```
title: "Remittance Analysis"
author: "Taj Cole"
date: '2022-06-16'
output:
  html_document: default
  pdf document: default
 word_document: default
```{r}
# Packages
library(dplyr)
library(ggplot2)
library(plotly)
# load data
load("~/2022-DSPG-LivDiv-/data/livdivdata.RData")
# financial
fd <- livdiv %>%
  select(-(4:967))
# subset
fd <- livdiv %>%
  select(-(4:967))
# remittance data
rmt <- fd %>%
  select(village, hhid, name, week, date,
         rmt_total, rmt_method_bank, rmt_method_person,
         rmt method mobile, rmt method moneyorder, rmt method oth,
rmt_purpose_food, rmt_purpose_food,
         rmt_purpose_tuition, rmt_purpose_asset, rmt_purpose_med,
rmt_purpose_oth, rmt_purpose_none)
rmt_nonZero <- rmt %>%
  filter(rmt_total != 0)
rmt_nonZero
- Remittance income is money sent to a person for the payment of goods and
services, or as a gift
# Method of recieving remittance
```{r}
rmt_method_summary <- rmt_nonZero %>%
  summarise("bank" = sum(rmt_method_bank), "in person" =
sum(rmt_method_person),
            "mobile" = sum(rmt_method_mobile), "money order" =
sum(rmt_method_moneyorder),
            "other" = sum(rmt_method_oth))
rmt_method_summary
# Histogram of methods to recieve remittance
```

```
```{r}
method_counts <- c(397, 472, 1, 1, 13)
Method <- c("Bank", "In person", "Mobile", "Money Order", "Other")</pre>
method_dat <- data.frame(Method, method_counts, stringsAsFactors = T)</pre>
rmt_method_plot <- ggplot(method_dat, aes( x= Method, y = method_counts, fill</pre>
= Method)) +
  geom_col() +
  labs(x = "", y = "Count") +
  theme_classic() +
  ggtitle("Method of Recieving Remittance")
rmt_method_plot
- Shows the count of how every transaction of remittance was received over
the data period
- Remittance income was primarily received through a bank or in person

    Many households may not have mobile phones to receive transactions through

- Money orders are often very secure ways of sending money. Many households
are more likely concerned about receiving their money quickly rather than
securely
- (It wouldn't make sense to do this by village because some villages dint
receive weekly remittance for a majority of the data set so there would be
many gaps and wouldn't show some of the villages)
# Purpose of recieving remittance
```{r}
rmt purpose summary <- rmt nonZero %>%
  summarise("Food/utility purchases" = sum(rmt_purpose_food), "tuition" =
sum(rmt_purpose_tuition),
             "asset/durable purchases" = sum(rmt_purpose_asset), "medical
expenses" =sum(rmt_purpose_med),
            "other reason" = sum(rmt_purpose_oth), "No reason" =
sum(rmt_purpose_none))
rmt_purpose_summary
# Histogram of purpose of recieving remittance
```{r}
Purpose <- c("Food/Utility Purchases", "Tuition", "Assets/Durable
Purchases", "Medical Expenses", "Other", "No Reason")</pre>
purpose_count <- c(594, 37, 27, 93, 128, 43)
purpose_dat <- data.frame(Purpose, purpose_count, stringsAsFactors = T)</pre>
rmt_purpose_plot <- ggplot(purpose_dat, aes(x = Purpose, y = purpose_count,</pre>
fill = Purpose)) + geom_col() +
  labs(x = "", y = "Count") +
  theme_classic() +
  ggtitle("Purpose for Recieving Remittance")+
  #rotate_x_text(angle = 22, size = rel(0.8))
  coord flip()
rmt_purpose_plot
```

- (Across all villages) Shows the count of what every transaction of remittance was used for over the data period - Remittance income was primarily being used for food and utility purchases It makes sense that a large majority of remittance is used on food, especially if the remittance is being received after devasttaion (cyclone) - (assets/durable purchases: appliances, furniture, goods that don't wear out quickly and are not purchased frequently) # Get average and total income by village for each week then compare across 49 week period # Villages vector ```{r} village <- c("Amrabati", "Beguakhali", "Bijoynagar", "Birajnagar", "Haridaskati</pre> Samsernagar", "Lakshmi Janardanpur", "Pargumti", "Purba Dwarokapur", "Sagar", "Shibpur") Villages <- rep(village, 49)</pre> # vector of weeks repeated weeks\_rep <- c(rep(1,10), rep(2, 10), rep(3,10), rep(4,10), rep(5,10),rep(6,10), rep(7,10), rep(8,10), rep(9,10), rep(10,10), rep(11,10), rep(12,10), rep(13, 10), rep(14,10), rep(15,10), rep(16, 10), rep(17, 10), rep(18,10), rep(19,10), rep(20,10), rep(21, 10), rep(22, 10), rep(23, 10), rep(24, 10), rep(25, 10),rep(26, 10), rep(27,10), rep(28,10), rep(29,10), rep(30,10),rep(31,10), rep(32,10), rep(33,10), rep(34,10), rep(35,10), rep(36,10), rep(37,10), rep(38,10), rep(39,10), rep(40,10), rep(41,10), rep(42,10), rep(43,10), rep(44,10), rep(45,10), rep(46,10), rep(47,10), rep(48,10), rep(49,10)## Average remittance each week (vectors) ```{r, include=FALSE}  $\text{week1\_rmt\_means} \leftarrow c(0, 5166.667, 11100, 0, 5250, 950, 5000, 0, 0, 0)$  $week2\_rmt\_means <- c(2000, 6000, 4071.429, 1600, 1300, 2700, 0, 2383.333,$ 3000, 2625) week3\_rmt\_means <- c(2000, 3000, 2033.333, 2500, 3250, 3000, 2000, 1250, 0,

```
week11_rmt_means <-c(0, 1000, 1542.857, 2500, 1166.667, 750, 0, 1400, 2000,
900)
\text{week12\_rmt\_means} \leftarrow c(0, 3500, 3333.333, 1800, 7000, 0, 0, 1600, 0, 600)
\text{week13\_rmt\_means} \leftarrow c(0, 9333.333, 3416.667, 1600, 6333.333, 500, 0, 1625, 0,
week14\_rmt\_means <- c(0, 2666.667, 3959.091, 2600, 0, 0, 2000, 1250, 5000,
week15_rmt_means <- c(0, 3000, 2500, 2700, 500, 1500, 0, 925, 2750, 8000)
week16_rmt_means <- c(0, 4125, 5000, 6800, 500, 0, 3200, 800, 500, 8750)
week17_rmt_means <- c(0, 3500, 4000, 1700, 500, 3000, 0, 1850, 0, 10000)
week18_rmt_means <- c(0, 30200, 3250, 6750, 20250, 2000, 3000, 4300, 0, 5750)
\text{week19\_rmt\_means} \leftarrow c(0, 6000, 3250, 3000, 700, 3192.5, 500, 2600, 0, 13000)
week20_rmt_means <- c(0, 1000, 2500, 0, 500, 3590, 0, 200, 0, 1500)
\text{week21\_rmt\_means} \leftarrow c(0, 5333.333, 3625, 3000, 300, 3500, 1000, 1730, 0, 0)
week22_rmt_means <- c(8000, 4501.250, 4250, 2633.333, 25000, 900, 1250,
3166.667, 0, 8000)
week23_rmt_means <- c(0, 2333.333, 4250, 3850, 0, 3166.667, 0, 2333.333, 0,
\text{week24\_rmt\_means} \leftarrow c(0, 6333.333, 2285.714, 6500, 0, 7500, 1050, 1300, 0,
6500)
week25_rmt_means <- c(0, 6416.667, 3333.333, 500, 0, 5000, 0, 10000, 0, 0)
week26_rmt_means <- c(0, 6500, 2800, 2100, 0, 3000, 8000, 500, 0, 30000)
week27_rmt_means <- c(0, 3000, 5460, 4000, 20000, 0, 0, 1400, 0, 0)
week28_rmt_means <- c(0, 0, 2300, 0, 0, 2750, 0, 3000, 2000, 7500)
week29_rmt_means <- c(0, 5000, 5062.5, 15000, 10000, 2562.5, 0, 2875, 5000,
week30_rmt_means <- c(0, 5375, 2928.571, 3500, 3675, 0, 0, 1033.333, 0, 3025)
week31_rmt_means <- c(0, 1500, 3366.667, 0, 500, 5666.667, 0, 750, 7000, 0)
week32_rmt_means <-c(0, 6000, 1833.333, 6500, 300, 2000, 0, 900, 0, 9900)
week33_rmt_means <-c(0, 6000, 2742.857, 0, 450, 7000, 0, 2500, 0, 5000)
week34_rmt_means <- c(0, 2000, 2214.286, 3000, 1450, 0, 0, 500, 2000, 0)
week35_rmt_means <- c(0, 3500, 4136.364, 4433.333, 2733.333, 3400, 0, 0, 0,
week36\_rmt\_means <- c(0, 1000, 3000, 2000, 475, 0, 0, 1000, 0, 10000)
week37_rmt_means <-c(0, 2500, 4927.273, 5000, 2637.5, 0, 0, 500, 0,
7666.667)
week38_rmt_means <- c(0, 7400, 2657.143, 4035, 3103.333, 1500, 0, 1500, 0, 0)
week39_rmt_means <- c(0, 1000, 3500, 3000, 1875, 0, 0, 3250, 0, 0)
week40\_rmt\_means <- c(500, 3533.333, 2600, 5035, 1943.333, 3100, 2500, 500,
week41_rmt_means <- c(0, 4500, 3242.857, 0, 1220, 1000, 0, 1500, 0, 5000)
week42_rmt_means <- c(5000, 6250, 2583.333, 2500, 1000, 2500, 5000, 2750, 0,
week43_rmt_means <- c(5000, 2500, 1750, 1945, 5000, 3000, 0, 3333.333, 0,
4000)
week44_rmt_means <- c(0, 3250, 2300, 800, 400, 0, 0, 1750, 0, 20500)
week45_rmt_means <- c(0, 5333.333, 3200, 7000, 2153.333, 0, 0, 2900, 0, 0)
week46\_rmt\_means <- c(0, 7333.333, 5400, 3335, 1250, 2500, 0, 1866.667, 1800,
15000)
week47_rmt_means <- c(0, 4000, 3582.500, 3833.333, 5000, 0, 0, 1100, 0, 0)
week48_rmt_means <- c(0, 3900, 6875, 0, 1210, 8500, 0, 1200, 0, 0)
week49_rmt_means < c(9700, 6500, 2228.571, 0, 500, 8000, 0, 1675, 0, 400)
```

```
# vector for all weeks of remittance averages
 ``{r}
mean_rmt_per_week <- c(week1_rmt_means, week2_rmt_means, week3_rmt_means,</pre>
week4_rmt_means, week5_rmt_means,
                         week6 rmt means, week7 rmt means, week8 rmt means,
week9_rmt_means, week10_rmt_means,
                         week11_rmt_means, week12_rmt_means,
week13_rmt_means, week14_rmt_means, week15_rmt_means,
                         week16_rmt_means, week17_rmt_means,
week18_rmt_means, week19_rmt_means, week20_rmt_means,
                         week21_rmt_means, week22_rmt_means,
week23_rmt_means, week24_rmt_means, week25_rmt_means,
                         week26_rmt_means, week27_rmt_means,
week28_rmt_means, week29_rmt_means, week30_rmt_means,
                         week31_rmt_means, week32_rmt_means,
week33 rmt means, week34 rmt means, week35 rmt means,
                         week36_rmt_means, week37_rmt_means,
week38_rmt_means, week39_rmt_means, week40_rmt_means,
                         week41_rmt_means, week42_rmt_means,
week43_rmt_means, week44_rmt_means, week45_rmt_means,
                         week46_rmt_means, week47_rmt_means,
week48_rmt_means, week49_rmt_means)
# Data frame for average remittance
```{r}
rmt_data_mean_weeks <- data.frame(Villages, weeks_rep, mean_rmt_per_week)</pre>
# Average remittance income plot
```{r}
average_rmt_plot <- ggplot(rmt_data_mean_weeks, aes(x = weeks_rep, y =
mean_rmt_per_week, color = Villages)) +
  geom_line() +
  theme classic() +
  labs(x = "Weeks", y = "Average Remittance Income [Rupee]") +
  ggtitle("Average Remittance Income Per Week Per Village (11/16/18 -
10/31/19)") +
  scale_color_brewer(palette = "Spectral")
average_rmt_plot

    Large spikes in remittance start around week 19 happen throughout the rest

of the data period
- Week 19-49 covered the dates march 22, 2019 - October 31, 2019

    Within this time period The Sundarbans was affected by 3 severe cyclones

that hit the Bengal Bay: Fani (category 4), Bulbul and Matmo (category 1)
- Fani was from April 25th to May 4th 2019 (weeks 22-24)
- Bulbul and Matmo were from October 28th to November 11th (week 48 and 49)
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- Matmo formed in the Philippine Sea on October 24th, dissipated as it went West over land (Cambodia), then regained energy and reached peak strength as it went over the Andaman Sea and into the Bengal bay, where it was renamed Bulbul
- citation: [Al Jazeera. "Tropical Cyclone Matmo Is Reborn As Bulbul."
  Climate Crisis News | Al Jazeera, Al Jazeera Media Network, 7 Nov. 2019,
  https://www.aljazeera.com/news/2019/11/7/tropical-cyclone-matmo-is-reborn-as-bulbul.]
- Villages could've also been impacted by two cylones that hit the Arabian sea during this time period: Vayu and Hikaa (category 1)
  - Vayu was from June 8-18th (week 29-31)
  - Hikaa was from September 20-26th (week 44)
    - Hikaa did not make direct landfall
- -While not reported as a region directly affected by these two cyclones, it is very likely that the Sundarbans still experienced the negative residuals of

these storms due to their proximity to the Arabian Sea

- -Villages with the largest spikes were most likely hit the hardest
- Cyclone dates citation : ["Most Recent Cyclones in India." Worlddata.info, Worlddata, https://www.worlddata.info/asia/india/cyclones.php.]

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## # Appendix

```
# Total remittance for each village each week (vectors)
week1_rmt_totals<- c(0, 15500, 22200, 0, 10500, 1900, 5000, 0, 0, 0)
week2_rmt_totals <- c(2000, 24000, 28500, 3200, 2600, 5400, 0, 14300, 3000,</pre>
10500)
week3_rmt_totals <- c(4000, 12000, 12200, 5000, 6500, 6000, 4000, 2500, 0,</pre>
week4 rmt totals \leftarrow c(0, 19000, 13500, 3300, 7000, 5000, 1500, 8500, 4500,
10000)
week5\_rmt\_totals <- c(0, 10000, 15800, 5700, 4000, 0, 0, 1750, 3000, 0)
week6_rmt_totals <- c(0, 31400, 22400, 5200, 0, 0, 1500, 700, 7000, 1000)
week7_rmt_totals <- c(0, 2000, 15500, 7100, 3000, 900, 5000, 8500, 0, 12600)</pre>
week8_rmt_totals <- c(0, 18500, 16500, 5800, 0, 0, 0, 16100, 0, 500)</pre>
week9_rmt_totals <- c(0, 7000, 29500, 11100, 0, 1000, 2000, 7300, 0, 16500)
week10_rmt_totals <- c(0, 20000, 28000, 18400, 0, 11500, 0, 6300, 0, 7000)
week11_rmt_totals <- c(0, 3000, 10800, 5000, 3500, 1500, 0, 4200, 2000, 900)
week12_rmt_totals <- c(0, 14000, 10000, 5400, 7000, 0, 0, 3200, 0, 600)
week13_rmt_totals <- c(0, 28000, 20500, 3200, 19000, 500, 0, 6500, 0, 3500)</pre>
week14_rmt_totals <- c(0, 8000, 43550, 13000, 0, 0, 2000, 2500, 5000, 1700)</pre>
week15_rmt_totals <- c(0, 6000, 7500, 8100, 500, 3000, 0, 3700, 5500, 24000)
week16_rmt_totals <- c(0, 16500, 35000, 20400, 500, 0, 3200, 2400, 500,
17500)
week17_rmt_totals <- c(0, 7000, 24000, 3400, 500, 3000, 0, 3700, 0, 10000)
week18 rmt totals <- c(0, 151000, 13000, 13500, 40500, 4000, 3000, 34400, 0,
week19_rmt_totals <- c(0, 12000, 19500, 9000, 1400, 6385, 500, 5200, 0,
13000)
```

```
week20_rmt_totals <- c(0, 1000, 10000, 0, 500, 7180, 0, 200, 0, 3000)
week21_rmt_totals <- c(0,16000, 14500, 3000, 300, 3500, 1000, 8650, 0, 0)
week22_rmt_totals <- c(8000, 18005, 17000, 7900, 25000, 900, 2500, 9500, 0,
8000)
week23_rmt_totals <- c(0, 7000, 17000, 7700, 0, 9500, 0, 7000, 0, 0)
week24_rmt_totals <- c(0, 19000, 16000, 13000, 0, 15000, 1050, 2600, 0,
13000)
week25_rmt_totals <- c(0, 38500, 10000, 500, 0, 5000, 0, 10000, 0, 0)
week26_rmt_totals <- c(0, 13000, 14000, 2100, 0, 6000, 8000, 500, 0, 30000)
week27_rmt_totals <- c(0, 3000, 27300, 4000, 20000, 0, 0, 2800, 0, 0)
week28_rmt_totals <- c(0, 0, 11500, 0, 0, 5500, 0, 9000, 2000, 15000)
week29_rmt_totals <- c(0, 10000, 40500, 15000, 15000, 5125, 0, 11500, 5000,
3500)
week30_rmt_totals <- c(0, 21500, 20500, 3500, 14700, 0, 0, 3100, 0, 6050)
week31_rmt_totals <- c(0, 1500, 20200, 0, 500, 17000, 0, 1500, 7000, 0)
week32_rmt_totals <- c(0, 30000, 5500, 13000, 300, 2000, 0, 3600, 0, 19800)
week33_rmt_totals <- c(0, 6000, 19200, 0, 450, 14000, 0, 7500, 0, 5000)
week34_rmt_totals <- c(0, 2000, 15500, 3000, 4350, 0, 0, 500, 2000, 0)
week35_rmt_totals <- c(0, 7000, 45500, 13300, 8200, 3400, 0, 0, 0, 32000)
week36_rmt_totals <- c(0, 1000, 27000, 2000, 950, 0, 0, 1000, 0, 10000)
week37_rmt_totals <- c(0, 5000, 54200, 5000, 10550, 0, 0, 500, 0, 23000)
week38_rmt_totals <- c(0, 37000, 18600, 8070, 9310, 1500, 0, 3000, 0, 0)
week39_rmt_totals <- c(0, 2000, 28000, 3000, 7500, 0, 0, 6500, 0, 0)</pre>
week40_rmt_totals <- c(500, 10600, 15600, 10070, 5830, 6200, 2500, 500, 0,
900)
week41_rmt_totals <- c(0, 9000, 22700, 0, 3660, 1000, 0, 3000, 0, 5000)
week42_rmt_totals <- c(5000, 25000, 15500, 2500, 1000, 7500, 5000, 5500, 0,
30000)
week43 rmt totals <- c(5000, 7500, 7000, 5835, 5000, 3000, 0, 10000, 0, 4000)
week44_rmt_totals <- c(0, 6500, 11500, 800, 400, 0, 0, 3500, 0, 41000)
week45_rmt_totals <- c(0, 22000, 27000, 3335, 2500, 5000, 0, 5600, 1800,
15000)
week46_rmt_totals <- c(0, 22000, 27000, 3335, 2500, 5000, 0, 5600, 1800,
15000)
week47_rmt_totals <- c(0, 4000, 28660, 11500, 5000, 0, 0, 3300, 0, 0)
week48_rmt_totals <- c(0, 19500, 27500, 0, 2420, 17000, 0, 2400, 0, 0)
week49_rmt_totals <- c(9700, 13000, 15600, 0, 500, 8000, 0, 3350, 0, 400)
. . .
# Vector of total remittance in each village per week
```{r}
total_rmt_per_week <- c(week1_rmt_totals, week2_rmt_totals, week3_rmt_totals,</pre>
week4_rmt_totals, week5_rmt_totals,
                        week6_rmt_totals, week7_rmt_totals, week8_rmt_totals,
week9_rmt_totals, week10_rmt_totals,
                        week11_rmt_totals, week12_rmt_totals,
week13_rmt_totals, week14_rmt_totals, week15_rmt_totals,
                        week16_rmt_totals, week17_rmt_totals,
week18 rmt totals, week19 rmt totals, week20 rmt totals,
                        week21_rmt_totals, week22_rmt_totals,
week23_rmt_totals, week24_rmt_totals, week25_rmt_totals,
```

```
week26_rmt_totals, week27_rmt_totals,
week28_rmt_totals, week29_rmt_totals, week30_rmt_totals,
                        week31_rmt_totals, week32_rmt_totals,
week33_rmt_totals, week34_rmt_totals, week35_rmt_totals,
                        week36_rmt_totals, week37_rmt_totals,
week38_rmt_totals, week39_rmt_totals, week40_rmt_totals,
                        week41 rmt totals, week42 rmt totals,
week43_rmt_totals, week44_rmt_totals, week45_rmt_totals,
                        week46_rmt_totals, week47_rmt_totals,
week48_rmt_totals, week49_rmt_totals)
# Data frame of Total remittance income for each village per week
rmt_data_total_weeks <- data.frame(Villages, weeks_rep, total_rmt_per_week)</pre>
. . .
# Plot of total rmeittance income for each village over 49 week period
```{r}
rmt_total_plot <- ggplot(rmt_data_total_weeks, aes(x = weeks_rep, y =</pre>
total_rmt_per_week, color = Villages))+
  geom_line() +
  theme_classic() +
  labs(x = "Weeks", y = "Total Village Remittance Income") +
  ggtitle("Total Village Remittance Income Over 49 Weeks")
rmt_total_plot

    There was a huge spike in remittance income in week 18 in the Beuakhali

village. This is mostly from one resident that recieved 132000 that week
# Tables for total and average remittance in each village each week
## Total remittance income in each village per week
````{r}
# 1
week 1 <- rmt nonZero %>%
  group_by(village) %>%
  filter(week == 1)
rmt_total_week1 <- week_1 %>%
  summarise("total remittance" = sum(rmt_total))
week_2 <- rmt_nonZero %>%
  group_by(village) %>%
  filter(week == 2)
rmt_total_week2 <- week_2 %>%
  summarise("total remittance" = sum(rmt_total))
# 3
week_3 <- rmt_nonZero %>%
  group by(village) %>%
  filter(week == 3)
```

```
rmt_total_week3 <- week_3 %>%
  summarise("total remittance" = sum(rmt_total))
week_4 <- rmt_nonZero %>%
  group_by(village) %>%
  filter(week == 4)
rmt total week4 <- week 4 %>%
  summarise("total remittance" = sum(rmt_total))
week_5 <- rmt_nonZero %>%
  group_by(village) %>%
  filter(week == 5)
rmt_total_week5 <- week_5 %>%
  summarise("total remittance" = sum(rmt_total))
week_6 <- rmt_nonZero %>%
  group_by(village) %>%
  filter(week == 6)
rmt_total_week6 <- week_6 %>%
  summarise("total remittance" = sum(rmt_total))
# 7
week_7 <- rmt_nonZero %>%
  group_by(village) %>%
  filter(week == 7)
rmt_total_week7 <- week_7 %>%
  summarise("total remittance" = sum(rmt_total))
week_8 <- rmt_nonZero %>%
  group by(village) %>%
  filter(week == 8)
rmt_total_week8 <- week_8 %>%
  summarise("total remittance" = sum(rmt_total))
week_9 <- rmt_nonZero %>%
  group_by(village) %>%
  filter(week == 9)
rmt_total_week9 <- week_9 %>%
  summarise("total remittance" = sum(rmt_total))
# 10
week 10 <- rmt nonZero %>%
  group_by(village) %>%
  filter(week == 10)
rmt_total_week10 <- week_10 %>%
  summarise("total remittance" = sum(rmt_total))
week_11 <- rmt_nonZero %>%
  group_by(village) %>%
  filter(week == 11)
rmt_total_week11 <- week_11 %>%
  summarise("total remittance" = sum(rmt_total))
# 12
week_12 <- rmt_nonZero %>%
  group_by(village) %>%
  filter(week == 12)
```

```
rmt_total_week12 <- week_12 %>%
  summarise("total remittance" = sum(rmt_total))
# 13
week_13 <- rmt_nonZero %>%
  group_by(village) %>%
  filter(week == 13)
rmt total week13 <- week 13 %>%
  summarise("total remittance" = sum(rmt_total))
# 14
week_14 <- rmt_nonZero %>%
  group_by(village) %>%
  filter(week == 14)
rmt_total_week14 <- week_14 %>%
  summarise("total remittance" = sum(rmt_total))
week_15 <- rmt_nonZero %>%
  group_by(village) %>%
  filter(week == 15)
rmt_total_week15 <- week_15 %>%
  summarise("total remittance" = sum(rmt_total))
# 16
week_16 <- rmt_nonZero %>%
  group_by(village) %>%
  filter(week == 16)
rmt_total_week16 <- week_16 %>%
  summarise("total remittance" = sum(rmt_total))
# 17
week_17 <- rmt_nonZero %>%
  group by(village) %>%
  filter(week == 17)
rmt_total_week17 <- week_17 %>%
  summarise("total remittance" = sum(rmt_total))
# 18
week_18 <- rmt_nonZero %>%
  group_by(village) %>%
  filter(week == 18)
rmt_total_week18 <- week_18 %>%
  summarise("total remittance" = sum(rmt_total))
# 19
week 19 <- rmt nonZero %>%
  group_by(village) %>%
  filter(week == 19)
rmt_total_week19 <- week_19 %>%
  summarise("total remittance" = sum(rmt_total))
week_20 <- rmt_nonZero %>%
  group_by(village) %>%
  filter(week == 20)
rmt_total_week20 <- week_20 %>%
  summarise("total remittance" = sum(rmt_total))
# 21
week_21 <- rmt_nonZero %>%
  group_by(village) %>%
  filter(week == 21)
```

```
rmt_total_week21 <- week_21 %>%
  summarise("total remittance" = sum(rmt_total))
# 22
week_22 <- rmt_nonZero %>%
  group_by(village) %>%
  filter(week == 22)
rmt total week22 <- week 22 %>%
  summarise("total remittance" = sum(rmt_total))
# 23
week_23 <- rmt_nonZero %>%
  group_by(village) %>%
  filter(week == 23)
rmt_total_week23 <- week_23 %>%
  summarise("total remittance" = sum(rmt_total))
week_24 <- rmt_nonZero %>%
  group_by(village) %>%
  filter(week == 24)
rmt_total_week24 <- week_24 %>%
  summarise("total remittance" = sum(rmt_total))
#25
week_25 <- rmt_nonZero %>%
  group_by(village) %>%
  filter(week == 25)
rmt_total_week25 <- week_25 %>%
  summarise("total remittance" = sum(rmt_total))
# 26
week_26 <- rmt_nonZero %>%
  group by(village) %>%
  filter(week == 26)
rmt_total_week26 <- week_26 %>%
  summarise("total remittance" = sum(rmt_total))
# 27
week_27 <- rmt_nonZero %>%
  group_by(village) %>%
  filter(week == 27)
rmt_total_week27 <- week_27 %>%
  summarise("total remittance" = sum(rmt_total))
# 28
week 28 <- rmt nonZero %>%
  group_by(village) %>%
  filter(week == 28)
rmt_total_week28 <- week_28 %>%
  summarise("total remittance" = sum(rmt_total))
# 29
week_29 <- rmt_nonZero %>%
  group_by(village) %>%
  filter(week == 29)
rmt_total_week29 <- week_29 %>%
  summarise("total remittance" = sum(rmt_total))
 # 30
week_30 <- rmt_nonZero %>%
  group_by(village) %>%
  filter(week == 30)
```

```
rmt_total_week30 <- week_30 %>%
  summarise("total remittance" = sum(rmt_total))
# 31
week_31 <- rmt_nonZero %>%
  group_by(village) %>%
  filter(week == 31)
rmt total week31 <- week 31 %>%
  summarise("total remittance" = sum(rmt_total))
# 32
week_32 <- rmt_nonZero %>%
  group_by(village) %>%
  filter(week == 32)
rmt_total_week32 <- week_32 %>%
  summarise("total remittance" = sum(rmt_total))
week_33 <- rmt_nonZero %>%
  group_by(village) %>%
  filter(week == 33)
rmt_total_week33 <- week_33 %>%
  summarise("total remittance" = sum(rmt_total))
# 34
week_34 <- rmt_nonZero %>%
  group_by(village) %>%
  filter(week == 34)
rmt_total_week34 <- week_34 %>%
  summarise("total remittance" = sum(rmt_total))
# 35
week_35 <- rmt_nonZero %>%
  group by(village) %>%
  filter(week == 35)
rmt_total_week35 <- week_35 %>%
  summarise("total remittance" = sum(rmt_total))
# 36
week_36 <- rmt_nonZero %>%
  group_by(village) %>%
  filter(week == 36)
rmt_total_week36 <- week_36 %>%
  summarise("total remittance" = sum(rmt_total))
# 37
week 37 <- rmt nonZero %>%
  group_by(village) %>%
  filter(week == 37)
rmt_total_week37 <- week_37 %>%
  summarise("total remittance" = sum(rmt_total))
week_38 <- rmt_nonZero %>%
  group_by(village) %>%
  filter(week == 38)
rmt_total_week38 <- week_38 %>%
  summarise("total remittance" = sum(rmt_total))
# 39
week_39 <- rmt_nonZero %>%
  group_by(village) %>%
  filter(week == 39)
```

```
rmt_total_week39 <- week_39 %>%
  summarise("total remittance" = sum(rmt_total))
# 40
week_40 <- rmt_nonZero %>%
  group_by(village) %>%
  filter(week == 40)
rmt total week40 <- week 40 %>%
  summarise("total remittance" = sum(rmt_total))
# 41
week_41 <- rmt_nonZero %>%
  group_by(village) %>%
  filter(week == 41)
rmt_total_week41 <- week_41 %>%
  summarise("total remittance" = sum(rmt_total))
week_42 <- rmt_nonZero %>%
  group_by(village) %>%
  filter(week == 42)
rmt_total_week42 <- week_42 %>%
  summarise("total remittance" = sum(rmt_total))
# 43
week_43 <- rmt_nonZero %>%
  group_by(village) %>%
  filter(week == 43)
rmt_total_week43 <- week_43 %>%
  summarise("total remittance" = sum(rmt_total))
# 44
week_44 <- rmt_nonZero %>%
  group by(village) %>%
  filter(week == 44)
rmt_total_week44 <- week_44 %>%
  summarise("total remittance" = sum(rmt_total))
# 45
week_45 <- rmt_nonZero %>%
  group_by(village) %>%
  filter(week == 45)
rmt_total_week45 <- week_45 %>%
  summarise("total remittance" = sum(rmt_total))
# 46
week 46 <- rmt nonZero %>%
  group_by(village) %>%
  filter(week == 46)
rmt_total_week46 <- week_46 %>%
  summarise("total remittance" = sum(rmt_total))
# 47
week_47 <- rmt_nonZero %>%
  group_by(village) %>%
  filter(week == 47)
rmt_total_week47 <- week_47 %>%
  summarise("total remittance" = sum(rmt_total))
# 48
week_48 <- rmt_nonZero %>%
  group_by(village) %>%
  filter(week == 48)
```

```
rmt_total_week48 <- week_48 %>%
  summarise("total remittance" = sum(rmt_total))
# 49
week_49 <- rmt_nonZero %>%
  group_by(village) %>%
  filter(week == 49)
rmt total week49 <- week 49 %>%
  summarise("total remittance" = sum(rmt_total))
## Average remittance income in each village per week (tables)
```{r}
# 1
rmt_mean_week1 <- week_1 %>%
  summarise("average remittance" = mean(rmt_total))
rmt mean week2 <- week 2 %>%
  summarise("average remittance" = mean(rmt_total))
#3
rmt_mean_week3 <- week_3 %>%
  summarise("average remittance" = mean(rmt_total))
#4
rmt_mean_week4 <- week 4 %>%
  summarise("average remittance" = mean(rmt_total))
rmt_mean_week5 <- week_5 %>%
  summarise("average remittance" = mean(rmt_total))
#6
rmt_mean_week6 <- week_6 %>%
  summarise("average remittance" = mean(rmt_total))
rmt_mean_week7 <- week_7 %>%
  summarise("average remittance" = mean(rmt_total))
rmt_mean_week8 <- week_8 %>%
  summarise("average remittance" = mean(rmt_total))
rmt_mean_week9 <- week_9 %>%
  summarise("average remittance" = mean(rmt_total))
#10
rmt_mean_week10 <- week_10 %>%
  summarise("average remittance" = mean(rmt_total))
#11
rmt_mean_week11 <- week_11 %>%
  summarise("average remittance" = mean(rmt_total))
#12
rmt_mean_week12 <- week_12 %>%
  summarise("average remittance" = mean(rmt_total))
rmt mean week13 <- week 13 %>%
  summarise("average remittance" = mean(rmt_total))
rmt_mean_week14 <- week_14 %>%
```

```
summarise("average remittance" = mean(rmt_total))
#15
rmt_mean_week15 <- week_15 %>%
  summarise("average remittance" = mean(rmt_total))
#16
rmt_mean_week16 <- week_16 %>%
  summarise("average remittance" = mean(rmt_total))
#17
rmt_mean_week17 <- week_17 %>%
  summarise("average remittance" = mean(rmt_total))
#18
rmt_mean_week18 <- week_18 %>%
  summarise("average remittance" = mean(rmt_total))
#19
rmt_mean_week19 <- week_19 %>%
  summarise("average remittance" = mean(rmt_total))
#20
rmt mean week20 <- week 20 %>%
  summarise("average remittance" = mean(rmt_total))
rmt_mean_week21 <- week_21 %>%
  summarise("average remittance" = mean(rmt_total))
#22
rmt_mean_week22 <- week_22 %>%
  summarise("average remittance" = mean(rmt_total))
#23
rmt_mean_week23 <- week_23 %>%
  summarise("average remittance" = mean(rmt_total))
#24
rmt_mean_week24 <- week_24 %>%
  summarise("average remittance" = mean(rmt_total))
#25
rmt_mean_week25 <- week_25 %>%
  summarise("average remittance" = mean(rmt_total))
#26
rmt_mean_week26 <- week_26 %>%
  summarise("average remittance" = mean(rmt_total))
rmt_mean_week27 <- week_27 %>%
  summarise("average remittance" = mean(rmt_total))
#28
rmt_mean_week28 <- week_28 %>%
  summarise("average remittance" = mean(rmt_total))
#29
rmt_mean_week29 <- week_29 %>%
  summarise("average remittance" = mean(rmt_total))
#30
rmt_mean_week30 <- week_30 %>%
  summarise("average remittance" = mean(rmt_total))
rmt mean week31 <- week 31 %>%
  summarise("average remittance" = mean(rmt_total))
rmt_mean_week32 <- week_32 %>%
```

```
summarise("average remittance" = mean(rmt_total))
#33
rmt_mean_week33 <- week_33 %>%
  summarise("average remittance" = mean(rmt_total))
#34
rmt_mean_week34 <- week_34 %>%
  summarise("average remittance" = mean(rmt_total))
#35
rmt_mean_week35 <- week_35 %>%
  summarise("average remittance" = mean(rmt_total))
#36
rmt_mean_week36 <- week_36 %>%
  summarise("average remittance" = mean(rmt_total))
#37
rmt_mean_week37 <- week_37 %>%
  summarise("average remittance" = mean(rmt_total))
#38
rmt mean week38 <- week 38 %>%
  summarise("average remittance" = mean(rmt_total))
#39
rmt_mean_week39 <- week_39 %>%
  summarise("average remittance" = mean(rmt_total))
#40
rmt_mean_week40 <- week_40 %>%
  summarise("average remittance" = mean(rmt_total))
#41
rmt_mean_week41 <- week_41 %>%
  summarise("average remittance" = mean(rmt_total))
#42
rmt_mean_week42 <- week_42 %>%
  summarise("average remittance" = mean(rmt_total))
#43
rmt_mean_week43 <- week_43 %>%
 summarise("average remittance" = mean(rmt_total))
rmt_mean_week44 <- week_44 %>%
  summarise("average remittance" = mean(rmt_total))
#45
rmt_mean_week45 <- week_45 %>%
  summarise("average remittance" = mean(rmt_total))
#46
rmt_mean_week46 <- week_46 %>%
  summarise("average remittance" = mean(rmt_total))
#47
rmt_mean_week47 <- week_47 %>%
  summarise("average remittance" = mean(rmt_total))
#48
rmt_mean_week48 <- week_48 %>%
  summarise("average remittance" = mean(rmt_total))
rmt mean week49 <- week 49 %>%
  summarise("average remittance" = mean(rmt_total))
```

. . .