Yes", ...
## $ Hips
                        <chr> "No", "No", "No", "No", "Yes", "No", "Yes", ...
## $ Legs
                        <chr> "No", "No", "No", "No", "No", "Yes", "No", "Yes",...
## $ Knees
                        <chr> "No", "No", "No", "No", "Yes", "No", "Yes",...
## $ Ankles.Feet
                       <chr> "No", "No", "No", "No", "No", "No", "No", "Yes", ...
## $ Buttocks
                        <chr> "RP", "RP", "RP", "RP", "RP", "RP", "RP", "RP", "RP", "...
## $ Site
data %>%
    select(-ID, -Site) %>%
    mutate_if(is.character, factor) %>%
    skim()
```

Table 1: Data summary

Name	Piped data
Number of rows	596
Number of columns	19
Column type frequency:	
factor	19
	_
Group variables	None

### Variable type: factor

skim_variable	n_missing	complete_rate	n_unique	top_counts
Head	0	1	2	No: 401, Yes: 195
Throat	0	1	2	No: 577, Yes: 19
Shoulder	0	1	2	No: 549, Yes: 47
Arms	0	1	2	No: 571, Yes: 25
Elbows	0	1	2	No: 574, Yes: 22
Wrists.Hands	0	1	2	No: 561, Yes: 35
Chest	0	1	2	No: 479, Yes: 117
$Upper\_back$	0	1	1	No: 596
Lower_back	0	1	2	No: 554, Yes: 42
Abdomen	0	1	2	No: 437, Yes: 159
Cervical_spine	0	1	2	No: 566, Yes: 30
Thoracic_spine	0	1	2	No: 522, Yes: 74
Lumbosacral_spine	0	1	2	No: 504, Yes: 92
Groin	0	1	2	No: 542, Yes: 54
Hips	0	1	2	No: 556, Yes: 40
Legs	0	1	2	No: 497, Yes: 99
Knees	0	1	2	No: 512, Yes: 84
Ankles.Feet	0	1	2	No: 412, Yes: 184
Buttocks	0	1	2	No: 577, Yes: 19

```
## Demographics
dim(demo)
## [1] 596
names (demo)
## [1] "ID"
                                                "Sex"
## [4] "Age"
                           "Employment_status" "CD4_recent"
## [7] "ART_currently"
                           "Education"
glimpse(demo)
## Rows: 596
## Columns: 8
                       <chr> "RPB73", "RPB74", "RPB75", "RPB76", "RPB77", "RPB...
## $ ID
                       <chr> "RP", "RP", "RP", "RP", "RP", "RP", "RP", "RP", "RP", "...
## $ Site
## $ Sex
                       <chr> "Female", "Female", "Female", "Female", "Female", ...
                       <dbl> 36, 27, 39, 36, 31, 32, 28, 37, 31, 25, 31, 24, 3...
## $ Age
## $ Employment_status <chr> "Other", "Unemployed", "Other", "Unemployed", "Un...
## $ CD4_recent
                       <dbl> 391, 571, 591, 207, 126, 225, 543, 410, 74, 212, ...
## $ ART_currently
                       <chr> "Yes", "Yes", "Yes", "Yes", "Yes", "Yes", "Yes", ...
## $ Education
                       <chr> "Tertiary", "Secondary", "Secondary", "Primary", ...
demo %>%
   select(-ID, -Site) %>%
   mutate_if(is.character, factor) %>%
   skim()
```

Table 3: Data summary

Name Piped data

Table 3: Data summary

Number of rows Number of columns	596 6
Column type frequency:	4
numeric  Group variables	2 None

## Variable type: factor

skim_variable	n_missing	complete_rate	n_unique	top_counts
Sex	0	1.00	2	Fem: 481, Mal: 115
Employment_status	49	0.92	4	Une: 330, Ful: 131, Par: 52, Oth: 34
ART_currently	5	0.99	2	Yes: 460, No: 131
Education	37	0.94	3	Sec: 395, Pri: 99, Ter: 65

#### Variable type: numeric

skim_variable	n_missing	$complete\_rate$	mean	$\operatorname{sd}$	p0	p25	p50	p75	p100
Age	8	0.99	37.28	9.06	19	31	36	42	76
$CD4\_recent$	99	0.83	320.71	238.92	1	155	261	432	1232

# 1.1 Add acute/chronic pain column to demographic data

Table 6: Data summary

Name Number of rows	Piped data 596		
Number of columns	1		
Column type frequency:			

Table 6: Data summary

factor	1
Group variables	None

## Variable type: factor

skim_variable	n_missing	complete_rate	n_unique	top_counts
Pain_def	0	1	2	Acu: 387, Chr: 209

# 2 Data analysis

## 2.1 Process data

# 2.2 Summary statistics

# 2.2.1 Total group

```
analysis_set %>%
select(Count) %>%
skim()
```

Table 8: Data summary

Name	Piped data
Number of rows	596
Number of columns	1
Column type frequency:	
numeric	1
	_
Group variables	None

# Variable type: numeric

skim_variable	n_missing	$complete\_rate$	mean	$\operatorname{sd}$	p0	p25	p50	p75	p100
Count	0	1	2.24	1.68	0	1	2	3	12

## 2.2.2 For each categorical variable

# 2.2.2.1 By sex

```
analysis_set %>%
  select(Count, Sex) %>%
  group_by(Sex) %>%
  skim()
```

Table 10: Data summary

Name	Piped data
Number of rows	596
Number of columns	2
Column type frequency:	
numeric	1
Group variables	Sex

# Variable type: numeric

skim_variable	Sex	n_missing	complete_rate	mean	$\operatorname{sd}$	p0	p25	p50	p75	p100
Count	Female	0	1	2.27	1.69	0	1	2	3	12
Count	Male	0	1	2.13	1.63	0	1	2	3	9

# 2.2.2.2 By HAART

```
analysis_set %>%
select(Count, ART_currently) %>%
filter(complete.cases(.)) %>%
group_by(ART_currently) %>%
skim()
```

Table 12: Data summary

Name	Piped data
Number of rows	591
Number of columns	2
Column type frequency:	
numeric	1
Group variables	ART_currently

## Variable type: numeric

skim_variable	ART_currently	n_missing	$complete\_rate$	mean	$\operatorname{sd}$	p0	p25	p50	p75	p100
Count	No	0	1	2.18	1.64	0	1	2	3	12
Count	Yes	0	1	2.27	1.70	0	1	2	3	9

## 2.2.2.3 By education

```
analysis_set %>%
  select(Count, Education) %>%
  filter(complete.cases(.)) %>%
  group_by(Education) %>%
  skim()
```

Table 14: Data summary

Name	Piped data
Number of rows	559
Number of columns	2
Column type frequency:	
numeric	1
Group variables	Education

## Variable type: numeric

skim_variable	Education	n_missing	complete_rate	mean	$\operatorname{sd}$	p0	p25	p50	p75	p100
Count	Primary	0	1	2.38	1.64	0	1	2	3	8
Count	Secondary	0	1	2.24	1.73	0	1	2	3	12
Count	Tertiary	0	1	2.06	1.52	0	1	2	3	6

## 2.2.2.4 By employment status

```
analysis_set %>%
  select(Count, Employment_status) %>%
  filter(complete.cases(.)) %>%
  group_by(Employment_status) %>%
  skim()
```

Table 16: Data summary

Name	Piped data
Number of rows	547
Number of columns	2
Column type frequency: numeric	1
Group variables	Employment_status

### Variable type: numeric

skim_variable	Employment_status	n_missing	complete_rate	mean	$\operatorname{sd}$	p0	p25	p50	p75	p100
Count	Full-time work	0	1	2.21	1.46	0	1	2	3	6
Count	Other	0	1	2.18	1.64	0	1	2	4	5
Count	Part-time work	0	1	2.50	1.89	0	1	2	3	9
Count	Unemployed	0	1	2.21	1.75	0	1	2	3	12

### 2.2.2.5 By pain definition

```
analysis_set %>%
  select(Count, Pain_def) %>%
  group_by(Pain_def) %>%
  skim()
```

Table 18: Data summary

Name	Piped data
Number of rows	596
Number of columns	2
Column type frequency:	
numeric	1
Group variables	Pain_def

#### Variable type: numeric

skim_variable	Pain_def	$n_{missing}$	$complete\_rate$	mean	$\operatorname{sd}$	p0	p25	p50	p75	p100
Count	Acute	0	1	2.19	1.74	0	1	2	3	12
Count	Chronic	0	1	2.33	1.57	0	1	2	3	9

# 2.3 Exploratory plots

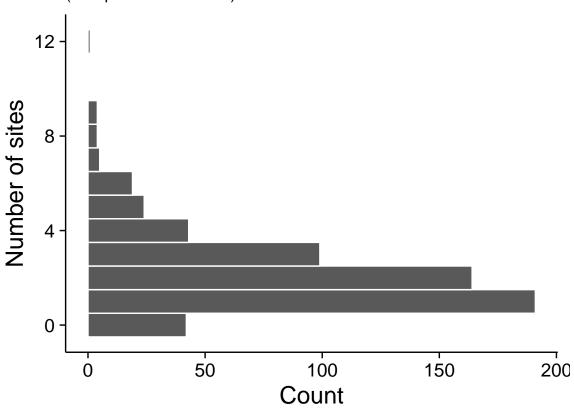
# 2.3.1 Distribution of pain counts

### 2.3.1.1 Overall count frequency

```
theme(plot.title = element_text(size = 18),
    plot.subtitle = element_text(size = 12),
    panel.grid = element_blank(),
    axis.text = element_text(colour = '#000000'),
    axis.line = element_line(size = 0.5),
    axis.ticks = element_line(size = 0.5)); all
```

# All participants

(Complete cases = 596)

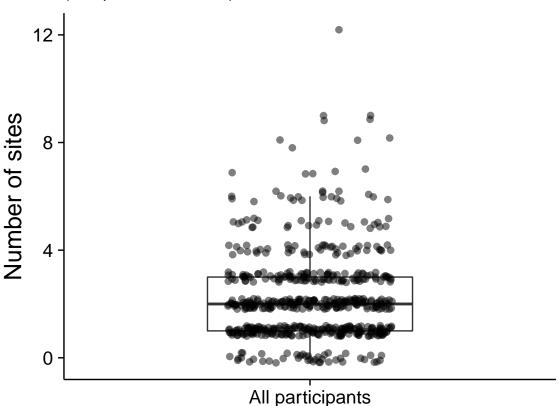


```
all <- analysis_set %>%
    select(Count) %>%
   filter(complete.cases(.)) %>%
   ggplot(data = .) +
   aes(y = Count,
        x = 'All participants') +
    geom_boxplot(outlier.colour = '#FFFFFF',
                 outlier.size = 0,
                 width = 0.5) +
    geom_point(size = 2,
               shape = 21,
               position = position_jitter(height = 0.2, width = 0.2),
               fill = '#000000',
               colour = '#000000',
               alpha = 0.5) +
   labs(title = 'All participants',
         subtitle = str_glue('(Complete cases = {nrow(analysis_set[!is.na(
                             analysis_set$Count), ])})'),
```

```
y = 'Number of sites') +
theme_minimal(base_size = 18) +
theme(plot.title = element_text(size = 18),
    plot.subtitle = element_text(size = 12),
    plot.caption = element_text(size = 12),
    panel.grid = element_blank(),
    axis.title.x = element_blank(),
    axis.title.x = element_text(colour = '#000000'),
    axis.line = element_line(size = 0.5),
    axis.ticks = element_line(size = 0.5)); all
```

# All participants

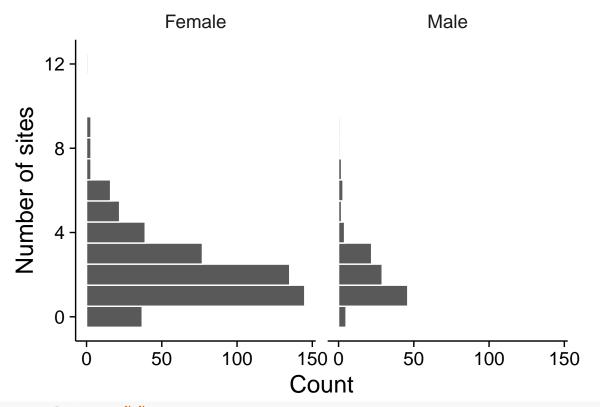
(Complete cases = 596)



## 2.3.1.2 Count by sex

# Sex

(Complete cases = 596)

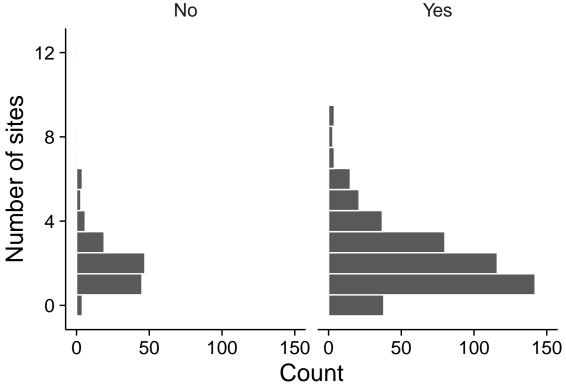


### 2.3.1.3 Count by HAART

```
analysis_set %>%
    select(Count, ART_currently) %>%
    filter(complete.cases(.)) %>%
   ggplot(data = .) +
   aes(Count) +
   geom_histogram(binwidth = 1,
                   colour = '#FFFFFF') +
    scale_y_continuous(limits = c(0, 150),
                      breaks = c(0, 50, 100, 150)) +
   labs(title = 'Currently on HAART',
         subtitle = str_glue('(Complete cases = {nrow(analysis_set[!is.na(
                             analysis_set$ART_currently), ])})'),
         x = 'Number of sites',
         y = 'Count') +
   facet_wrap(~ART_currently) +
    coord_flip() +
   theme_minimal(base_size = 18) +
   theme(plot.title = element_text(size = 18),
         plot.subtitle = element_text(size = 12),
          panel.grid = element_blank(),
          axis.text = element_text(colour = '#000000'),
          axis.line = element line(size = 0.5),
          axis.ticks = element_line(size = 0.5))
```

# Currently on HAART

(Complete cases = 591)



```
haart <- analysis_set %>%
    select(Count, ART_currently) %>%
   filter(complete.cases(.)) %>%
   ggplot(data = .) +
    aes(y = Count,
        x = ART_currently) +
   geom boxplot(outlier.colour = '#FFFFFF',
                 outlier.size = 0) +
   geom_point(size = 2,
               shape = 21,
               position = position_jitter(height = 0.2, width = 0.2),
               fill = '#000000',
               colour = '#000000',
               alpha = 0.5) +
    scale_x_discrete(labels = c('No', 'Yes')) +
   labs(title = 'Currently on HAART',
         subtitle = str_glue('(Complete cases = {nrow(analysis_set[!is.na(
                             analysis_set$ART_currently), ])})'),
         y = 'Number of sites') +
   theme_minimal(base_size = 18) +
    theme(plot.title = element_text(size = 18),
          plot.subtitle = element_text(size = 12),
          plot.caption = element_text(size = 12),
          panel.grid = element_blank(),
          axis.title.x = element_blank(),
```

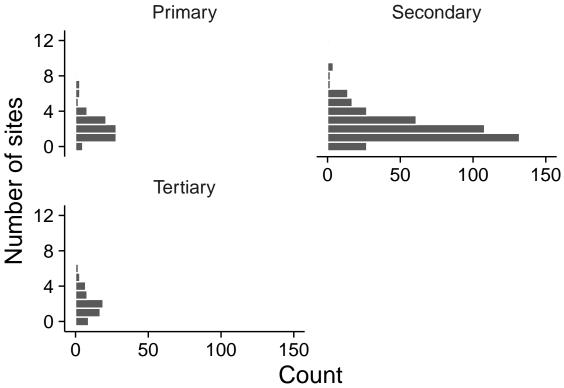
```
axis.text = element_text(colour = '#000000'),
axis.line = element_line(size = 0.5),
axis.ticks = element_line(size = 0.5))
```

#### 2.3.1.4 Count education

```
analysis_set %>%
    select(Count, Education) %>%
   filter(complete.cases(.)) %>%
   ggplot(data = .) +
   aes(Count) +
   geom_histogram(binwidth = 1,
                  colour = '#FFFFFF') +
    scale_y_continuous(limits = c(0, 150),
                       breaks = c(0, 50, 100, 150)) +
   labs(title = 'Level of education',
        subtitle = str_glue('(Complete cases = {nrow(analysis_set[!is.na(
                             analysis_set$Education), ])})'),
        x = 'Number of sites',
        y = 'Count') +
   facet_wrap(~Education, ncol = 2) +
    coord flip() +
   theme_minimal(base_size = 18) +
   theme(plot.title = element_text(size = 18),
         plot.subtitle = element_text(size = 12),
         panel.grid = element_blank(),
         axis.text = element_text(colour = '#000000'),
         axis.line = element_line(size = 0.5),
         axis.ticks = element_line(size = 0.5))
```

# Level of education

(Complete cases = 559)



```
edu <- analysis_set %>%
    select(Count, Education) %>%
   filter(complete.cases(.)) %>%
   ggplot(data = .) +
    aes(y = Count,
        x = Education) +
   geom boxplot(outlier.colour = '#FFFFFF',
                 outlier.size = 0) +
   geom_point(size = 2,
               shape = 21,
               position = position_jitter(height = 0.2, width = 0.2),
               fill = '#000000',
               colour = '#000000',
               alpha = 0.5) +
    scale_x_discrete(labels = c('Primary', 'Secondary',
                                'Tertiary')) +
   labs(title = 'Level of Education',
         subtitle = str_glue('(Complete cases = {nrow(analysis_set[!is.na(
                             analysis_set$Education), ])})'),
         y = 'Number of sites') +
   theme_minimal(base_size = 18) +
    theme(plot.title = element_text(size = 18),
          plot.subtitle = element_text(size = 12),
          plot.caption = element_text(size = 12),
          panel.grid = element_blank(),
```

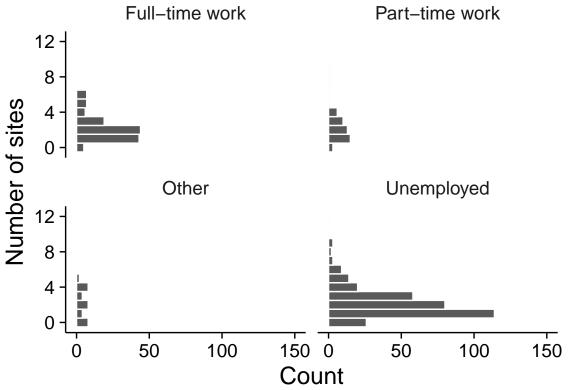
```
axis.title.x = element_blank(),
axis.text = element_text(colour = '#000000'),
axis.line = element_line(size = 0.5),
axis.ticks = element_line(size = 0.5))
```

## 2.3.1.5 Count by employment status

```
analysis_set %>%
   select(Count, Employment_status) %>%
   filter(complete.cases(.)) %>%
   mutate(Employment_status = factor(Employment_status,
                                      levels = c('Full-time work',
                                                 'Part-time work',
                                                 'Other',
                                                 'Unemployed'),
                                      ordered = TRUE)) %>%
   ggplot(data = .) +
   aes(Count) +
   geom_histogram(binwidth = 1,
                   colour = '#FFFFFF') +
   scale_y_continuous(limits = c(0, 150),
                       breaks = c(0, 50, 100, 150)) +
   labs(title = 'Employment status',
         subtitle = str_glue('(Complete cases = {nrow(analysis_set[!is.na(
                             analysis_set$Employment_status), ])})'),
         x = 'Number of sites',
         y = 'Count') +
   facet_wrap(~Employment_status, ncol = 2) +
    coord_flip() +
   theme_minimal(base_size = 18) +
   theme(plot.title = element_text(size = 18),
         plot.subtitle = element_text(size = 12),
          panel.grid = element_blank(),
          axis.text = element_text(colour = '#000000'),
          axis.line = element_line(size = 0.5),
          axis.ticks = element_line(size = 0.5))
```

# **Employment status**

(Complete cases = 547)



```
employment <- analysis_set %>%
    select(Count, Employment_status) %>%
    filter(complete.cases(.)) %>%
   mutate(Employment_status = factor(Employment_status,
                                      levels = c('Full-time work',
                                                  'Part-time work',
                                                  'Other',
                                                  'Unemployed'),
                                      ordered = TRUE)) %>%
    ggplot(data = .) +
    aes(y = Count,
        x = Employment_status) +
   geom_boxplot(outlier.colour = '#FFFFFF',
                 outlier.size = 0) +
   geom_point(size = 2,
               position = position_jitter(height = 0.2, width = 0.2),
               fill = '#000000',
               colour = '#000000',
               alpha = 0.5) +
    scale_x_discrete(labels = c('Full-time\nwork', 'Part-time\nwork',
                                 'Other', 'Unemployed')) +
   labs(title = 'Employment status',
         subtitle = str_glue('(Complete cases = {nrow(analysis_set[!is.na(
                             analysis_set$Employment_status), ])})'),
```

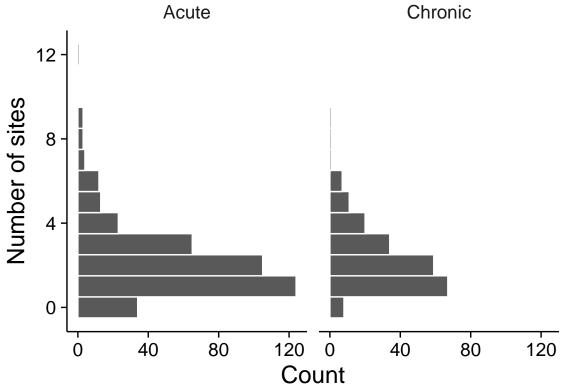
```
caption = 'Other: receiving social grants or were students',
    y = 'Number of sites') +
theme_minimal(base_size = 18) +
theme(plot.title = element_text(size = 18),
    plot.subtitle = element_text(size = 12),
    plot.caption = element_text(size = 12),
    panel.grid = element_blank(),
    axis.title.x = element_blank(),
    axis.text = element_text(colour = '#000000'),
    axis.ticks = element_line(size = 0.5),
    axis.ticks = element_line(size = 0.5))
```

### 2.3.1.6 Count by pain definition

```
analysis_set %>%
    select(Count, Pain_def) %>%
   filter(complete.cases(.)) %>%
   ggplot(data = .) +
   aes(Count) +
   geom_histogram(binwidth = 1,
                   colour = '#FFFFFF') +
   labs(title = 'Pain definition',
         subtitle = str_glue('(Complete cases = {nrow(analysis_set[!is.na(
                             analysis_set$Pain_def), ])})'),
         x = 'Number of sites',
         y = 'Count') +
   facet_wrap(~Pain_def) +
   coord_flip() +
   theme_minimal(base_size = 18) +
   theme(plot.title = element_text(size = 18),
          plot.subtitle = element_text(size = 12),
          panel.grid = element_blank(),
          axis.text = element_text(colour = '#000000'),
          axis.line = element_line(size = 0.5),
          axis.ticks = element_line(size = 0.5))
```

# Pain definition

(Complete cases = 596)



```
pain_def <- analysis_set %>%
    select(Count, Pain_def) %>%
   filter(complete.cases(.)) %>%
   ggplot(data = .) +
    aes(y = Count,
        x = Pain_def) +
   geom boxplot(outlier.colour = '#FFFFFF',
                 outlier.size = 0) +
   geom_point(size = 2,
               shape = 21,
               position = position_jitter(height = 0.2, width = 0.2),
               fill = '#000000',
               colour = '#000000',
               alpha = 0.5) +
    scale_x_discrete(labels = c('Acute', 'Chronic')) +
   labs(title = 'Pain definition',
         subtitle = str_glue('(Complete cases = {nrow(analysis_set[!is.na(
                             analysis_set$Pain_def), ])})'),
         y = 'Number of sites') +
   theme_minimal(base_size = 18) +
   theme(plot.title = element_text(size = 18),
          plot.subtitle = element_text(size = 12),
          plot.caption = element_text(size = 12),
          panel.grid = element_blank(),
          axis.title.x = element_blank(),
```

```
axis.text = element_text(colour = '#0000000'),
axis.line = element_line(size = 0.5),
axis.ticks = element_line(size = 0.5))
```

## 2.4 Regression analysis

#### 2.4.1 Process data

```
analysis_set <- analysis_set %>%
  # Re-factor Employment_status
mutate(Employment_status = case_when(
          Employment_status == 'Unemployed' ~ ' 1_unemployed',
          Employment_status == 'Part-time work' ~ ' 3_employed PT',
          Employment_status == 'Full-time work' ~ ' 2_employed FT',
          Employment_status == 'Other' ~ ' 4_other'
)) %>%
filter(complete.cases(.))
```

#### 2.4.2 Mixed model negative binomial regression for count data

```
# Null model
null <- glmer.nb(Count ~ 1 +</pre>
                        (1|Site),
                    data = analysis_set)
# Full model
mod.mmnb <- glmer.nb(Count ~ Sex +</pre>
                        # Center and scale age
                        scale(Age) +
                        # Center and scale CD4
                        scale(CD4_recent) +
                        Employment_status +
                        Education +
                        ART_currently +
                        Pain_def +
                        (1|Site),
                    data = analysis_set)
## Fit is singular, print model to check SD of random effect
summary(mod.mmnb)
```

```
## Generalized linear mixed model fit by maximum likelihood (Laplace
    Approximation) [glmerMod]
## Family: Negative Binomial(9.4933) ( log )
## Formula: Count ~ Sex + scale(Age) + scale(CD4_recent) + Employment_status +
      Education + ART_currently + Pain_def + (1 | Site)
##
     Data: analysis_set
##
##
       AIC
                      logLik deviance df.resid
                BIC
    1608.4
##
            1661.2 -791.2
                              1582.4
                                           417
##
## Scaled residuals:
                               ЗQ
##
      Min
           1Q Median
                                      Max
```

```
## -1.5293 -0.7298 -0.1483 0.4427 6.0224
##
## Random effects:
                      Variance Std.Dev.
## Groups Name
           (Intercept) 4.952e-12 2.225e-06
## Site
## Number of obs: 430, groups: Site, 5
## Fixed effects:
##
                                  Estimate Std. Error z value Pr(>|z|)
                                              0.11990 7.229 4.88e-13 ***
## (Intercept)
                                   0.86675
## SexMale
                                  -0.11757
                                              0.10083 -1.166
                                                                0.2436
## scale(Age)
                                   0.06919
                                              0.04101
                                                        1.687
                                                                0.0916
## scale(CD4_recent)
                                   0.01573
                                              0.03825
                                                        0.411
                                                                0.6809
## Employment_status 2_employed FT -0.01663
                                              0.09150 -0.182
                                                                0.8558
## Employment_status 3_employed PT 0.18433
                                                       1.532
                                                                0.1256
                                              0.12035
## Employment_status 4_other
                                   0.02330
                                              0.14169
                                                        0.164
                                                                0.8694
                                              0.09809 -0.610
## EducationSecondary
                                  -0.05979
                                                                0.5422
## EducationTertiary
                                  -0.20110
                                              0.14198 -1.416
                                                                0.1567
## ART_currentlyYes
                                   0.01669
                                              0.09858
                                                        0.169
                                                                0.8656
## Pain defChronic
                                   0.08157
                                              0.09758
                                                        0.836
                                                                0.4032
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##
              (Intr) SexMal scl(A) s(CD4_ E_2_FT E_3_PT Emp_4_ EdctnS EdctnT
## SexMale
              -0.141
## scale(Age) -0.051 -0.149
## scl(CD4_rc) 0.023 0.191 -0.033
## Emplym_2_FT -0.158 -0.009 -0.087 -0.035
## Emplym_3_PT -0.178  0.001 -0.003 -0.018  0.224
## Emplymnt_4_ -0.079  0.038  0.049 -0.047  0.148
## EdctnScndry -0.646 0.038 0.221 0.047 -0.088 0.050 -0.070
## EductnTrtry -0.399 0.061 0.220 0.016 -0.148 0.032 -0.132 0.586
## ART_crrntly -0.619 0.029 0.019 0.032 0.105 0.036 -0.027 0.013 -0.017
## Pan_dfChrnc 0.029 -0.162 -0.377 -0.334 -0.165 -0.108 0.092 -0.063 -0.153
              ART cY
## SexMale
## scale(Age)
## scl(CD4_rc)
## Emplym_2_FT
## Emplym_3_PT
## Emplymnt_4_
## EdctnScndry
## EductnTrtry
## ART_crrntlY
## Pan_dfChrnc -0.297
## convergence code: 0
## boundary (singular) fit: see ?isSingular
```

## 2.4.3 Negative binomial regression for count data

```
# Full model
mod.nb <- glm.nb(Count ~ Sex +</pre>
                     # Center and scale age
                     scale(Age) +
                     # Center and scale CD4
                     scale(CD4_recent) +
                     Employment_status +
                     Education +
                     ART_currently +
                     Pain_def,
                 data = analysis_set)
# Compare models
anova(null, mod.nb)
## Likelihood ratio tests of Negative Binomial Models
## Response: Count
##
                                                                                                 Model
## 1
## 2 Sex + scale(Age) + scale(CD4_recent) + Employment_status + Education + ART_currently + Pain_def
       theta Resid. df
                                           Test
                           2 x log-lik.
                                                   df LR stat.
                                                                 Pr(Chi)
## 1 8.192033
                    429
                              -1596.185
## 2 9.493334
                    419
                              -1582.363 1 vs 2
                                                  10 13.82139 0.1812956
# Summary of coefficients
cbind(Estimate = coef(mod.nb)[-1],
      confint(mod.nb)[-1, ],
      summary(mod.nb)$coefficients[-1, 3:4]) %>%
   kable(caption = 'Coefficients and 95% CI',
          digits = 3,
          col.names = c('Estimate', 'Lower 95%CI', 'Upper 95%CI',
                        'z-value', 'P-value'))
```

Table 20: Coefficients and 95% CI

	Estimate	Lower 95%CI	Upper $95\%$ CI	z-value	P-value
SexMale	-0.118	-0.317	0.078	-1.166	0.243
scale(Age)	0.069	-0.012	0.149	1.692	0.091
scale(CD4_recent)	0.016	-0.060	0.090	0.413	0.680
Employment_status 2_employed FT	-0.017	-0.197	0.162	-0.181	0.856
Employment_status 3_employed PT	0.184	-0.055	0.417	1.528	0.127
Employment_status 4_other	0.023	-0.260	0.296	0.164	0.869
EducationSecondary	-0.060	-0.250	0.134	-0.609	0.542
EducationTertiary	-0.201	-0.481	0.076	-1.418	0.156
ART_currentlyYes	0.017	-0.175	0.212	0.169	0.866
Pain_defChronic	0.082	-0.110	0.272	0.842	0.400

### 2.5 Publication plot

```
# Generate plots for age and CD4
## Count by age
age <- analysis_set %>%
```

```
select(Count, Age) %>%
   filter(complete.cases(.)) %>%
   ggplot(data = .) +
   aes(y = Count,
        x = Age) +
   geom_point(size = 2,
              shape = 21,
              position = position jitter(height = 0.2, width = 0.2),
              fill = '#000000',
               colour = '#000000',
               alpha = 0.5) +
   labs(title = 'Age',
         subtitle = str_glue('(Complete cases = {nrow(analysis_set[!is.na(
                             analysis_set$Age), ])})'),
        x = 'Age (years)',
        y = 'Number of sites') +
    scale_x_continuous(limits = c(18, 80),
                      breaks = seq(20, 80, by = 20)) +
   theme_minimal(base_size = 18) +
   theme(plot.title = element_text(size = 18),
          plot.subtitle = element_text(size = 12),
         plot.caption = element_text(size = 12),
         panel.grid = element_blank(),
         axis.text = element_text(colour = '#000000'),
         axis.line = element line(size = 0.5),
         axis.ticks = element_line(size = 0.5))
## Count by CD4 T-cell count
cd4 <- analysis_set %>%
   select(Count, CD4_recent) %>%
   filter(complete.cases(.)) %>%
   ggplot(data = .) +
   aes(y = Count,
       x = CD4_recent) +
   geom_point(size = 2,
               shape = 21,
              position = position_jitter(height = 0.2, width = 0.2),
              fill = '#000000',
               colour = '#000000',
              alpha = 0.5) +
   labs(title = 'Recent CD4 T-cell count',
         subtitle = str_glue('(Complete cases = {nrow(analysis_set[!is.na(
                             analysis_set$CD4_recent), ])})'),
        x = expression('CD4 T-cell count (cells.mm'^-3*')'),
        y = 'Number of sites') +
   scale_x_continuous(limits = c(0, 1400),
                       breaks = seq(0, 1400, by = 400)) +
   scale_y_continuous(breaks = seq(0, 12, 4)) +
   theme_minimal(base_size = 18) +
   theme(plot.title = element_text(size = 18),
         plot.subtitle = element_text(size = 12),
         plot.caption = element_text(size = 12),
         panel.grid = element_blank(),
```

```
axis.text = element_text(colour = '#000000'),
          axis.line = element_line(size = 0.5),
          axis.ticks = element_line(size = 0.5))
# Fix figures for patchwork plot
all2 <- all
sex2 <- sex +
 theme(axis.title.y = element_blank())
haart2 <- haart
employment2 <- employment</pre>
edu2 <- edu +
 theme(axis.title.y = element_blank())
pain_def2 <- pain_def +</pre>
 theme(axis.title.y = element_blank())
age2 <- age
cd42 <- cd4 +
 theme(axis.title.y = element_blank())
pubs <- all2 + sex2 + haart2 + edu2 + employment2 + pain_def2 + age2 + cd42 +</pre>
 plot_layout(ncol = 2)
ggsave(filename = 'figures/figure_3.png',
       width = 10,
       height = 18)
```

# 3 Session information

#### sessionInfo()

```
## R version 3.6.3 (2020-02-29)
## Platform: x86_64-pc-linux-gnu (64-bit)
## Running under: Debian GNU/Linux 10 (buster)
##
## Matrix products: default
## BLAS/LAPACK: /usr/lib/x86_64-linux-gnu/libopenblasp-r0.3.5.so
##
## locale:
   [1] LC CTYPE=en US.UTF-8
                                   LC NUMERIC=C
##
   [3] LC TIME=en US.UTF-8
                                   LC COLLATE=en US.UTF-8
##
   [5] LC MONETARY=en US.UTF-8
                                   LC MESSAGES=C
##
   [7] LC_PAPER=en_US.UTF-8
                                   LC_NAME=C
##
  [9] LC_ADDRESS=C
                                   LC TELEPHONE=C
## [11] LC MEASUREMENT=en US.UTF-8 LC IDENTIFICATION=C
## attached base packages:
## [1] stats
                 graphics grDevices utils
                                               datasets methods
                                                                    base
##
## other attached packages:
   [1] patchwork_1.0.0 knitr_1.28
                                        skimr_2.1.1
                                                         forcats 0.5.0
   [5] stringr_1.4.0
                                                         readr_1.3.1
##
                        dplyr_0.8.5
                                        purrr_0.3.4
   [9] tidyr 1.0.2
                        tibble_3.0.1
                                        ggplot2_3.3.0
                                                         tidyverse_1.3.0
## [13] MASS_7.3-51.5
                        lme4_1.1-23
                                        Matrix_1.2-18
## loaded via a namespace (and not attached):
   [1] Rcpp 1.0.4.6
                         lubridate 1.7.8 lattice 0.20-38 utf8 1.1.4
                                          R6 2.4.1
##
   [5] assertthat_0.2.1 digest_0.6.25
                                                            cellranger_1.1.0
## [9] repr_1.1.0
                         backports_1.1.6 reprex_0.3.0
                                                            evaluate 0.14
## [13] highr_0.8
                         httr_1.4.1
                                          pillar_1.4.3
                                                            rlang_0.4.5
## [17] readxl_1.3.1
                         minqa_1.2.4
                                          rstudioapi_0.11 nloptr_1.2.2.1
                                          splines 3.6.3
## [21] rmarkdown 2.1
                         labeling 0.3
                                                            statmod 1.4.34
## [25] munsell_0.5.0
                         broom 0.5.6
                                          compiler_3.6.3
                                                            modelr 0.1.6
## [29] xfun_0.13
                         base64enc_0.1-3
                                          pkgconfig_2.0.3
                                                            htmltools_0.4.0
## [33] tidyselect_1.0.0 fansi_0.4.1
                                          crayon_1.3.4
                                                            dbplyr_1.4.3
## [37] withr_2.2.0
                         grid_3.6.3
                                          nlme_3.1-144
                                                            jsonlite_1.6.1
## [41] gtable_0.3.0
                         lifecycle_0.2.0
                                          DBI_1.1.0
                                                            magrittr_1.5
## [45] scales_1.1.0
                         cli_2.0.2
                                          stringi_1.4.6
                                                            farver_2.0.3
## [49] fs_1.4.1
                         xm12_1.3.2
                                          ellipsis_0.3.0
                                                            generics_0.0.2
## [53] vctrs_0.2.4
                         boot_1.3-24
                                          tools_3.6.3
                                                            glue_1.4.0
## [57] hms_0.5.3
                         yaml_2.2.1
                                          colorspace_1.4-1 rvest_0.3.5
## [61] haven_2.2.0
```

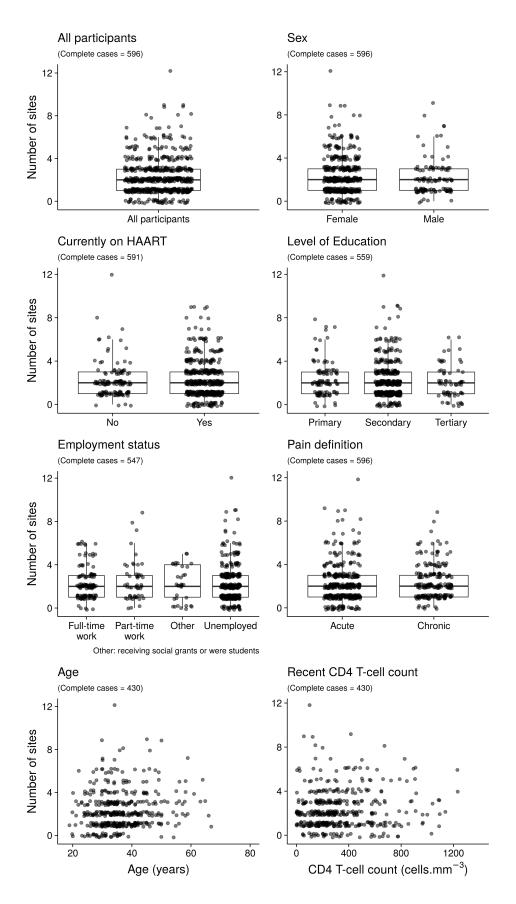


Figure 1: Number of pain sites by predictor variable  $\phantom{0}26$