Supplement 4

Analysis of the primary outcome

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The primary outcome was the reduction in pain severity score (PSS) between week 0 (BL) and week 48 (Wk48). The PSS is a composite measure of the mean of: pain at the assessment, and worst, least, and average pain in the last week.

The baseline observation carried forward (BOCF) analysis used data from week 0 to interpolate missing week 48 data. Some participants did not arrive for their week 0 appointment and therefore did not have data at baseline. These missing data were interpolated using the week 0 study site average PSS. No heed was paid to whether data were available at intermediate time points between week 0 and week 48.

The per protocol (PP) analysis included all individuals with complete data (i.e., no interpolation required).

Statistical significance was assessed using a permutation test that tested the independence of the therapeutic relationship only group (T, control) and the P groups (positive-living programme and therapeutic relationship). The conditional null distribution of the test statistic was calculated using Monte Carlo resampling (n = 100000).

Dosage was calculated as the number of assessments attended, on average) by participants in each group, and the number of intervention sessions attended by participasnts in group P and the change in NRS. *Note that the latter analysis expludes site U1 because the data were inadvertantly destroyed.*

Import and sort data

```
# Get BPI data
## BPI
bpi <- read_rds('data-cleaned/bpi.rds') %>%
        select(ID, contains('Pain'))
```

```
## Demographics
```

```
demo <- read_rds('data-cleaned/demographics.rds') %>%
    select(ID, Study_site, Group)
## Dosage
dosage <- read_rds('data-cleaned/dosage.rds')
## Join BPI and demographics
data <- demo %>%
    left_join(bpi)
## Add dosage data
data %<>% left_join(dosage)
## Primary outcome
primary <- data %>%
```

```
select(-contains('present'))
```

Quick look

```
glimpse(primary)
```

```
## Observations: 160
## Variables: 28
                      <chr> "J1", "J3", "J4", "J5", "J6", "J7", "J9", "J...
## $ ID
                      <chr> "U1", "U1", "U1", "U1", "U1", "U1", "U1", "U1...
## $ Study_site
                      <chr> "P", "T", "P", "P", "T", "T", "T", "P",...
## $ Group
                      <int> 8, 9, 5, 7, 7, 8, 10, 10, 9, 0, 10, NA, 9, 8...
## $ Worst_pain.BL
## $ Worst pain.Wk4
                      <int> NA, NA, 3, NA, NA, 8, 8, 9, 8, NA, NA, NA, 8...
                      <int> NA, NA, O, NA, NA, NA, 8, 9, 10, 4, NA, NA, ...
## $ Worst_pain.Wk8
## $ Worst_pain.Wk12
                      <int> NA, NA, 3, NA, NA, NA, 7, 9, 10, 7, 10, NA, ...
## $ Worst_pain.Wk24
                      <int> NA, NA, 6, NA, NA, NA, 7, 9, NA, 6, NA, NA, ...
## $ Worst_pain.Wk48
                      <int> NA, NA, 6, NA, NA, NA, 7, 8, NA, 8, NA, NA, ...
## $ Least_pain.BL
                      <int> 4, 1, 1, 10, 2, 3, 3, 5, 3, 0, 3, NA, 3, 3, ...
## $ Least_pain.Wk4
                      <int> NA, NA, 1, NA, NA, 3, 5, 4, 2, NA, NA, NA, 3...
                      <int> NA, NA, O, NA, NA, NA, 3, 4, 5, 1, NA, NA, 4...
## $ Least_pain.Wk8
## $ Least_pain.Wk12
                      <int> NA, NA, 1, NA, NA, NA, 3, 6, 5, 4, 5, NA, NA...
## $ Least_pain.Wk24
                      <int> NA, NA, 2, NA, NA, NA, 3, 5, NA, 2, NA, NA, ...
## $ Least_pain.Wk48
                      <int> NA, NA, 2, NA, NA, NA, 3, 5, NA, 2, NA, NA, ...
                      <int> 4, 4, 3, 5, 4, 6, 6, 7, 6, 0, 6, NA, 6, 6, 5...
## $ Average_pain.BL
## $ Average_pain.Wk4
                      <int> NA, NA, 1, NA, NA, 5, 5, 6, 5, NA, NA, NA, 5...
## $ Average_pain.Wk8
                      <int> NA, NA, O, NA, NA, NA, 5, 7, 8, 2, NA, NA, 6...
## $ Average_pain.Wk12 <int> NA, NA, 2, NA, NA, NA, 4, 7, 7, 6, 8, NA, NA...
## $ Average pain.Wk24 <int> NA, NA, 4, NA, NA, NA, 5, 7, NA, 4, NA, NA, ...
## $ Average_pain.Wk48 <int> NA, NA, 4, NA, NA, NA, 5, 7, NA, 5, NA, NA, ...
## $ Pain now.BL
                      <int> 6, 4, 0, 5, 2, 9, 0, 0, 9, 0, 3, NA, 0, 8, 6...
## $ Pain_now.Wk4
                      <int> NA, NA, O, NA, NA, 8, 8, 4, 6, NA, NA, NA, O...
## $ Pain_now.Wk8
                      <int> NA, NA, O, NA, NA, NA, 3, 5, 10, 1, NA, NA, ...
## $ Pain_now.Wk12
                      <int> NA, NA, 2, NA, NA, NA, 8, 5, 10, 3, 8, NA, N...
## $ Pain now.Wk24
                      <int> NA, NA, 2, NA, NA, NA, 3, 8, NA, 3, NA, NA, ...
                      <int> NA, NA, 4, NA, NA, NA, 3, 2, NA, 5, NA, NA, ...
## $ Pain_now.Wk48
## $ dosage
```

BOCF data processing and analysis

```
# Calculate Pain Severity Score (PSS) for weeks 0 (BL) and 48 (Wk48)
BOCF <- primary %>%
    gather(key = 'Time',
           value = 'NRS',
           - ID, -Study_site, -Group) %>%
    separate(col = Time,
             into = c('Pain', 'Period'),
             sep = '\\.') %>%
    group_by(ID, Study_site, Group, Period) %>%
    nest() %>%
    mutate(PPS = map(.x = data,
                     ~ summarise(.x, Mean = mean(NRS, na.rm = TRUE)))) %>%
    select(-data) %>%
    unnest() %>%
    filter(Period %in% c('BL', 'Wk48')) %>%
    mutate(Mean = ifelse(is.nan(Mean),
                         yes = NA,
                          no = Mean))
# Missing baseline data
BOCF[is.na(BOCF$Mean) & BOCF$Period == 'BL', ]
## # A tibble: 16 x 5
            Study_site Group Period Mean
##
      ID
##
      <chr> <chr>
                       <chr> <chr> <dbl>
## 1 J18
            U1
                       Т
                             BL
                                        NA
## 2 J29
                       Т
            U1
                             BL
                                        NA
## 3 J59
                       Т
                             BL
            U1
                                        NA
## 4 J67
                       Т
            U1
                             BL
                                        NA
## 5 M1
                       Т
            R1
                             BL
                                        NA
## 6 M8
                       Т
                             BL
            R1
                                        NA
## 7 M9
            R1
                       Т
                             BL
                                        NA
## 8 M12
                       Т
                             BL
            R1
                                        NA
## 9 M20
                       Ρ
                             BL
                                        NA
            R1
                       Ρ
## 10 M21
            R1
                             BL
                                        NA
## 11 M23
            R1
                       Ρ
                             BL
                                        NA
## 12 M25
            R1
                       Ρ
                             BL
                                        NA
            R1
## 13 M29
                       Ρ
                             BL
                                        NA
## 14 M38
            R1
                       Ρ
                             BL
                                        NA
## 15 M41
                       Ρ
                             BL
                                        NA
            R.1
## 16 M45
            R1
                       Т
                             BL
                                        NA
# Number of participants with missing baselines
nrow(BOCF[is.na(BOCF$Mean) & BOCF$Period == 'BL', ])
## [1] 16
# Calculate baseline mean for study sites R1 and U1 (sites with missing data)
R1 <- mean(BOCF$Mean[!is.na(BOCF$Mean) &
```

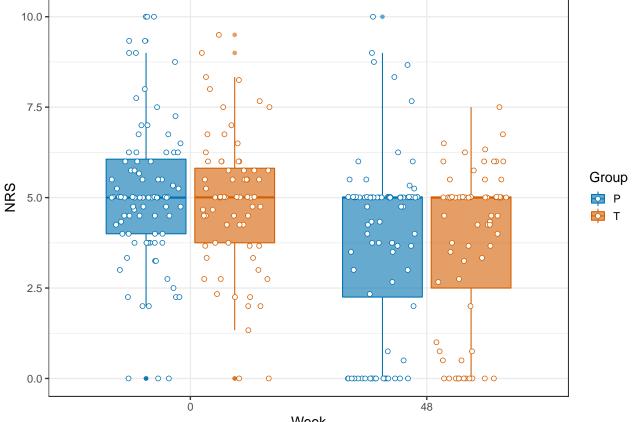
BOCF\$Study_site == 'R1' &

```
BOCF$Period == 'BL'])
U1 <- mean(BOCF$Mean[!is.na(BOCF$Mean) &
                         BOCF$Study_site == 'U1' &
                         BOCF$Period == 'BL'])
# Substitute missing baselines with Study site means
BOCF %<>%
    mutate(Mean = ifelse(is.na(Mean) & Study_site == 'R1',
                         yes = R1,
                         no = Mean),
           Mean = ifelse(is.na(Mean) & Study_site == 'U1',
                         yes = U1,
                         no = Mean))
# Spread data
BOCF_wide <- BOCF %>%
    spread(key = Period,
           value = Mean)
# Number of participants
BOCF_wide %>%
    summarise(Count = n())
## # A tibble: 1 x 1
   Count
##
##
     <int>
## 1
     160
# Number with missing data at week 48
BOCF_wide %>%
    filter(is.na(Wk48)) %>%
    summarise(Count = n())
## # A tibble: 1 x 1
   Count
##
##
     <int>
## 1
       35
# Number with missing data at week 48 (by intervention)
BOCF_wide %>%
    filter(is.na(Wk48)) %>%
    group by(Group) %>%
    summarise(Count = n())
## # A tibble: 2 x 2
##
   Group Count
   <chr> <int>
##
## 1 P
              20
## 2 T
              15
# Create BOCF dataframe
BOCF_wide %<>%
    mutate(Wk48 = ifelse(is.na(Wk48),
                         yes = BL,
                         no = Wk48))
```

```
# Plot of BOCF NRS data at weeks 0 and 48 (by intervention)
BOCF_wide %>%
   gather(key = Period,
           value = NRS,
           BL, Wk48) %>%
   ggplot(data = .) +
    aes(x = Period),
        y = NRS,
        colour = Group,
        fill = Group) +
   geom_boxplot(alpha = 0.6) +
   geom_point(position = position_jitterdodge(jitter.height = 0,
                                                jitter.width = 0.3),
               shape = 21,
               fill = '#FFFFFF',
               size = 2) +
   labs(title = 'BOCF: NRS at week 0 and week 48',
         subtitle = 'P: Positive-living programme and therapeutic relationship\nT: Therapeutic relation
         x = 'Week') +
    scale fill manual(values = pal) +
    scale_colour_manual(values = pal) +
    scale_x_discrete(labels = c(0, 48))
```

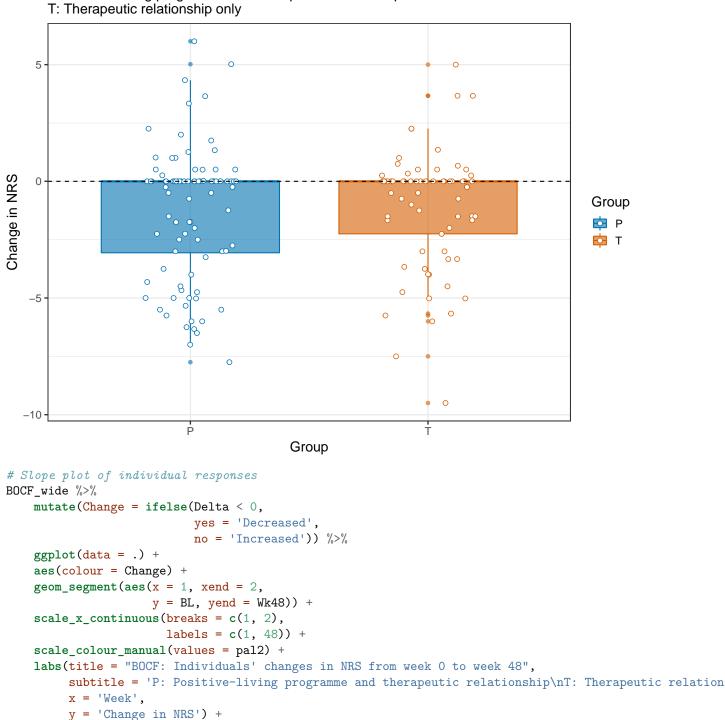
BOCF: NRS at week 0 and week 48

P: Positive-living programme and therapeutic relationship T: Therapeutic relationship only



Week

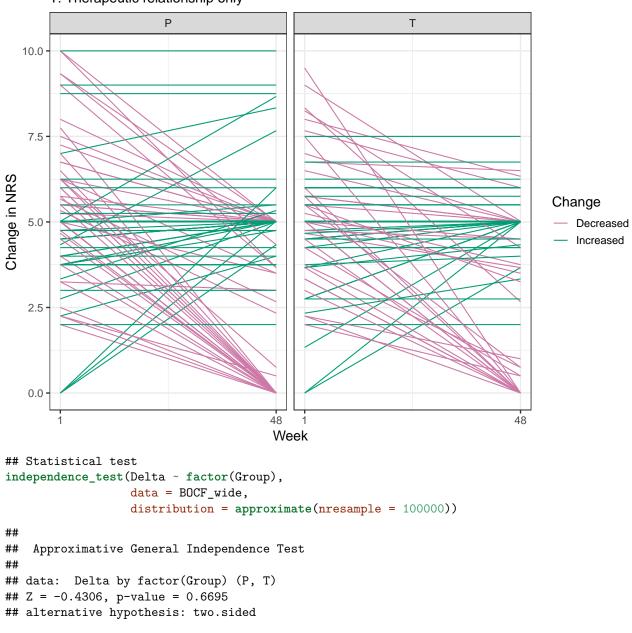
```
# Calculate change from baseline to Wk48
BOCF_wide %<>%
    mutate(Delta = Wk48 - BL)
# Compare the P and T groups
## Plot of BOCF change in NRS between week 0 and 48 (by intervention)
ggplot(data = BOCF_wide) +
    aes(x = Group,
       y = Delta,
       fill = Group,
       colour = Group) +
    geom_boxplot(alpha = 0.6) +
    geom_point(position = position_jitterdodge(jitter.height = 0,
                                               jitter.width = 0.4),
               shape = 21,
               size = 2,
               fill = '#FFFFFF') +
    geom_hline(yintercept = 0,
               linetype = 2) +
    labs(title = 'BOCF: Change in NRS from week 0 to week 48',
         subtitle = 'P: Positive-living programme and therapeutic relationship\nT: Therapeutic relation
         y = 'Change in NRS') +
    scale_fill_manual(values = pal) +
    scale_colour_manual(values = pal)
```



BOCF: Change in NRS from week 0 to week 48 P: Positive–living programme and therapeutic relationship

7

facet_wrap(~ Group)

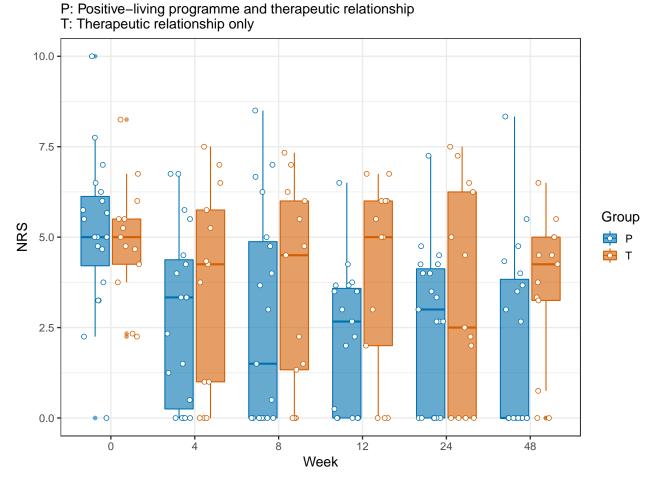


BOCF: Individuals' changes in NRS from week 0 to week 48 P: Positive–living programme and therapeutic relationship T: Therapeutic relationship only

Per protcol data processing and analysis

```
into = c('Pain', 'Period'),
             sep = '\\.') %>%
    group_by(ID, Study_site, Group, Period) %>%
    nest() %>%
    mutate(PPS = map(.x = data,
                    ~ summarise(.x, Mean = mean(NRS, na.rm = TRUE)))) %>%
    select(-data) %>%
    unnest() %>%
    mutate(Mean = ifelse(is.nan(Mean),
                         yes = NA,
                         no = Mean)) %>%
    mutate(Period = case_when(
        Period == 'BL' ~ 'TOO',
        Period == 'Wk4' ~ 'T04',
        Period == 'Wk8' ~ 'T08',
        Period == 'Wk12' ~ 'T12',
        Period == 'Wk24' ~ 'T24',
        Period == 'Wk48' ~ 'T48'
    )) %>%
    spread(key = Period,
           value = Mean) %>%
    select(- `<NA>`)
# PP cohort
PP_wide %<>%
    filter(complete.cases(.))
# Number of participants with complete data
PP_wide %>%
    summarise(Count = n())
## # A tibble: 1 x 1
##
   Count
##
     <int>
## 1
        32
# Number of participants with complete data (by intervention)
PP wide %>%
    group_by(Group) %>%
    summarise(Count = n())
## # A tibble: 2 x 2
##
   Group Count
     <chr> <int>
##
## 1 P
              19
## 2 T
              13
# Plot of BOCF NRS data at weeks 0 and 48 (by intervention)
PP_wide %>%
    gather(key = Period,
           value = NRS,
           -ID, -Study_site, -Group) %>%
    ggplot(data = .) +
    aes(x = Period),
        y = NRS,
        colour = Group,
```

PP: NRS at weeks 0 through to 48

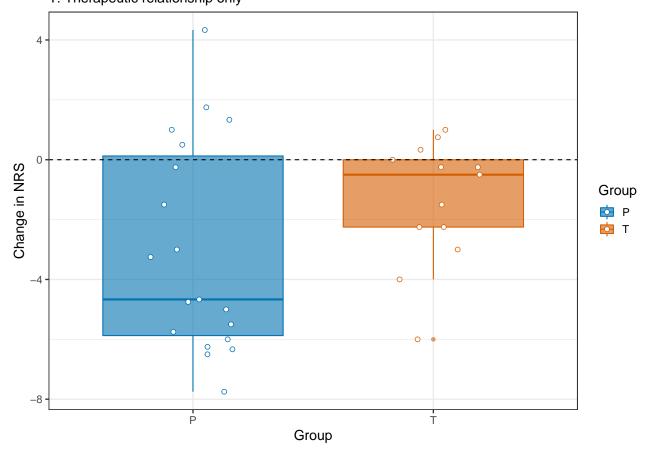


Calculate change from baseline to Wk48
PP_wide %<>%
 mutate(Delta = T48 - T00)

```
# Compare the P and T groups
## Plot of BOCF change in NRS between week 0 and 48 (by intervention)
ggplot(data = PP_wide) +
    aes(x = Group,
        y = Delta,
```

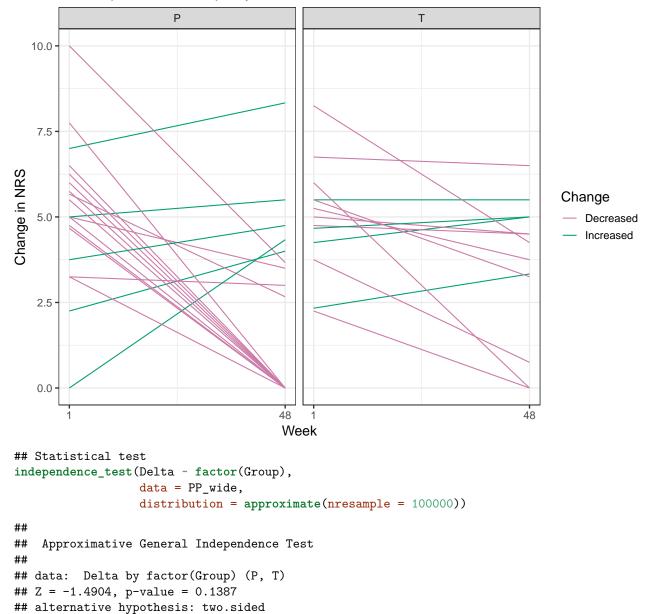
PP: Change in NRS from week 0 to week 48

P: Positive–living programme and therapeutic relationship T: Therapeutic relationship only



PP: Individuals' changes in NRS from week 0 to week 48

P: Positive–living programme and therapeutic relationship



T: Therapeutic relationship only

Treatment dose

Number of assessments attended on average

```
dose <- primary %>%
   gather(key = 'Time',
          value = 'NRS',
          - ID, -Study_site, -Group) %>%
    separate(col = Time,
            into = c('Pain', 'Period'),
            sep = '\\.') %>%
   group_by(ID, Study_site, Group, Period) %>%
   nest() %>%
   mutate(PPS = map(.x = data,
                    ~ summarise(.x, Mean = mean(NRS, na.rm = TRUE)))) %>%
   select(-data) %>%
   unnest() %>%
   mutate(Mean = ifelse(is.nan(Mean),
                        yes = NA,
                        no = Mean)) %>%
   mutate(Period = case_when(
       Period == 'BL' ~ 'TOO',
       Period == 'Wk4' ~ 'T04',
       Period == 'Wk8' ~ 'T08',
       Period == 'Wk12' ~ 'T12',
       Period == 'Wk24' ~ 'T24',
       Period == 'Wk48' ~ 'T48'
   ))
# Calculate dose per individual
dose_b <- dose %>%
   mutate(Mean = ifelse(is.na(Mean),
                        yes = 0,
                        no = 1)) \% > \%
   group_by(Group, Study_site, ID) %>%
   summarise(Count = sum(Mean))
# Calculate summary stat for dose
dose_b %>%
   ungroup() %>%
   summarise(Median = median(Count),
             q25 = quantile(Count, probs = 0.25),
             a75 = quantile(Count, probs = 0.75),
             min = min(Count),
             \max = \max(Count))
## # A tibble: 1 x 5
##
    Median q25 a75 min max
     ##
## 1
         5
               3
                    6
                          0
                                7
# Calculate summary stat for dose (by intervention)
dose_b %>%
   group_by(Group) %>%
```

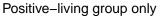
```
summarise(Median = median(Count),
           q25 = quantile(Count, probs = 0.25),
           a75 = quantile(Count, probs = 0.75),
           min = min(Count),
           \max = \max(Count))
## # A tibble: 2 x 6
## Group Median q25 a75 min
                               max
  ##
## 1 P
             5
                  3
                       6
                          1
                                 7
                  2
## 2 T
             4
                       5
                            0
                                 6
```

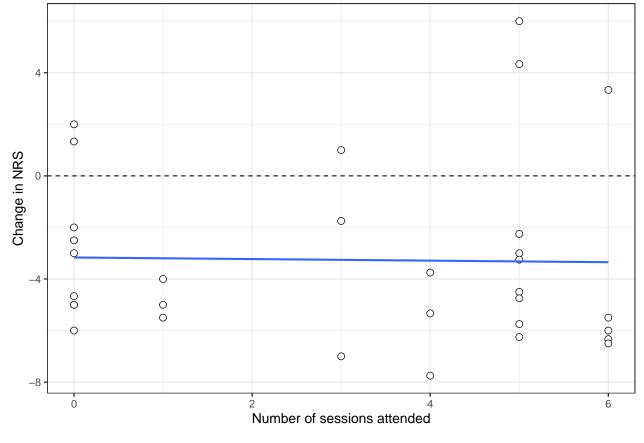
Relationship between treatment session attendance (Group P) and outcome

```
Excludes site U1 (data missing).
```

```
# Spread dose data
dose %<>%
    spread(key = Period,
           value = Mean)
# Calculate change from baseline to Wk48, PL group only
dose %<>%
   mutate(Delta = T48 - T00) %>%
   left_join(dosage) %>%
   filter(Group == 'P') %>%
   filter(Delta != 'NA') %>%
   filter(dosage != 'NA') # filters out 5 participants all from J site
# Plot of P group's change in NRS between week 0 and 48
# (by PL session attendance)
ggplot(data = dose) +
   aes(x = dosage)
       y = Delta) +
   geom_point(shape = 21,
               fill = '#FFFFFF',
               size = 3) +
   geom_hline(yintercept = 0,
               linetype = 2) +
   geom_smooth(method = 'lm',
                se = FALSE) +
   labs(title = 'Change in NRS from week 0 to week 48\nby number of sessions attended',
         subtitle = 'Positive-living group only',
         y = 'Change in NRS',
         x = 'Number of sessions attended')
```

Change in NRS from week 0 to week 48 by number of sessions attended





Session information

```
## R version 3.6.0 (2019-04-26)
## Platform: x86_64-apple-darwin15.6.0 (64-bit)
## Running under: macOS Mojave 10.14.4
##
## Matrix products: default
## BLAS:
           /Library/Frameworks/R.framework/Versions/3.6/Resources/lib/libRblas.0.dylib
## LAPACK: /Library/Frameworks/R.framework/Versions/3.6/Resources/lib/libRlapack.dylib
##
## locale:
## [1] en_US.UTF-8/en_US.UTF-8/en_US.UTF-8/C/en_US.UTF-8/en_US.UTF-8
##
## attached base packages:
## [1] stats
                graphics grDevices utils
                                               datasets methods
                                                                   base
##
## other attached packages:
  [1] coin_1.3-0
                          survival_2.44-1.1 forcats_0.4.0
##
##
  [4] stringr_1.4.0
                          dplyr_0.8.0.1
                                            purrr_0.3.2
  [7] readr_1.3.1
                          tidyr_0.8.3
                                            tibble_2.1.1
##
```

## ##	[10]	ggplot2_3.1.1	tidyverse_1.2.1 m	agrittr_1.5
##	loade	ed via a namespace	(and not attached):	
##	[1]	Rcpp_1.0.1	lubridate_1.7.4	mvtnorm_1.0-10
##	[4]	lattice_0.20-38	zoo_1.8-5	$assertthat_0.2.1$
##	[7]	digest_0.6.18	utf8_1.1.4	R6_2.4.0
##	[10]	cellranger_1.1.0	plyr_1.8.4	backports_1.1.4
##	[13]	stats4_3.6.0	evaluate_0.13	httr_1.4.0
##		pillar_1.3.1	rlang_0.3.4	<pre>lazyeval_0.2.2</pre>
##	[19]	multcomp_1.4-10	readxl_1.3.1	rstudioapi_0.10
##	[22]	Matrix_1.2-17	rmarkdown_1.12	labeling_0.3
##	[25]	splines_3.6.0	munsell_0.5.0	broom_0.5.2
##	[28]	compiler_3.6.0	modelr_0.1.4	xfun_0.6
##	[31]	pkgconfig_2.0.2	libcoin_1.0-4	htmltools_0.3.6
##	[34]	tidyselect_0.2.5	codetools_0.2-16	matrixStats_0.54.0
##	[37]	fansi_0.4.0	crayon_1.3.4	withr_2.1.2.9000
##	[40]	MASS_7.3-51.4	grid_3.6.0	nlme_3.1-139
##	[43]	jsonlite_1.6	gtable_0.3.0	scales_1.0.0
##	[46]	cli_1.1.0	stringi_1.4.3	xml2_1.2.0
##	[49]	generics_0.0.2	<pre>sandwich_2.5-1</pre>	TH.data_1.0-10
##	[52]	tools_3.6.0	glue_1.3.1	hms_0.4.2
##	[55]	parallel_3.6.0	yaml_2.2.0	colorspace_1.4-1
##	[58]	rvest_0.3.3	knitr_1.22	haven_2.1.0
##	[61]	modeltools_0.2-22		