Notes on data processing and analysis of available Papua New Guinea datasets

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1. Processing and preparing data for analysis and mapping

Given Microsoft Excel files containing data from the Papua New Guinea NHIS per month for 2015 and 2016, we would like to read each of these files and then concatenate them into a single dataset. This can be done in R as follows:

```
# Get the filenames of all .XLSX files in folder named "data"
fileNames <- list.files(path = "data/")
# Create a concatenating object
png maternal <- NULL
# Loop through each of the XLSX files in data and read them
for(i in fileNames) {
  # Use read_xlsx() to read current filename
  temp <- read_xlsx(path = paste("data/", i, sep = ""),</pre>
                    col_names = FALSE,
                    skip = 3)
  #
  # extract month of current data
  month <- str_split(string = i, pattern = " ")[[1]][1]</pre>
  # extract year of current data
  year <- str_split(string = str_split(string = i, pattern = " ")[[1]][2],</pre>
                    pattern = ".xlsx")[[1]][1]
  # Add month variable to temp dataset
  temp$month <- month
```

```
# Add year variable to temp dataset
#
temp$year <- year
#
# concatenate current dataset with png_maternal
#
png_maternal <- data.frame(rbind(png_maternal, temp))
}</pre>
```

This results in a data frame object called png_maternal with 36 columns and 16096 rows. The resulting data frame is as follows (first 60 rows):

X1 X2	X3	X4	X	5 X	6 X	7 }	(8)	9 X	10 X	11 X_	12 X	13 X	14 X	15 X	16 X	17 X	18 X	19 X	Z_20 X	21 X	22 X_	23 X_	_24 X	25 X_	_26 X_	_27 X_	28 X_	29 X_	_30 X_	31 X	_32 X;	33 X3	4 month	year
10102 BALIMO HP	1	2		0	17	1	0	0	0	0	7	28	62	0	0	0	44	32	165	0	0	0	1	42	0	6	1	0	0	4	0	0	2 apr	2015
10104 BOSSET SC	1	1		0	0	0	0	0	0	0	1	2	18	0	0	0	90	4	20	1	2	5	0	5	0	0	0	0	0	0	0		0 apr	2015
10106 KAMASI CLINIC	1	0		3	5	0	0	0	0	0	0	6	10	0	0	0	6 10	6	3	0	0	0	0	3	0	0	1	0	0	0			0 apr	2015
10107 MAPODA SC 10108 NOMAD HC	1	0		0	0	0	0	0	0	0	1	0	4	0	0	0	10	0	14	0	0	0	0	10	0	0	0	0	0	0	0		0 apr 0 apr	2015 2015
		0			0	0	0	0	0	0	1	0		0	0	0		0	- 0	0	0	0	0	0	0		0	0	0	0	0	0		
10109 OBO/ KAVINANGA SC 10110 WASUA SC	1	3	1	9	5	0	0	0	0	0	9	8	23	0	0	0	9	0	38	2	0	3	0	5	0	1	0	0	0	0	0	0	0 apr 1 apr	2015 2015
10110 WASOA SC 10111 WAWOIL SC	1	0		0	3	0	0	0	0	0	0	0	19	0	0	0	9	0	11	0	0	0	0	1	0	0	0	0	0	0	0		1 apr 1 apr	2015
10112 ADIBA SC	1			1	0	0	0	0	0	0	5	3	22	0	0	ő	3	0	18	0	0	0	0	5	0	0	2	ő	0	0			0 apr	2015
10114 TAPILA SC	1	0		0	5	0	0	0	0	0	0	5	5	0	0	0	11	6	22	0	1	2	0	1	0	0	1	7	0	0	0	0	0 apr	2015
10202 DEBEPARI SC	1	3		0	0	0	0	2	0	0	1	1	0	25	0	0	3	1	10	0	0	0	0	0	0	0	0	0	0	0	0	0	0 apr	2015
10203 DOME SC	1	0		0	0	0	0	0	0	0	0	0	0	0	0	0	0	2	7	0	0	3	0	0	0	0	0	0	0	0	0	0	0 apr	2015
10204 GOLGOBIP SC	1	0		0	0	0	0	0	0	0	0	0	7	0	0	0	1	0	0	1	0	0	0	0	0	0	0	0	0	0	0		0 apr	2015
10205 HAEWENAE SC	1	1		0	1	0	0	0	0	0	0	0	10	0	0	0	3	0	11	2	0	0	0	0	0	0	0	0	0	0	0	0	0 apr	2015
10206 IOWARA SC	1	1		0	0	0	0	0	0	0	1	1	1	0	0	0	8	1	29	1	0	2	0	3	0	0	0	0	0	2	0	0	1 apr	2015
10208 KIUNGA HP 10209 KIUNGA CATHOLIC UC	1	9		0	10	0	0	0	0	0	19	71 21	513 83	0	0	0	54	30 5	287 71	21	18	35	0	72	0	0	0	0	0	2	6	0	1 apr 0 apr	2015
10210 KUNGIM SC	1	0		9	4	0	0	0	0	0	0	5	27	0	0	0	24	0	0	9	0	14	0	9	0	1	0	0	0	1	0		0 apr 0 apr	2015 2015
10211 MATKOMNAI SC	1	4		0	2	0	0	0	0	0	19	31	35	0	0	0	4	7	6	1	3	5	0	10	0	0	0	0	0	1	1	0	1 apr	2015
10212 MOUGULU SC	1	2		1	9	0	0	0	0	0	2	3	26	0	0	0	16	9	29	0	0	0	0	9	0	0	0	0	0	0	0	0	1 apr	2015
10213 NINGERUM HC	1	1		1	5	0	0	0	0	0	11	26	93	0	0	0	13	9	38	7	0	6	0	5	0	0	0	0	0	0	0	0	0 apr	2015
10214 OLSOBIP HC	1	0		0	2	0	0	0	0	0	0	0	3	0	0	0	1	1	0	0	0	0	0	1	0	0	0	0	0	0	0	0	0 apr	2015
10215 RUMGINAE HC	1			1	20	0	0	0	0	0	1	14	68	0	0	0	6	2	38	3	1	4	0	15	0	0	0	0	0	0	-		0 apr	2015
10216 TABUBIL (OTM LTD) HE	1	11		0	27	0	0	0	0	0	25	63	382	0	0	0	77	0	370	24	0	57	0	69	0	5	0	1	0	3	7		0 apr	2015
10217 TARAKBITS SC	1	0		0	2	0	0	0	0	0	0	2	12	U	U	0	4	1	1	0	0	0	0	0	0	0	0	0	0	0	0	0	p.	2015
10218 MEMBOK SC 10219 TABUBIL UC	1	0		0	1	0	0	0	0	0	2	8	4	0	0	0	6	0	17	4	0	2	0	2	0	0	0	0	0	1	1	0	1 apr 0 apr	2015
10301 DARU HOSPITAL	1	0		0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	64	0	1	0	0	0	2	0		0 apr 0 apr	2015 2015
10301 DARU UC	1	4		0	12	0	0	0	0	0	62	0	259	0	0	2	61	30	167	22	4	31	0	0.4	0	0	0	0	0	0			0 apr	2015
10304 KUNINI SC	1	4		8	6	0	0	0	0	0	24	37	54	0	1	4	6	16	21	0	0	0	0	0	0	0	0	0	0	0	0	0	0 apr	2015
10305 MABUDAWAN HC	1	0		0	0	0	0	0	0	0	13	9	18	0	0	11	3	5	0	0	0	0	0	5	0	0	0	0	0	0	0	0	0 apr	2015
10306 MOREHEAD SC	1	3		2	4	0	0	0	0	0	1	2	53	0	0	0	6	0	22	0	0	0	0	2	0	0	0	0	0	0	0	0	0 apr	2015
10308 SUKI (GIGWA) SC	1			20	41	0	0	0	0	0	14	42	31	0	0	0	35	13	21	8	5	25	0	5	0	0	0	0	0	0	0		0 apr	2015
10310 TEAPOPO HC	1	13		0	52	0	0	0	0	0	30	24	81	0	0	0	66	0	13	53	8	0	0	1	0	0	0	0	1	0	0		0 apr	2015
10311 UPIARA SC	1	0		0	1	0	0	0	0	0	0	1	6	0	0	0	4	0	8	0	0	0	0	3	0	1	0	0	0	2	1	0	0 apr	2015
20101 BEMA SC	1	0		0	0	0	0	0	0	0	0	0	0	0	0	0	9	0	29	0	0	0	0	3	0	0	0	0	0	5	1	0	1 apr	2015
20102 HAWABANGO SC 20104 KAINTIBA HC	1	0		0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0 11	0	0	0	0	0	0	0	0	0	0	0			0 apr 0 apr	2015 2015
20104 KAINTIBA HC 20106 KAMINA HC	1			0	-4-	0	0	0	0	0	0	2	3	0	0	0	8	0	27	0	0	0	0	6	0	0	0	0	0	6	0		0 apr 0 apr	2015
20107 KANABEA HC	1	0		0	4	0	0	0	0	0	0	0	28	0	0	0	26	10	42	15	4	6	0	2	0	1	0	0	2	6	0	0	2 apr	2015
20108 KEREMA HOSPITAL	1	0		0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	36	0	13	1	0	2	0	7	0	3 apr	2015
20109 KOARU SC	1	0		2	2	0	0	0	0	0	0	3	7	0	0	0	19	8	38	0	0	0	0	8	0	0	0	0	1	2	0		0 apr	2015
20110 KUKIPI HC	1	0		0	3	0	0	0	0	0	0	0	5	0	0	0	6	18	0	3	0	1	0	8	0	0	0	0	0	0	0		0 apr	2015
20111 LESE AVIHARA SC	1	0		0	3	0	0	0	0	0	0	1	14	0	0	0	6	4	14	0	0	0	0	2	0	0	0	0	0	0	0	0	1 apr	2015
20112 MALALAUA HC	1	0		0	5	0	0	0	0	0	2	6	34	0	0	0	14	12	58	2	0	8	0	14	0	0	0	0	0	1	0	0	4 apr	2015
20113 MURUA SC	1	1		0	0	0	0	0	0	0	0	0	1	0	0	0	1	2	5	0	0	0	0	3	0	0	0	0	0	0	0	0	0 apr	2015
20114 PUTEI SC 20115 TERAPO SC	1	0		0	0	0	0	0	0	0	0	0	0	0	0	0	4 25	10 30	0 45	0	0	0	0	0 15	0	0	0	0	0	0	0		0 apr 0 apr	2015 2015
20116 KEREMA UC	1			2	13	0	0	0	0	0	1	20	101	0	0	0	48	33	130	17	3	18	0	0	0	0	0	0	0	0	0		0 apr 0 apr	2015
20201 BAIMURU HC	1	0		0	0	0	0	0	0	0	9	17	32	0	0	4	7	1	62	0	0	0	0	3	0	0	0	0	0	2	0		0 apr	2015
20202 IHU HC	1	0		0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	7	1	2	0	0	0	4	1		0 apr	2015
20202 HIC HC 20203 KAPUNA HC	1	2		1	5	0	0	9	1	1	0	5	40	0	0	0	17	22	55	4	0	3	0	28	0	4	1	0	0	0	10		0 apr	2015
20204 KARAUTI SC	1	8		0	6	0	0	0	0	144	11	15	53	0	0	300	10	3	36	0	0	0	0	5	0	0	0	0	0	0	0		0 apr	2015
20205 KIKORI HC	1	16		8	17	0	0	4	0	0	8	30	129	0	0	0	33	18	82	6	0	12	0	26	0	4	0	0	0	0	2		0 apr	2015
20208 OROKOLO SC	1	0		0	4	0	0	0	0	0	5	3	66	0	0	0	14	10	24	0	0	0	0	16	0	0	1	0	0	1	0	0	0 apr	2015
30101 BORU SC	1	0		0	0	0	0	0	0	0	0	1	7	0	0	0	1	2	0	1	2	0	0	0	0	0	0	0	0	0	0	0	0 apr	2015
30102 IRUNA HC	1	0		0	4	0	0	0	0	0	0	7	19	0	0	0	12	2	30	6	0	0	0	5	0	0	0	0	0	1	0		0 apr	2015
30103 KUPIANO HC 30104 MOREGUINA HC	1	3		1	16	0	0	0	0	0	0	3	56 42	0	0	0	33 12	15 3	122 30	3	0	9	0	16	0	0	0	0	0	0	3		5 apr 0 apr	2015 2015
30104 MOREGUINA HC 30105 UPULIMA SC	1			0	0	0	0	0	0	0	2	9	18	0	0	0	4	8	13	5	3	4	0	2	0	1	1	0	0	0			0 apr 1 apr	2015
															-						~	-												

The object png_maternal lacks meaningful column names. Also, it would be good to create codes corresponding to the province code and district codes from the code specified in the first column of png_maternal. These can be processed in R as follows:

```
# Get province, district and facility codes
# Extract first two digits from code
png_maternal$pcode <- floor(png_maternal$X_1 / 10000)</pre>
# pad the pcode with a O at the start
png_maternal$pcode <- str_pad(string = png_maternal$pcode,</pre>
                       width = 2, side = "left", pad = "0")
# Extract first 4 digits from code
#
png_maternal$dcode <- floor(png_maternal$X_1 / 100)</pre>
# pad the dcode with a O at the start
png_maternal$dcode <- str_pad(string = png_maternal$dcode,</pre>
                       width = 4, side = "left", pad = "0")
# pad the code with a 0 at the start
png maternal$X 1 <- str_pad(string = png maternal$X 1,</pre>
                      width = 6, side = "left", pad = "0")
#
# Created codebook for PNG maternal mortality data
longName <- c("Five to six-digit facility code",</pre>
           "Name of facility",
           "Report recieved? 1 = YES; 2 = NO",
           "New attendance breastfeeding pills",
           "New attendance combined pills",
           "New attendance injection",
           "Unkown Number 1",
           "Permanent vasectomy",
           "New attendance IUD",
           "New attendance ovulation",
```

```
"New attendance condom",
               "Re-attendance breastfeeding pills",
               "Re-attendance combined pills",
               "Re-attendance injection",
               "Re-attendance IUD",
               "Re-attendance ovulation",
               "Re-attendance condom",
               "Antenatal first visit",
               "Antenatal fourth visit",
               "Antenatal other",
               "Antenatal TT1",
               "Antenatal TT2",
               "Antenatal booster",
               "Unknown Number 2",
               "Deliveries in health facility",
               "Maternal deaths in facility",
               "Birthweight less than 2500 grams",
               "Stillbirths",
               "Village births supervised",
               "Village births complications",
               "Born before arrival",
               "Delivery complications",
               "Maternal deaths not in facility",
               "Transferred to hospital",
               "Month",
               "Year",
               "Province code",
               "District code")
shortName <- c("code", "facility", "report",</pre>
               "bfpills1", "combpills1", "inj1", "uno1", "vasectomy", "iud1",
               "ovulation1", "condom1", "bfpills2", "combpills2", "inj2",
               "iud2", "ovulation2", "condom2", "anc1", "anc4", "ancother",
               "tt1", "tt2", "ttbooster", "uno2", "delhf", "deadhf", "lbw",
               "still", "vbsup", "vbcomp", "bba", "delcomp", "deadnothf",
               "transhop", "month", "year", "pcode", "dcode")
names(png_maternal) <- shortName</pre>
```

Checking png_maternal object again, we see that the columns have been labelled with more meaningful names and corresponding province and district codes have been added.

SHIPPING SAMENON 1	op month year pco	onth yea	month	mo	p :	юр	юр	ор	p	p r	mo	mon	nontl	onth	h y	year	r pr	cod	le
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90 90 WASSILANS 1	0 apr 2015 01	or 201	apr	apr	0 :	0	0	0	0	0 a	ap	apr	ipr	r	2	2015	5 07	/1	
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020113 MURUA SC	1 apr 2015 02																		
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COLDING TERROP SC 1	0 apr 2015 02	or 201	apr	арл	0	0	0	0	0	0 a	ap	apr	pr	r	2	2015	5 09	12	
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a. Processing png_maternal to create datasets that can be mapped

Further processing of png_maternal can be done to allow for mapping of the data at province and district level.

Province level data can be produced from png maternal as follows:

```
#
# Aggregate data by province and per year
#
provincedata <- aggregate(
   cbind(bfpills1, combpills1, inj1, uno1, vasectomy, iud1, ovulation1,
        condom1, bfpills2, combpills2, inj2, iud2, ovulation2, condom2,
        anc1, anc4, ancother, tt1, tt2, ttbooster, uno2,
        delhf, deadhf, lbw, still, vbsup, vbcomp, bba,
        delcomp, deadnothf, transhop) ~ pcode + year,
   data = png_maternal, FUN = sum)</pre>
```

This produces a data frame object named provincedata.

	pcode	year	bfpills1	combpills1	inj1	uno1	vasectomy	iud1	ovulation1	condom1	bfpills2	combpills2	inj2	iud2	ovulation2	condom2	anc1	anc4	ancother	tt1	tt2	ttbooster	uno2	delhf	deadhf	lbw	still	vbsup	vbcomp	bba	delcomp	deadnothf	transhop
1	01	2015	844		2187	433		196	80	390	2995		15682	82	128	694		1835	11048		466	2036	433	3393	4	166	33	101		142	165	3	48
2	02	2015 2015	315 141	177 171	721 855	3		46 93			475 218	795 568	4154 4438	3 10	6	3319 335	2780	1654 1558		574 1185		815 1491	3 72	1625 2232	5	120 63	44 45			257 167	209 121	7	67 179
4	04	2015	411		1863	72 31	0			110 50	466		18634	19	56 174		8240		23229			4561	31	2232	8			94	0		0	0	0
5	05	2015	547		2021	93	17		200	485	972	3490	9394	4	1529	260	6904		29183			3861	93	5287		590		467		200	792	14	213
6	06	2015	312	300	1061	118	0	27	123	543	534	1316	6158	43	70	524	4132	2000	10936	1708	856	1626	118	2575	3	219	65	127	26	218	349	1	485
7	07	2015	1316		2818	43	7		133	1438	4004		11454	64	231	1551	7116		11922			2550	43	4344		230	91	79		305	448	5	196
8	08 09	2015 2015	458 330		1233 2797	938 69	644	104 85	263 263	898 642	622 291	2689	4753 16185	103	213 92	1185 744	6708 8010			2384 2975		3047 3567	938 69	3860 6472		171 285		118 31	29	358 137	402 275	22	191 231
10	10	2015	829		2398	190		428	131	3027	855	1845	6932	51	130	1985	6069		13548			3284	190	4872		292		91		241	794	3	408
11	11	2015	1782		3284	145		1310	51	1608	3058		14809	350	120		10323		23602			4921	145	7990		160	90	492		145	731	6	293
12	12	2015	1397		3329	87		284	112	3195	2228	10355		7	175		13667		33783			6861	87	4505		545		939		291	227	22	213
13	13	2015	1035		4584	190	48		61	825	1505	3066	10910	148	120		11060			5178		5507	190	5004	15			213		390	645	14	178
14 15	14 15	2015 2015	495 272		1226 1245	16	4 16		81 36	623 41	808 412	4061 1625	6901 6966	0 28	93 12	262 98	6598 7678		13053	3090 1694		2369 2169	16	4203 2815		278 324		99 253		275 363	262 331	6 11	129 150
16	16	2015	27		280	15		39	36	87	22	69	1526	38	26					318		545		1161				222		43	232	0	28
17	17	2015	9		415	32	0		88	161	5	75	1902	64	154	58	1264 3861		14568		614	1940	32	2332		140		184		128	232	0	152
18	18	2015	199	186	1649	80	52	170	1556	524	123	700	4076	3	1084	139	9292	5200	29804	2545	2080	6417	80	8562	15	410	137	234		324	968	11	494
19	19	2015	518		1129	448	40		80	638	854	2450	4924	191	77	926	8311			2378		4627	448	5086		381		179		310	521	3	282
20	20	2015	125		383	49		6		780	214	381	1910	4	313		3617					1597		2707		123		50		140	278	2	148
21 22	21 22	2015 2015	1127 717		1577 1375	14	0	10	1069 65	16645 3099	2494 867	3148 2106	4478 6588	0 17	213 37	8874	4752 3977			1998 1560	948 719	2263 2039	0	3074 2561	6	71 111	41 58	235 53		356 107	138 355	5 10	127 130
23	01	2016	580		1345	12	57			292	1676	4098	13623	3	299	414	3461			1197	342	1594	12	2174	15		28	101	39		141	9	72
24	02	2016	294		922	19			33	1348	408	667	5827	0	17	2706				781	219	736		1279	8					168	148	1	51
25	03	2016	108	150		67	2		7	180	182	657	4965	3	33	262				1133	297	1363		2168	3			119	35	123	134	7	189
26	04	2016	336		2590	15		42	95	151	438		22417	80	104		11966		29444			6021		14359		1200		3		408	106	0	35
27 28	05 06	2016 2016	398 160	393 176	1842 921	94 90	39 0		41	236 531	748 199	3073 922	9885 6502	1 39	994 166		6561 3443		27858 11267		557	3523 1152		5105 2183		634 125		332 106		205 138	733 269	5 1	236 432
29	07	2016	1181		2474	39	0			823	3086	10093		1	315	2231		2881		2248		3002	39	3939		197				325	364	1	151
30	08	2016	390	672	1463	247	24	597	536	1715	389	2604	4389	29	365	1554	6783	2853	9773	2619	942	3652	247	4008	10	125	41	93	44	304	424	8	144
31	09	2016	504		3452	64	2		187	1465	529		20108	34	90	969	8233		18014			3916	64	6344		180		14		134	253	0	156
32 33	10 11	2016 2016	979 1699		2996 3270	128 97	94 254	711 695	57 30	1863 2225	923 3114	1990	7081 14623	12 135	76 115		6281 11659		14650 29322			3214 5700	128 97	4917 7823	1 28	188 359		39 259		129 277	794 1078	2 12	396 948
34	12	2016	1167		3478	330	28			1000	1477		22415	0	226		11663		25090			8654	330	8088	38			1496		641	234	38	239
35	13	2016	804		3552	96	25		50	306	1261		12492	44	102		9817		18051			4878	96	6332		775		139		392	879	16	189
36	14	2016	475		1533	15			60	1397	605	3858	8123	0	73		5895		11799		852	2254		4018		160		82		264	376	12	81
37	15	2016	427		1515	6	27		12	52	261	718	9216	23	13	94	4528		10533		544	1750 468	6 38	2828		310		225		238	195	10	131
38 39	16 17	2016 2016	34 10	25	318 480	38 50	1	16 125	28 34	146 33	29 5	121 51	1234 2187	3	5 54	108	1195 4033		14304	346 1581		468 2997	50 50	926 3355	0 5	91 268		34 245		48 170	77 386	4	23 234
40	18	2016	153		1071	120		184	3530	326	59	385	3835	0	1648		8289		27101			5884	120			604		176		328	1060	6	452
41	19	2016	282	299	1038	268	18	109	39	418	460	1989	5882	5	59	1133	7760	4026	24983	2186	1019	4342	268	6170	4	431	171	269	68	295	805	2	509
42	20	2016	51		501	56		37	94	209	70	219	1687	15	348	338	3828		10087			1566	56	2773		123		82		115	320	0	133
43	21 22	2016 2016	754 922		1130 1887	2	0	49 90	148 118	4103 1703	2051 1071	2996 2486	4383 7719	0 111	130 161	3739 2153		2456 2177		1601 1816	687 767	2075 5799	2	2876 2586	1 12	65 165	42 35	189 46		348 131	150 505	6	102 104
NA	NA	NA	NA		NA			NA	NA	NA	NA	NA		NA	NA	NA		NA		NA		NA	NA	NA		NA				NA	NA	NA	NA
	NA	NA	NA		NA			NA	NA	NA	NA	NA		NA	NA	NA	NA				NA	NA	NA	NA		NA		NA		NA	NA	NA	NA
NA.2	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
NA.3		NA	NA		NA			NA	NA	NA	NA	NA		NA	NA	NA	NA			NA		NA	NA	NA		NA				NA	NA	NA	NA
NA.4 NA.5	NA NA	NA NA	NA NA	NA NA	NA NA	NA NA		NA NA	NA NA	NA NA	NA NA	NA NA	NA NA	NA NA	NA NA	NA NA	NA NA	NA NA	NA NA	NA NA		NA NA	NA NA	NA NA	NA NA	NA NA		NA NA		NA NA	NA NA	NA NA	NA NA
	NA	NA	NA	NA		NA		NA	NA	NA	NA	NA		NA	NA	NA	NA	NA	NA			NA	NA	NA	NA		NA	NA		NA	NA	NA	NA
NA.7		NA	NA		NA			NA	NA	NA	NA	NA		NA	NA	NA	NA	NA	NA		NA	NA	NA	NA		NA		NA		NA	NA	NA	NA
NA.8	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
NA.9		NA	NA	NA				NA	NA	NA	NA	NA		NA	NA	NA	NA		NA	NA		NA	NA	NA	NA		NA	NA		NA	NA	NA	NA
NA.10		NA	NA		NA			NA	NA	NA	NA	NA		NA	NA	NA	NA	NA	NA		NA	NA	NA	NA		NA		NA		NA	NA	NA	NA
NA.11 NA.12		NA NA	NA NA	NA NA	NA NA			NA NA	NA NA	NA NA	NA NA	NA NA		NA NA	NA NA	NA NA	NA NA	NA NA	NA NA	NA NA	NA NA	NA NA	NA NA	NA NA	NA NA	NA NA	NA NA	NA NA		NA NA	NA NA	NA NA	NA NA
NA.12		NA	NA	NA				NA	NA	NA	NA	NA		NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA		NA	NA		NA	NA	NA	NA
NA.14		NA	NA	NA				NA	NA	NA	NA	NA		NA	NA	NA	NA	NA	NA	NA		NA	NA	NA	NA		NA	NA		NA	NA	NA	NA
NA.15	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA

District level data can be produced from png_maternal as follows:

This produces a data frame object named districtdata.

dcode	year	bfpills1	combpills1	inj1	uno1	vasectomy	iud1 (ovulation1	condom1	bfpills2	combpills2	inj2	iud2	ovulation2	condom2	anc1	anc4	ancother	tt1	tt2	ttbooster	uno2	delhf	deadhf	lbw	still	vbsup	vbcomp	bba d	elcomp d	eadnothf	transhop
0101	2015	139	115	383	3		162	6	32	378		2649	2	2	116		432	2311	228	81	480	3		0	35	16	32	5	26	11	2	9
0102	2015	345		1135	426	2	25	6	12	1023	2547		80	99		2781	777	6576	714	331	1150		1904	4		13	63	8	81	149	0	32
0103	2015	360	259	669	4	1	9	68	346	1594	1152	3543	0	27	562	1235	626	2161	451	54	406	4	833	0	40	4	6	6	35	5	1	7
0201	2015	40	39	413	0	0	0	0	26	51	208	1622	0	2	28		1055	3621	397	92	582	0	837	1	55	26	44	51	180	63	1	52
0202	2015	275	138	308	3	0	46	5	1190	424	587	2532	3	4	3291	858	599	2282	177	43	233	3	788	4	65	18	31	22	77	146	6	15
0301	2015	31	36	192	2	10	0	3	3	23	195	1201	0	12	8	740	305	1908	263	105	267	2	381	1	3	10	14	4	45	27	1	45
0302	2015	8	33	33	1	0	0	2	48	9	24	32	0	1	40	349	116	435	88	102	145	1	66	1	3	3	27	20	38	4	0	1
0303	2015	54	40	413	48	0	93	4	51	98		1577	10	33		2411	938	4868	681	197	901		1241	2		15	37	28	48	66	1	60
0304	2015	48	62	217	21	0	0	0	8	88		1628	0	10		753	199	1804	153	27	178	21		4		17	16	7	36	24	1	73
0401	2015	195	135	634	0	0	40	11	15	201	1664	7196	19	165	322	2806	1362	7530	1051	490	1510	0	0	0	0	0	0	0	0	0	0	0
0402	2015	147	166	681	14	0	25	120	34	181		6893	0	7			1707	10683		612	2036	14	0	0	0	0	0	0	0	0	0	0
0403	2015	69	116	548	17	0	12	2	1	84		4545	0	2		1809	708	5016	674	359	1015	17		0	0	0	0	0	0	0	0	0
0501	2015	184	159	683	54	3	2	40	88	188	1001		0	29			1816	11505		606	1547		2444	2		59	46	47	81	377	0	72
0502 0503	2015 2015	115	75		16 9	14 0	20	2	234	266		1652	0	108		1466 1816	952	6061 6432	423 566	218 280	748		1075		126	31	27	19	35 36	151 139	9	42
		206	241	382	9		3	152	130	482	1602		0	768			1165				842	9		5		21	327	22	30		3	49
0504	2015	42	31	465	14	0	28	6	33	36		1668	4	624		1122		5185	351	199	724		893		100	26	67	29	48	125	2	50
0601	2015	182	181		45	0	4	11	497	363		3690	1	59		2509	1120	7798	981	568	910	45			215	65	59		182	327	0	266
0602	2015	130	119		73 0	0	23	112	46	171		2468	42	11				3138		288	716	73		0	-	6	68		36	22	1	219
0701 0702	2015 2015	156 141	211 72	683 226	7	0	12	37 11	146 270	235 542	1275	3066 1537	64	13 7		1340 812	378 594	2358 902	652 348	347 102	597 202	7	598 283	0	23	1	19	13 6	44 10	20 9	0	44 21
																										-						
0703	2015	163	121	354	0	6	0	4	354	342	1059		0	64		1063	382	1824	470	112	397		243	0	164	3	2	13	27	11	2	15
0706 0707	2015 2015	409 447	417 294	957 598	5 31	1 0	17 0	21 60	474 194	2066 819	2104 1555		0	44 103		1993 1908	597 1052	3107 3731	525 676	175 414	608 746	5 31	2418 802	1 2	164 36	49 32	12 46	8 17	58 166	354 54	0 2	35 81
0801	2015	81	76	68	0	3	0	55	46	68	85	76	0	57		468	200	474	129	84	118	0		0	3	1	40	2	45	3	0	18
0802	2015	47	163	339	334	168	68	159	659	111		1053	75	89		1044	540	1344	444	130	469	334		8	26	11	35	3	70	47	0	30
0803	2015	196	136	161	71	123	10	6	14	254	821	738	0	16		2486	210	3131	824	268	1280		1319	2		21	13		116	145	13	52
0804 0805	2015 2015	46 88	179 174	261 404	162 371	98 252	0 26	43	30 149	76 113	389 1016	1192	28	51		1290 1420	442	1739 1936	536 451	74 140	571 609	371	1290 680	1 2	13 32	23 16	25 5	12 2	75 52	108 99	8	29 62
0902	2015	70	169	463	37	0	13	16	501	60		1929	24	43		1079	598	2171	613	391	487	37	462	0		4	22	2	31	26	3	102
0903	2015	47	92		0	0	0	142	7	70		1952	0	6		747	386	1410	240	87	310		272	0		4	2	9	10	15	0	26
0904	2015	129			25	0	42		18	103	3074		16	19			2093			1028						63	1	3	88	211	0	33
0904	2015	84	104	1396 574	7	2	30	11 94	116	58		3225	0	24		1581	719	11341 3363	1644 478	208	2154 616	25 7			262 10	7	6	10	8	23	0	70
1001	2015	116	61	300	63	18	5	5	795	153		1284	0	5	577	867	424	1745	425	138	430	63	453	0	16	13	13	1	15	30	0	14
1002	2015	34	37		5	0	10	4	1	33	90	407	0	0	0	554	187	1305	195	54	311	5		0	1	0	0	2	1	3	0	12
1003	2015	66	64	341	6	63	51	34	64	131	147	863	8	81	88	652	417	619	264	248	254	6	185	0	5	1	23	3	36	16	0	25
1004	2015	178	156	464	63	4	21	86	307	119	260	1231	0	37	285	1676	972	3177	649	232	1079	63	1324	1	33	10	49	5	34	135	1	278
1005	2015	389	488	991	52		341	0	1783	353		2700	43	1	990	1979	739	5953	710	287	985		2495	7		73	6		143	596	2	69
1006	2015	46	121		1	0	0	2	77	66	236	447	0	6	45	341		749	175		225		129	0		1	0		12	14	0	10
1101	2015	50	69	186	0	0	57	1	6	76	258	864	0	4	76	814	513	2236	249	107	404	0		0	2	1	3	2	2	5	1	35
1102	2015	395	254	985	20	130	582	29	250	1190	5348	5422	270	69	1630	3055	1573	8766	1286	868	1490	20	4734	10	0	47	0	1	24	505	0	1
1103	2015	311	250	438	15	14	151	3	438	534	1230	1960	14	3	716	1290	1187	2237	894	340	751	15	455	0	37	3	124	14	14	26	0	36
1104	2015	282	195	398	3	53	43	5	57	319	1376		3	8		1612		3589	642	338	942		1420	2	93	31	142	3		132	0	98
1105	2015	80	130	197	5		201	1	56	129	518	926	1	11	65	850	492	2029	436	133	556	5		3	5	2	12	6	5	3	3	17
1106	2015	419	467	730	7		138	5	540	469	1002		35	16		1656	936	3177	659	282	400	7		0		3	203	29	20	37	0	71
1107	2015	231	164	309	30	26	33	0	208	326	517	1400	0	0	417	695	335	986	420	168	254	30	156	0	3	1	4	15	8	4	2	8
1108	2015	14	18	41	65	26	105	7	53	15	98	188	27	9	135	351	318	582	239	77	124	65	96	0	5	2	4	6	20	19	0	27
1201	2015	245	312	528	7		125	35	203	217	1045		0	6	227	2430	739	3497	868	380	959	7		1	6	5	217	7	12	16	1	30
1202	2015	119	128	383	36	0	30	3	1242	92		1067	0	17	1059	996	584	1543	381		514	36		5		11	6	7	29	9	3	52
1203	2015	45	107	149	14	0	8	12	288	36		1106	0	28		1220	384	2324	390	136	790	14	70	1	3	1	37	14	14	2	0	28
1204	2015	51	133	217	6	3	4	1	38	115	569	1194	0	9	27	558	325	1729	220	72	173	6	163	1	1	3	181	4	14	2	1	3
1205	2015	488	329	776	1	2	4	12	40	809	5378	6787	0	110	129	3726	1682	9531	1642	691	2098	1	1800	0	393	110	38	2	131	17	0	0
1206	2015	180	128	349	1	18	63	0	471	603	1674		7	2		1656	915	7682	773	463	1055	1		0		15	123	10		54	1	18
1207	2015	104	69	411	22	15	21	10	422	108		1135	0	2		1211	558	2973	371	145	383	22	336	1	50	10	167	62	21	23	2	14
1208	2015	73	64		0	2	29	3	104	96		1932	0	0		1033	548	2439	369	239	486	0			6	5	50	3		29	2	29
1209	2015	92	95	343	0	42	0	36	387	152	227	1473	0	1	603	837	518	2065	306	238	403	0	458	3	41	32	120	10	14	75	12	39
1301	2015	37	39	507	3	30	76	3	25	26	166	899	121	1	36	1582	427	2287	523	129	784	3	458	0	68	17	15	31	44	34	2	26
1302	2015	736		2271	154	8	10	23	247	1014	1423	4989	2	71			2518	7561		1574	3230	154			399	36	73		183	360	1	40
1303	2015	42	28	302	0	0	12	8	26	180	294	723	0	0		1100	435	1837	328	64	186	0	316	1	13	6	0	20	24	11	5	8
	2015	126	192	290	29	6	12	18	486	169	429	888	5	22			715	1608	226	83	198	29	289	0		16	53	26	32	22	2	18
1305	2015	61	59	299	0	0	1	7	5	46	251	1746	0	16	12	1578	638	3083	589	207	558	0	972	1	120	23	24	7	33	106	1	26

To be able to use provincedata and districtdata, we will need to standardise the raw counts. The usual way to standardise is by calculating rates usually expressed per 10,000 or per 100,000 of a particular population. However, data on specific populations required for the indicators of interest are not available.

As an alternative, we can use available population data per province and per district as a standardising factor to be able to compare the raw counts with each other. It should be made clear that these are not the same as the standard rates hence are not comparable to those. However, standardised values will allow comparison of values across provinces and districts to show general trends rather than on specific absolute values.

Population data at province and district level of Papua New Guinea is available via the papuanewguinea R package. Province population data can be accessed in R via a call to pop_adm1. District population data can be accessed in R via a call to pop_adm2.

The data frame pop_adm1 is as follows:

ADM1_EN	ADM1_PCODE	ADM0_EN	${\rm ADM0_PCODE}$	WARDS	UNITS	HOUSEHOLDS	PERSONS	MALES	FEMALES	WRA
Western Province	PG01	Papua New Guinea	PG	307	872	36800	234745	120970	113775	59618
Gulf Province	PG02	Papua New Guinea	PG	144	531	25819	158197	81814	76383	40025
Central Province	PG03	Papua New Guinea	PG	209	1109	41309	269756	142231	127525	66823
National Capital District	PG04	Papua New Guinea	PG	9	520	47559	364125	194834	169291	88708
Milne Bay Province	PG05	Papua New Guinea	PG	391	1169	55262	276512	143791	132721	69546
Northern (Oro) Province	PG06	Papua New Guinea	PG	158	948	34117	186309	97188	89121	46699
Southern Highlands Province	PG07	Papua New Guinea	PG	465	2628	88041	510245	263523	246722	129282
Hela Province	PG21	Papua New Guinea	PG	260	1616	65271	249449	128895	120554	63170
Enga Province	PG08	Papua New Guinea	PG	365	1015	76421	432045	224663	207382	108668
Jiwaka Province	PG22	Papua New Guinea	PG	190	1191	49298	243886	125458	118628	62161
Western Highlands Province	PG09	Papua New Guinea	PG	298	1145	85101	362850	183791	179059	93827
Chimbu (Simbu) Province	PG10	Papua New Guinea	PG	315	1494	81668	407567	214219	193348	101314
Eastern Highlands Province	PG11	Papua New Guinea	PG	259	3393	136992	579825	301048	278777	146079
Morobe Province	PG12	Papua New Guinea	PG	521	2429	130109	674810	350902	323908	169728
Madang Province	PG13	Papua New Guinea	PG	442	1708	86140	493906	257581	236325	123834
East Sepik Province	PG14	Papua New Guinea	PG	625	1481	87465	450343	225933	224468	117621
West Sepik (Sandaun) Province	PG15	Papua New Guinea	PG	336	926	44934	248411	127771	120640	63215
Manus Province	PG16	Papua New Guinea	PG	122	224	10360	60485	31161	29324	15366
New Ireland Province	PG17	Papua New Guinea	PG	138	631	29634	194067	102494	91573	47984
East New Britain Province	PG18	Papua New Guinea	PG	370	790	58458	328369	168760	159609	83635
West New Britain Province	PG19	Papua New Guinea	PG	107	734	50744	264264	138942	125322	65669
Autonomous Region of Bougainville	PG20	Papua New Guinea	PG	81	1069	48233	249358	127990	121368	63597

The data frame pop_adm2 is as follows:

ADM2_EN	ADM2_PCODE	ADM1_EN	ADM1_PCODE	ADM0_EN	${\rm ADM0_PCODE}$	WARDS	UNITS	HOUSEHOLDS	PERSONS	MALES	FEMALES	WRA
North Fly District	PG0102	Western Province	PG01	Papua New Guinea	PG	110	336	15828	96244	49596	46648	24444
Middle Fly District	PG0101	Western Province	PG01		PG	125	308	11661	79349	40891	38458	
South Fly District	PG0103	Western Province	PG01	Papua New Guinea	PG	72	228	9311	59152	30483	28669	
Kikori District Kerema District	PG0202 PG0201	Gulf Province Gulf Province	PG02 PG02	Papua New Guinea Papua New Guinea	PG PG	77 67	215 316	7810 18009	50966 107231	25812 56002	25154 51229	13181 26844
				-								
Kairuku - Hiri District	PG0303 PG0304	Central Province Central Province	PG03 PG03	Papua New Guinea		76 68	448 259	18147 9099	121586 56509	64150 29539	57436 26970	
Rigo District Abau District	PG0301	Central Province	PG03	Papua New Guinea Papua New Guinea		42	194	7925	55569	29339	26248	13754
Goilala District	PG0302	Central Province	PG03	Papua New Guinea		23	208	6138	36092	19221	16871	8840
National Capital District	PG0401	National Capital District	PG04	Papua New Guinea		9	520	47559	364125	194834	169291	
Alotau District	PG0501	*	PG05		PG	153	470	19226	99539	52076	47463	
Samarai-Murua District	PG0502	Milne Bay Province Milne Bay Province	PG05	Papua New Guinea Papua New Guinea	PG	89	341	11258	58590	30138	28452	
Esa'ala District	PG0504	Milne Bay Province	PG05	Papua New Guinea	PG	88	171	11570	54467	28323	26144	
Kiriwina-Goodenough District	PG0503	Milne Bay Province	PG05	Papua New Guinea	PG	61	187	13208	63916	33254	30662	16067
Sohe District	PG0602	Northern (Oro) Province	PG06	Papua New Guinea		74	461	14413	86547	44977	41570	21783
Ijivitari District	PG0601	Northern (Oro) Province	PG06	Papua New Guinea		84	487	19704	99762	52211	47551	24917
Nipa/Kutubu District	PG0707	Southern Highlands Province	PG07	Papua New Guinea		122	1014	22682	147005	75426	71579	37507
Ialibu/Pangia District	PG0701	Southern Highlands Province	PG07	Papua New Guinea		92	442	12821	63478	32913	30565	16016
Kagua/Erave District	PG0703	Southern Highlands Province	PG07	Papua New Guinea	PG	93	483	14389	74139	39069	35070	18377
Mendi/Munihu District	PG0706	Southern Highlands Province	PG07	Papua New Guinea	PG	92	287	23236	144629	74865	69764	36556
Imbonggu District	PG0702	Southern Highlands Province	PG07	Papua New Guinea	PG	66	402	14913	80994	41250	39744	20826
Koroba/Kopiago District	PG2105	Hela Province	PG21	Papua New Guinea	PG	93	651	22616	73855	38478	35377	18538
Tari/Pori District	PG2108	Hela Province	PG21	Papua New Guinea	PG	69	492	15844	79441	40977	38464	20155
Komo/Magarima District	PG2104	Hela Province	PG21	Papua New Guinea	PG	98	473	26811	96153	49440	46713	24478
Kompiam District	PG0802	Enga Province	PG08	Papua New Guinea	PG	77	194	10465	54624	28404	26220	13739
Wabag District	PG0804	Enga Province	PG08	Papua New Guinea	PG	59	207	12736	73649	38297	35352	18524
Lagaip/Pogera District	PG0803	Enga Province	PG08	Papua New Guinea		103	280	26238	158873	82614	76259	
Kandep District	PG0801	Enga Province	PG08	Papua New Guinea		75	222	12177	73102	38013	35089	
Wapenamanda District	PG0805	Enga Province	PG08	Papua New Guinea	PG	51	112	14805	71797	37335	34462	18058
Anglimp/South Waghi District	PG2201	Jiwaka Province	PG22	Papua New Guinea	PG	84	665	19229	94008	48042	46166	24191
Mul/Baiyer District	PG0905	Western Highlands Province	PG09	Papua New Guinea	PG	119	217	19387	83036	41886	41150	21563
Tambul/Nebilyer District	PG0907	Western Highlands Province	PG09	Papua New Guinea	PG	86	217	16969	75499	37800	37699	19754
Mt Hagen District	PG0903	Western Highlands Province	PG09	Papua New Guinea	PG	42	367	25822	123299	62787	60512	31708
Dei District	PG0902	Western Highlands Province	PG09	Papua New Guinea		51	344	22923	81016	41318	39698	20802
Jimi District	PG2204	Jiwaka Province	PG22	Papua New Guinea	PG	61	176	14177	71379	37381	33998	17815
North Waghi District	PG2206	Jiwaka Province	PG22	Papua New Guinea	PG	45	350	15892	78499	40035	38464	20155
Karimui/Nomane District	PG1003	Chimbu (Simbu) Province	PG10	Papua New Guinea		64	216	10276	52159	27359	24800	
Chuave District	PG1001	Chimbu (Simbu) Province	PG10	Papua New Guinea		57	253	10461	39021	20151	18870	9888
Kundiawa/Gembogl District	PG1005	Chimbu (Simbu) Province	PG10	Papua New Guinea	PG	48	311	14327	78521	41394	37127	19455
Sina Sina Yonggomugl District	PG1006	Chimbu (Simbu) Province	PG10	Papua New Guinea	PG	49	222	11790	56805	29539	27266	14287
Gumine District	PG1002	Chimbu (Simbu) Province	PG10	Papua New Guinea	PG	45	177	12728	56860	30136	26724	14003
Kerowagi District	PG1004	Chimbu (Simbu) Province	PG10	Papua New Guinea		52	315	22086	124201	65640	58561	30686
Obura/Wonenara District	PG1106	Eastern Highlands Province	PG11	Papua New Guinea	PG	64	284	8126	39919	20537	19382	10156
Kainanatu District	PG1104	Eastern Highlands Province	PG11	Papua New Guinea		42	382	28686	126248	65723	60525	31715
Goroka District	PG1102	Eastern Highlands Province	PG11	Papua New Guinea	PG	14	611	23565	103396	53292	50104	26254
Daulo District	PG1101	Eastern Highlands Province	PG11	Papua New Guinea	PG	23	534	14378	45783	24029	21754	11399
Unggai/Benna District	PG1108	Eastern Highlands Province	PG11	Papua New Guinea	PG	20	573	15391	67125	35050	32075	16807
Henganofi District	PG1103	Eastern Highlands Province	PG11	Papua New Guinea	PG	30	543	15191	62904	32503	30401	15930
Lufa District	PG1105	Eastern Highlands Province	PG11	Papua New Guinea	PG	32	209	15387	61057	31749	29308	15357
Okapa District	PG1107	Eastern Highlands Province	PG11	Papua New Guinea	PG	34	257	16268	73393	38165	35228	18459
Tawae/Siassi District	PG1209	Morobe Province	PG12	Papua New Guinea	PG	57	193	10727	54340	28257	26083	13667
Nawae District	PG1208	Morobe Province	PG12	Papua New Guinea		46	190	9030	44556	23169	21387	11207
Huon District	PG1203	Morobe Province	PG12	Papua New Guinea		65	250	16075	77564	40333	37231	19509
Menyamya District	PG1207	Morobe Province	PG12	Papua New Guinea		60	342	17163	87209	45349	41860	
Bulolo District	PG1201	Morobe Province	PG12	Papua New Guinea	PG	91	393	20865	101568	52816	48752	25546
Kabwum District	PG1204	Morobe Province	PG12	Papua New Guinea	PG	66	193	9242	43472	22605	20867	10934
Markham District	PG1206	Morobe Province	PG12	Papua New Guinea	PG	63	306	13352	62495	32497	29998	15719
Lae District	PG1205	Morobe Province	PG12	Papua New Guinea	PG	2	298	21901	148934	77446	71488	37460
Finschafen District	PG1202	Morobe Province	PG12	Papua New Guinea	PG	71	264	11754	54672	28430	26242	
Bogia District	PG1301	Madang Province	PG13	Papua New Guinea	PG	91	260	13770	75067	39035	36032	18881
Sumkar District	PG1305	Madang Province	PG13	Papua New Guinea		64	235	15530	84944	45021	39923	20920
Rai Coast District	PG1304	Madang Province	PG13	Papua New Guinea		89	349	14138	83218	42877	40341	
Usino Bundi District	PG1306	Madang Province	PG13	Papua New Guinea		66	230	10372	60807	32424	28383	
Middle Ramu District	PG1303	Madang Province	PG13	Papua New Guinea		93	282	13334	78892	40515	38377	20110
Madang District	PG1302	Madang Province	PG13	Papua New Guinea		39	352	18996	110978	57709	53269	27913
Ambunti/Drekikier District	PG1401	East Sepik Province	PG14	Papua New Guinea	PG	123	326	15150	71304	35662	35700	18707
Wewak District	PG1404	East Sepik Province	PG14	Papua New Guinea	PG	84	319	16278	87761	43955	43806	22954
Angoram District	PG1402	East Sepik Province	PG14	Papua New Guinea		153	294	16603	98135	50331	47804	25049
Yangoru Saussia District	PG1406	East Sepik Province	PG14	Papua New Guinea		96	186	11680	58878	29247	29631	15527
Wosera Gawi District	PG1405	East Sepik Province	PG14	Papua New Guinea		104	189	12506	62030	30807	31223	16361
Maprik District	PG1403	East Sepik Province	PG14	Papua New Guinea		65	167	15248	72235	35931		19023
Vanimo/Green River District	PG1504	West Sepik (Sandaun) Province	PG15	Papua New Guinea		95	293	11609	69052	36105		17264
Aitape/Lumi District	PG1501	West Sepik (Sandaun) Province	PG15			79	226	11919	72319	37218	35101	18393
Nuku District	PG1502	West Sepik (Sandaun) Province	PG15	Papua New Guinea	PG	80	193	13459	58158	29722	28436	14900
Telefomin District	PG1503	West Sepik (Sandaun) Province	PG15	Papua New Guinea		82	214	7947	48882	24726	24156	
Manus District	PG1601	Manus Province	PG16	Papua New Guinea		122	224	10360	60485	31161		15366
Namatanai District	PG1702	New Ireland Province	PG17	Papua New Guinea	PG	90	348	16437	110905	59043	51862	
Kavieng District	PG1701	New Ireland Province	PG17	Papua New Guinea		48	283	13197	83162	43451	39711	
Gazelle District	PG1801	East New Britain Province	PG18	Papua New Guinea		129	211	24823	129317	66428	62889	
Pomio District	PG1803	East New Britain Province	PG18	Papua New Guinea		118	333	13315	71836	36865	34971	
Kokopo District	PG1802	East New Britain Province	PG18	Papua New Guinea	PG	84	146	13591	87829	45284	42545	
Rabaul District	PG1804	East New Britain Province	PG18	Papua New Guinea	PG	39	100	6729	39387	20183	19204	10063
Kandrian/Gloucester District	PG1901	West New Britain Province	PG19	Papua New Guinea	PG	52	332	16278	74265	38412	35853	18787
Talasea District	PG1902	West New Britain Province	PG19	Papua New Guinea	PG	55	402	34466	189999	100530	89469	46882
South Bougainville District	PG2003	Autonomous Region of Bougainville	PG20	Papua New Guinea		28	351	16330	81675	41654	40021	20971
North Bougainville District	PG2001	Autonomous Region of Bougainville	PG20	Papua New Guinea		27	276	18892	109023	55833	53190	27872
Central Bougainville District	PG2002	Autonomous Region of Bougainville	PG20	Papua New Guinea		26	442	13011	58660	30503	28157	14754

We will need to extract the appropriate information from these population datasets to use for standardising the raw counts. For the type of indicators we will be looking at, the data on population of women of reproductive age would be the most appropriate. This is the data identified as WRA in the population datasets.

The most efficient way to work with this population data and the province and district data we produced awhile ago will be to extract the population of women of reproductive age and then attaching it to the province and district data accordingly.

To do this, we will first need to organise the population data in such a way that it can be merged with the province data.

First, we need to extract the data columns that we need from the population data. These will be the province name, the province code and the women of reproductive age population.

For the province population, this can be done as follows:

```
wra_adm1 <- pop_adm1[ , c("ADM1_PCODE", "ADM1_EN", "WRA")]</pre>
```

This produces the following data frame object:

ADM1_PCODE	ADM1_EN	WRA
PG01	Western Province	59618
PG02	Gulf Province	40025
PG03	Central Province	66823
PG04	National Capital District	88708
PG05	Milne Bay Province	69546
PG06	Northern (Oro) Province	46699
PG07	Southern Highlands Province	129282
PG21	Hela Province	63170
PG08	Enga Province	108668
PG22	Jiwaka Province	62161
PG09	Western Highlands Province	93827
PG10	Chimbu (Simbu) Province	101314
PG11	Eastern Highlands Province	146079
PG12	Morobe Province	169728
PG13	Madang Province	123834
PG14	East Sepik Province	117621
PG15	West Sepik (Sandaun) Province	63215
PG16	Manus Province	15366
PG17	New Ireland Province	47984
PG18	East New Britain Province	83635
PG19	West New Britain Province	65669
PG20	Autonomous Region of Bougainville	63597

For the district population, this can be done as follows:

```
wra_adm2 <- pop_adm2[ , c("ADM2_PCODE", "ADM2_EN", "ADM1_PCODE", "ADM1_EN", "WRA")]</pre>
```

This produces the following data frame object:

ADM2_PCODE	ADM2_EN	ADM1_PCODE	ADM1_EN	WRA
PG0102	North Fly District	PG01	Western Province	24444
PG0101	Middle Fly District	PG01	Western Province	20152
PG0103	South Fly District	PG01	Western Province	15023
PG0202	Kikori District	PG02	Gulf Province	13181
PG0201	Kerema District	PG02	Gulf Province	26844
PG0303	Kairuku - Hiri District	PG03	Central Province	30096
PG0304	Rigo District	PG03	Central Province	14132
PG0301	Abau District	PG03	Central Province	13754
PG0302	Goilala District	PG03	Central Province	8840
PG0401	National Capital District	PG04	National Capital District	88708
PG0501	Alotau District	PG05	Milne Bay Province	24871
PG0502	Samarai-Murua District	PG05	Milne Bay Province	14909
PG0504	Esa'ala District	PG05	Milne Bay Province	13699
PG0503	Kiriwina-Goodenough District	PG05	Milne Bay Province	16067
PG0602	Sohe District	PG06	Northern (Oro) Province	21783
PG0601	Ijivitari District	PG06	Northern (Oro) Province	24917
PG0707	Nipa/Kutubu District	PG07	Southern Highlands Province	37507
PG0701	Ialibu/Pangia District	PG07	Southern Highlands Province	16016
PG0703	Kagua/Erave District	PG07	Southern Highlands Province	18377
PG0706	Mendi/Munihu District	PG07	Southern Highlands Province	36556
PG0702	Imbonggu District	PG07	Southern Highlands Province	20826
PG2105	Koroba/Kopiago District	PG21	Hela Province	18538
PG2108	Tari/Pori District	PG21	Hela Province	20155
PG2104	Komo/Magarima District	PG21	Hela Province	24478
PG0802	Kompiam District	PG08	Enga Province	13739
PG0804	Wabag District	PG08	Enga Province	18524
PG0803	Lagaip/Pogera District	PG08	Enga Province	39960
PG0801	Kandep District	PG08	Enga Province	18387
PG0805	Wapenamanda District	PG08	Enga Province	18058
PG2201	Anglimp/South Waghi District	PG22	Jiwaka Province	24191
PG0905	Mul/Baiyer District	PG09	Western Highlands Province	21563
PG0907	Tambul/Nebilyer District	PG09	Western Highlands Province	19754
PG0903	Mt Hagen District	PG09	Western Highlands Province	31708
PG0902	Dei District	PG09	Western Highlands Province	20802
PG2204	Jimi District	PG22	Jiwaka Province	17815
PG2206	North Waghi District	PG22	Jiwaka Province	20155
PG1003	Karimui/Nomane District	PG10	Chimbu (Simbu) Province	12995
PG1001	Chuave District	PG10	Chimbu (Simbu) Province	9888
PG1005	Kundiawa/Gembogl District	PG10	Chimbu (Simbu) Province	19455
PG1006	Sina Sina Yonggomugl District	PG10	Chimbu (Simbu) Province	14287

Then, we need to organise the population datasets such that the rows of data are in the same sequence as that of the province and district data. This means, the population datasets will have to be ordered in such a way that the province code and district code are sequential. We see that the population datasets are not sequential and as such will need to be re-ordered. This can be done in R as follows:

```
wra_adm1 <- wra_adm1[order(wra_adm1$ADM1_PCODE), ]
wra_adm2 <- wra_adm2[order(wra_adm2$ADM2_PCODE), ]</pre>
```

We now need to adjust the admin codes to match the admin codes in the province and district data. We notice that the population admin codes start with PG whilst the province and district data don't have this. So, we should adjust the population codes by removing the appended PG. This can be done as follows:

```
wra_adm1$ADM1_PCODE <- as.numeric(str_replace(wra_adm1$ADM1_PCODE, "PG", ""))
wra_adm2$ADM2_PCODE <- as.numeric(str_replace(wra_adm2$ADM2_PCODE, "PG", ""))
wra_adm2$ADM1_PCODE <- as.numeric(str_replace(wra_adm2$ADM1_PCODE, "PG", ""))</pre>
```

Once the admin codes have been adjusted, we should now calculate a standardising factor which we will call sf. Using the population size for women of reproductive age (WRA), we divide this by 100,000 to get a standardising factor that will give an indicator value that is per 100,000 WRA population. This can be done as follows:

```
wra_adm1$sf <- wra_adm1$WRA / 100000
wra_adm2$sf <- wra_adm2$WRA / 100000</pre>
```

Once the standardising factor (sf) is calculated, the population data can now be merged with the province and district data respectively.

For the province data, this can be done in R as follows:

For the district data, we will need to do some processing of the district data because there are two additional districts for the National Capital District whilst in the population data and the map data, there is only one. This can be adjusted in such a way that we can collapse the district data for the National Capital District into a single district. This can be done as follows:

```
xy[2,2] <- 2016

districtdata <- data.frame(rbind(
    districtdata[!districtdata$dcode %in% c(401, 402, 403), ], xy))</pre>
```

Once the district data has been adjusted, we can now merge the district data with the district population data. This can be done in R as follows:

We now have processed datasets for province and district data that has all the information needed to produce various analysis outputs.

b. Processing png_maternal to create datasets that can be used for time series analysis

Further processing of png_maternal can be done to allow for time series analysis of the data at monthly intervals for year 2015 and 2016.

Monthly province level data can be produced from png_maternal as follows:

This produces a data frame object named mProvince.

Monthly district level data can be produced from png_maternal as follows:

This produces a data frame object named mDistrict.

month	dcode	year	bfpills1	combpills1	inj1	uno1	l vasectomy	y iud1	ovulation1	condom1	bfpills2	combpills2	inj2	iud2	ovulation2 c	ondom2	anc1	anc4	ancothe	r tt1	tt2	ttbooster	uno2	delhf	deadhf	lbw	still	vbsup	vbcomp	bba	delcomp	deadnothf	transhop
apr	0101	2015	8	24	36	1	L (0 0	0	0	30	59	305	0	0	1	181	51	30			18	1	78	0	7	5	7	0	4	1	0	4
aug	0101	2015	3	10				0 0	-			37	136	0	1	5	41			1 18		34	0	31	0		0	3	2		1	0	0
dec	0101	2015	6	7				0 109					150 159	0	0	85 3	27	13 42		1 17 1 64		8 249	0	78 71			1	2	0		0	0	
feb jan	0101 0101	2015 2015	15 24	1 8				0 0	0	1	60	57 84	352	0	1	2	137 114	69			2	41	1	66	0	3	2	3	1	4	0	0	1
jul	0101	2015	25		37			0 0	0	5	38		175	0	0	5	36	13		7 10		12	0	31	0	0	1	0	0	-	1	0	2
jun	0101	2015	25 15	8				0 0	~	0		55 44	194	0	0	6	86	31	13.			17	0	98	0	0	1	1	0		1	0	0
mar	0101	2015	10		44							48	252	0	0	3	148					34	0	81			3	7			4	2	0
may	0101	2015	18	9	36	1	L (0 0	0	0	30	66	278	0	0	1	91	50	320	5 8	8	24	1	52	0	0	1	4	0	2	0	0	1
nov	0101	2015	11	8	36	0) (0 45	1	2	26	32	115	0	0	0	40	20	11	5 9	3	21	0	20	0	2	0	0	1	0	0	0	0
oct	0101	2015	2	3	14	0) (0 6	0	0	31	23	128	0	0	0	28	14	9	5 7	1	8	0	30	0	1	0	3	0	3	0	0	0
sep	0101	2015	2	8				0 0	1	19		29	405	0	0	5	28	27	9:			14	0	20	0	0	0	2	0		2	0	0
apr	0102	2015	34	10				0 2	0			246		25	0	0	223	68				128	0	188	0		0	1	0		19	0	
aug	0102 0102	2015 2015	22 16	11	25			$\begin{array}{ccc} 0 & 0 \\ 0 & 5 \end{array}$	0	0 5			787 555	0	53 2	2	150 68	53 46				85 31	0	90 56	1	9	0	11 5	1	5 8	10 16	0	2
																												-	1		10	0	0
feb ian	0102 0102	2015 2015	30 37	21	97 55			0 0 1 10	~	0	88 158	244 345		0	6 10		215 817	61 98	62:	5 68 2 98		100 195	426 0	92 123	0		0	4	0		14	0	1
jul		2015	16	12					0	1		193	932	0	10	0						110	0	125	0		2	10	1		12	0	
jun	0102	2015	15	6				0 3	4	2	65	157	607	50	23	0	123	55				74	0	694	0	6	2	4	0		19	0	2
mar	0102	2015	47	9	99	0) (0 0	0	0	98	236	943	0	1	0	247	96	710	94		155	0	138	0	10	1	5	2	10	17	0	3
may	0102	2015	41	6	67	0) (0 0	0	0	93	272	921	0	0	2	173	86	70	5 55	45	114	0	183	3	6	2	12	0	6	10	0	11
nov	0102	2015	34		163				0	0		119	422	0	1	6		65				55	0	78		3	1	2	1		10	0	
oct	0102	2015	28		249			0 0	2	2		188	437	5	1	2	376	40				39	0	74	0	3	1	1	2		10	0	2
sep		2015	25		62			0 0	0	1	70	173	691	0	1	0	128	46				64	0	63	0	6	2	1	0		3	0	0
apr	0103	2015	25		116			0 0	0	0	144	115	502	0	1	17	181	64	25			56	0	80	0	2	0	0	1	5	1	1	0
aug	0103	2015	42		182			0 0	0		130	46	189	0	2	43	86	49	149			25	0	71	0	3	1	0	0		1	0	2
dec feb	0103 0103	2015 2015	21 60	14 76	36			$\begin{array}{ccc} 0 & 0 \\ 0 & 8 \end{array}$	0 58	0 5		46 98	299 107	0	4	30 14	85 85			9 36 5 26		22 38	0	75 76		3	1	0	0		0	0	-
ian	0103	2015	43	33				0 0				175	637	0	1	11	38	36				11	0	69	0	7	1	0	1		0	0	
jul	0103	2015	28		42		1	1 0				43	313	0	4	28	106	46		3 31		48	4	65	0	5	0	0	0	3	1	0	2
jun	0103	2015	21	9	33	0) (0 0	0	0	66	77	153	0	0	11	63	27	7	2 13	0	9	0	74	0	10	0	0	0	3	0	0	0
mar	0103	2015	16	21) (0 0	0	1		76	101	0	0	11	107	77) 29		46	0	70	0	1	0	0	0	6	0	0	0
may	0103	2015	17	16	14	0) (0 0	0	2	203	195	349	0	0	12	72	42	19	22	2	30	0	72	0	2	0	4	1	6	0	0	0
nov	0103	2015	22	21				0 0	~	0		104		0	0	38	129	53		3 49	5		0	32		0	1	0	3		2	0	1
oct	0103	2015	43	11	45	0) (0 1	2	309	172	106	370	0	6	222	141	61	23	7 65	3	48	0	75	0	3	0	2	0	3	0	0	0
sep	0103	2015	22	4				0 0	6					0	8		142			1 49		28	0	74		1	0	0	0		0	0	1
apr	0201	2015	2	5				0 0	-			35	201	0	0	0	167			37		33	0	97	0		2	0			8	0	
aug dec	0201 0201	2015 2015	1	3				0 0		0	3 5	15 10	127 135	0	1	0	171 68	91 74	26 19			45 28	0	73 26	0	1	2	1 3	0		1	0	
feb	0201	2015	7	_	40			0 0	V			21	208	0	1		194	67	330			82	0	76		7	5	3	3		9	0	4
jan	0201	2015	5	2				0 0	0	8	9	25	155	0	0	3	184	43	210		3	66	0	76		7	3	8	13		9	1	11
jul	0201	2015	1	1				0 0	0	0	5	7	84	0	0	0	75					2	0	69	0	3	2	1	2		1	0	
jun	0201	2015	7	0	21	0) (0 0	0	0	3	15	124	0	0	1	209	105	37			61	0	96	0	3	2	1	5	15	3	0	3
mar	0201	2015	3	13	41	0) (0 0	0	2	1	22	111	0	0	0	186	133	33.	5 31	13	54	0	69	0	1	2	0	3	20	7	0	6
may	0201	2015	2	0	29	0) (0 0	0	0	10	22	154	0	0	0	163	96	38	8 28	6	45	0	83	1	8	3	1	6	9	12	0	3
nov	0201	2015	1	5				0 0					132	0	0		147			3 34		29	0	44	0		1	7			0	0	0
oct	0201	2015	4				,	0 0				8	73	0	0	0	157	96				64	0	41	0	3	0	11	2		4	0	3
sep	0201	2015	6	1				0 0			4		118	0	0	1						73	0	87		4	4	8		24	14	0	5
apr	0202 0202	2015 2015	26 13	9				0 13	1 0	145	33 51	70 48	320 227	0	0	304 0	81 66	54 53				15 18	0	85 76	1	10	2	0	0		13 10	0	
-				3		3		0 0	4	164				0	1	437	30					7	3	50	0	3	1	0	0		10	0	
dec feb	0202 0202	2015 2015	10 51	29		-			1	164		10 28	163 209	2	1	300	107	18 34		1 10 2 24		4	3	50	0	3	2	6	2		16	0	
ian	0202	2015	30	23				0 1	0	3		54	213	0	0	0	105	80				12	0	73	0	1	0	6	5	0	12	0	
jul	0202	2015	35	10				0 0	0	301		74	351	0	0	301	94					27	0	65	0	7	1	0	2		13	0	-
jun	0202	2015	19	7) (0 5	0	21	33	56	116	0	0	576	58	27	21			7	0	67	2	9	0	0	0	2	12	1	0
mar	0202	2015	13	0	21	0) (0 2	0	200	67	20	162	1	0	102	65	53	230	5 16	5	73	0	72	0	4	2	4	4	9	14	2	0
may	0202	2015	17	5				0 13	2	0			180	0	0	435	80	59				16	0	86	1	7	3	9	5		24	1	_
nov	0202	2015	14	8				0 0	-	0	23		119	0	0	436	41				1	38	0	30		1	0	0	0		3	0	
oct	0202 0202	2015 2015	32 15	26	31			$\begin{array}{ccc} 0 & 7 \\ 0 & 3 \end{array}$		_	20 32	75 63	262 210	0	0	400	88 43	91 65			4	9	0	71 59		12 4	2	3	3	-	10 18	1	
sep	0202	2019	15	8	31	0	,	0 3	0	541	32	03	210	U	9	400	43	0.0	19.	۷ (1	- (U	59	0	4	9	0	1	0	18	1	0

We can then merge these datasets with the population data that contains the standardising factor. This can be done as follows:

```
mProvince <- merge(wra_adm1, mProvince, by.x = "ADM1_PCODE", by.y = "pcode")</pre>
```

For the district data, we will need to make the same adjustments we did to the district data. This can be done as follows:

```
x <- aggregate(</pre>
       cbind(dcode, year, bfpills1, combpills1, inj1, uno1, vasectomy, iud1,
             ovulation1, condom1, bfpills2, combpills2, inj2, iud2, ovulation2,
             condom2, anc1, anc4, ancother, tt1, tt2, ttbooster, uno2,
             delhf, deadhf, lbw, still, vbsup, vbcomp, bba, delcomp, deadnothf,
             transhop) ~ month,
       data = mDistrict[mDistrict$dcode %in% c(401, 402, 403) &
                           mDistrict$year == 2015, ],
       FUN = sum)
y <- aggregate(
       cbind(dcode, year, bfpills1, combpills1, inj1, uno1, vasectomy, iud1,
             ovulation1, condom1, bfpills2, combpills2, inj2, iud2, ovulation2,
             condom2, anc1, anc4, ancother, tt1, tt2, ttbooster, uno2,
             delhf, deadhf, lbw, still, vbsup, vbcomp, bba, delcomp, deadnothf,
             transhop) ~ month,
       data = mDistrict[mDistrict$dcode %in% c(401, 402, 403) &
                           mDistrict$year == 2016, ],
       FUN = sum)
xy \leftarrow rbind(x, y)
xy$dcode <- 401
xy[1:12, 3] \leftarrow 2015
xy[13:24, 3] \leftarrow 2016
mDistrict <- data.frame(rbind(</pre>
  mDistrict[!mDistrict$dcode %in% c(401, 402, 403), ], xy))
```

The resulting mDistrict data can now be merged with the district population data as follows:

```
mDistrict <- merge(wra_adm2, mDistrict, by.x = "ADM2_PCODE", by.y = "dcode")</pre>
```

We now have processed datasets for time series province and district data that has all the

information needed to produce various analysis outputs.

3. Indicators

Given the Papua New Guinea NHIS data, the following indicators can be possibly calculated:

• Number of pregnant women who has had at least one antenatal care visit (ANC) with a trained health worker per 100,000 women of reproductive age

$$n_{anc1} \div \frac{n_{WRA}}{100000}$$

where:

 n_{anc1} = Number of pregnant women who has had at least 1 ANC visit with a trained health worker

 n_{WRA} = Number of women of reproductive age

• Number of pregnant women who has had at least four antenatal care visits (ANC) with any service provider per 100,000 women of reproductive age

$$n_{anc4} \div \frac{n_{WRA}}{100000}$$

where:

 n_{anc4} = Number of pregnant women who has had at least 4 ANC visits with any service provider

 n_{WRA} = Number of women of reproductive age

• Number of pregnant women who received first tetanus toxoid vaccination per 100,000 women of reproductive age

$$n_{tt1} \div \frac{n_{WRA}}{100000}$$

where:

 n_{tt1} = Number of pregnant women who received first tetanus toxoid vaccination n_{WRA} = Number of women of reproductive age

• Number of pregnant women who received second tetanus toxoid vaccination per 100,000 women of reproductive age

$$n_{tt2} \div \frac{n_{WRA}}{100000}$$

where:

 n_{tt2} = Number of pregnant women who received second tetanus toxoid vaccination n_{WRA} = Number of women of reproductive age

 Number of pregnant women who received tetanus toxoid booster vaccination per 100,000 women of reproductive age

$$n_{ttbooster} \div \frac{n_{WRA}}{100000}$$

where:

 $n_{ttbooster}$ = Number of pregnant women who received tetanus toxoid booster vaccination n_{WRA} = Number of women of reproductive age

• Number of pregnant women who delivered in a health facility per 100,000 women of reproductive age

$$n_{delhf} \div \frac{n_{WRA}}{100000}$$

where:

 n_{delhf} = Number of pregnant women who delivered in a health facility

 n_{WRA} = Number of women of reproductive age

• Number of pregnant women who delivered a low birth weight child per 100,000 women of reproductive age

$$n_{lbw} \div \frac{n_{WRA}}{100000}$$

where:

 n_{lbw} = Number of pregnant women who delivered a low birth weight child

 n_{WRA} = Number of women of reproductive age

 $\bullet\,$ Number of pregnant women who delivered a still birth per 100,000 women of reproductive age

$$n_{still} \div \frac{n_{WRA}}{100000}$$

where:

 n_{still} = Number of pregnant women who delivered

a stillbirth

 n_{WRA} = Number of women of reproductive age

• Number of pregnant women who died during childbirth per 100,000 women of reproductive age

```
n_{deadhf} + n_{deadnothf} \div \frac{n_{WRA}}{100000}
where:
n_{deadhf} = \text{Number of pregnant women who died}
during childbirth at health facility
n_{deadnothf} = \text{Number of pregnant women who died}
during childbirth outside of health facility
n_{WRA} = \text{Number of women of reproductive age}
```

2. Time-series analysis of monthly NHIS data

Using the data frame objects mProvince and mDistricts, we can now produce time-series analysis of specific indicators specified above.

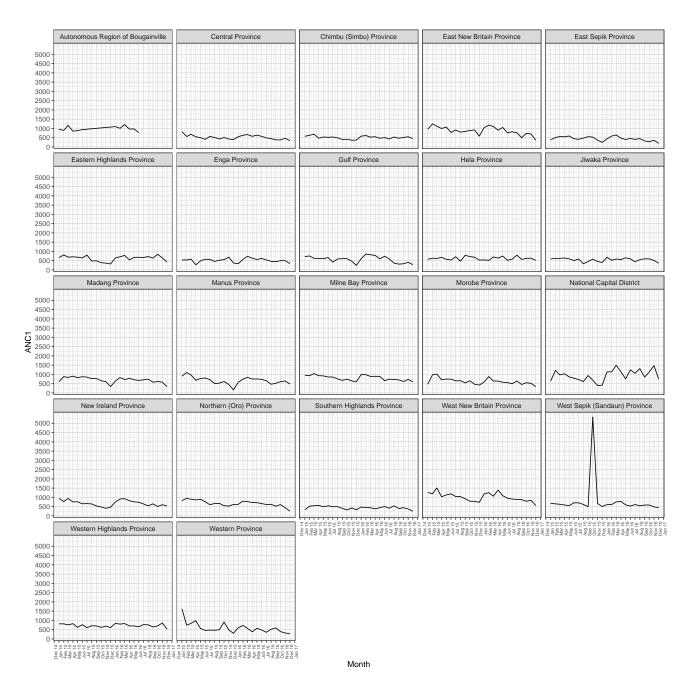
At least one antenatal care visit with a trained health worker

We first work with the province data.

We will work with the data columns labelled anc1 and sf

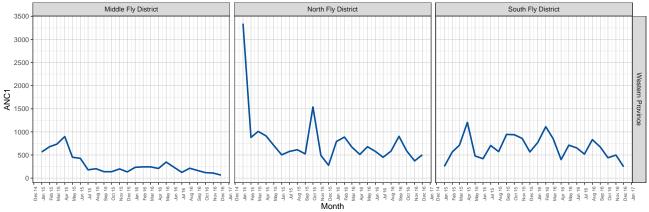
```
temp1 <- aggregate(anc1 ~ ADM1_PCODE + ADM1_EN + month + year,</pre>
                    data = mProvince,
                    FUN = sum)
temp2 <- aggregate(sf ~ ADM1_PCODE + ADM1_EN + month + year,</pre>
                    data = mProvince,
                    FUN = unique)
temp1$anc1Std <- temp1$anc1 / temp2$sf</pre>
temp1$month <- as.character(temp1$month)</pre>
temp1$month[temp1$month == "jan"] <- 1
temp1$month[temp1$month == "feb"] <- 2</pre>
temp1\$month[temp1\$month == "mar"] <- 3
temp1$month[temp1$month == "apr"] <- 4
temp1$month[temp1$month == "may"] <- 5</pre>
temp1$month[temp1$month == "jun"] <- 6
temp1$month[temp1$month == "jul"] <- 7</pre>
temp1$month[temp1$month == "aug"] <- 8</pre>
temp1$month[temp1$month == "sep"] <- 9</pre>
temp1$month[temp1$month == "oct"] <- 10</pre>
temp1$month[temp1$month == "nov"] <- 11
temp1$month[temp1$month == "dec"] <- 12
```

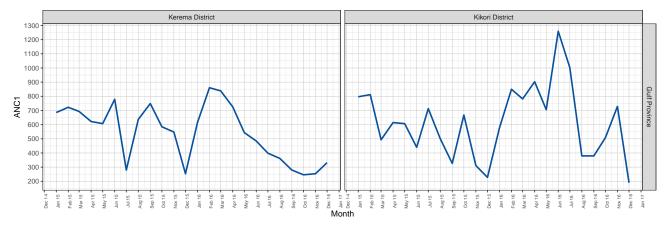
```
temp1$date <- paste(temp1$year, temp1$month, sep = "-")</pre>
temp1$date <- zoo::as.yearmon(temp1$date)</pre>
temp1 <- temp1[order(temp1$date), ]</pre>
temp1 <- temp1 %>%
  group_by(ADM1_EN) %>%
 mutate(anc1Sm = rollmean(x = anc1Std, k = 3, na.pad = TRUE))
temp1long <- gather(data = temp1,</pre>
                     key = "anc1",
                     value = "value",
                     anc1Std, anc1Sm,
                     factor_key = TRUE)
themeSettings <- theme_bw() +</pre>
                    theme(panel.grid.major = element_line(linetype = 1,
                                                            size = 0.2,
                                                            colour = "gray80"),
                          axis.text.x = element_text(size = 6, angle = 90),
                          legend.key = element_rect(linetype = 0),
                          legend.key.size = unit(1, "cm"),
                          legend.position = "top")
```



We now work with the district data.

```
dist1 <- aggregate(anc1 ~ ADM2_PCODE + ADM2_EN + ADM1_PCODE + ADM1_EN + month + year,
                     data = mDistrict,
                     FUN = sum)
dist2 <- aggregate(sf ~ ADM2_PCODE + ADM2_EN + ADM1_PCODE + ADM1_EN + month + year,</pre>
                     data = mDistrict,
                     FUN = unique)
dist1$anc1Std <- dist1$anc1 / dist2$sf</pre>
dist1$month <- as.character(dist1$month)</pre>
dist1$month[dist1$month == "jan"] <- 1</pre>
dist1$month[dist1$month == "feb"] <- 2</pre>
dist1$month[dist1$month == "mar"] <- 3</pre>
dist1$month[dist1$month == "apr"] <- 4</pre>
dist1$month[dist1$month == "may"] <- 5</pre>
dist1$month[dist1$month == "jun"] <- 6</pre>
dist1$month[dist1$month == "jul"] <- 7</pre>
dist1$month[dist1$month == "aug"] <- 8</pre>
dist1$month[dist1$month == "sep"] <- 9</pre>
dist1$month[dist1$month == "oct"] <- 10</pre>
dist1$month[dist1$month == "nov"] <- 11</pre>
dist1$month[dist1$month == "dec"] <- 12</pre>
dist1$date <- paste(dist1$year, dist1$month, sep = "-")</pre>
dist1$date <- zoo::as.yearmon(dist1$date)</pre>
```





```
facet_grid(ADM1_EN ~ ADM2_EN) +
  themeSettings
            Abau District
                                 Goilala District
                                                      Kairuku – Hiri District
                                                                              Rigo District
 1100
 1000
  900
  700
  600
  500
  400
  300
  200
ggplot(dist1[dist1$ADM1_PCODE == 4, ], aes(as.Date(date), anc1Std)) +
  geom_line(colour = "#08519c", size = 1) +
  scale_x_date(name = "Month", date_breaks = "1 month", date_labels = "%b %y") +
  scale_y_continuous(name = "ANC1",
                         breaks = seq(from = 0,
                                        to = max(dist1$anc1Std),
                                        by = 100)) +
  facet_grid(ADM1_EN ~ ADM2_EN) +
  themeSettings
                                           National Capital District
 1500
 1400
 1300
 1200
                                                                                            National Capital District
 1100
  800
  600
  500
  400
ggplot(dist1[dist1$ADM1_PCODE == 5, ], aes(as.Date(date), anc1Std)) +
  geom_line(colour = "#08519c", size = 1) +
  scale_x_date(name = "Month", date breaks = "1 month", date labels = "%b %y") +
  scale_y_continuous(name = "ANC1",
                         breaks = seq(from = 0,
                                        to = max(dist1$anc1Std),
```

by = 100) +

facet_grid(ADM1 EN ~ ADM2 EN) +

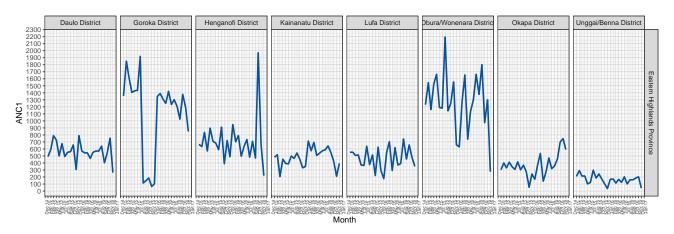
```
Alotau District
                                  Esa'ala District
                                                     Kiriwina-Goodenough District
                                                                             Samarai-Murua District
 1500
 1400
 1200
 1100
 900
  800
  700
  600
  400
ggplot(dist1[dist1$ADM1_PCODE == 6, ], aes(as.Date(date), anc1Std)) +
  geom_line(colour = "#08519c", size = 1) +
  scale_x_date(name = "Month", date_breaks = "1 month", date_labels = "%b %y") +
  scale_y_continuous(name = "ANC1",
                         breaks = seq(from = 0,
                                         to = max(dist1$anc1Std),
                                         by = 100)) +
  facet_grid(ADM1_EN ~ ADM2_EN) +
  themeSettings
                       Ijivitari District
                                                                    Sohe District
 1100
 1000
  900
  800
 600
  500
  400
  300
                                   Jul 16
Jul 16
Aug 16
Sep 16
Oct 16
Nov 16
ggplot(dist1[dist1$ADM1_PCODE == 7, ], aes(as.Date(date), anc1Std)) +
  geom_line(colour = "#08519c", size = 1) +
  scale_x_date(name = "Month", date_breaks = "1 month", date_labels = "%b %y") +
  scale_y_continuous(name = "ANC1",
                         breaks = seq(from = 0,
                                         to = max(dist1$anc1Std),
                                         by = 100)) +
  facet_grid(ADM1_EN ~ ADM2_EN) +
```

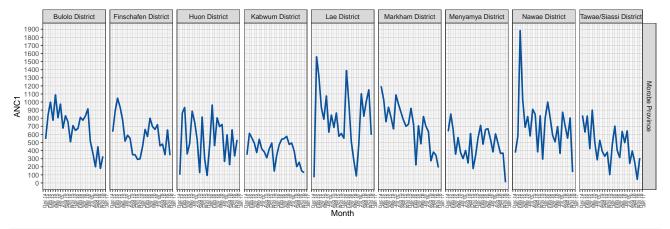
```
| Ialibu/Pangia District | Imbonggu District | Kagua/Erave District | Mendi/Munihu District | Nipa/Kutubu District
```



```
Dei District
                                              Mt Hagen District
 800
 700
 600
ANC 500
 400
 300
ggplot(dist1[dist1$ADM1_PCODE == 10, ], aes(as.Date(date), anc1Std)) +
  geom_line(colour = "#08519c", size = 1) +
  scale_x_date(name = "Month", date_breaks = "1 month", date_labels = "%b %y") +
  scale_y_continuous(name = "ANC1",
                          breaks = seq(from = 0,
                                          to = max(dist1$anc1Std),
                                          by = 100)) +
  facet_grid(ADM1_EN ~ ADM2_EN) +
  themeSettings
        Chuave District
                       Gumine District
                                     Karimui/Nomane District
                                                      Kerowagi District
                                                                  Kundiawa/Gembogl District
                                                                                 Sina Sina Yonggomugl District
 1200
 1100
 1000
  900
  800
  700
  500
  400
  300
  100
    84244778584844778584844778584844778584844778584844778584844778584844778584844778584844778584844778584844778584
ggplot(dist1[dist1$ADM1_PCODE == 11, ], aes(as.Date(date), anc1Std)) +
  geom_line(colour = "#08519c", size = 1) +
  scale_x_date(name = "Month", date_breaks = "1 month", date_labels = "%b %y") +
  scale_y_continuous(name = "ANC1",
                          breaks = seq(from = 0,
                                          to = max(dist1$anc1Std),
                                          by = 100)) +
```

facet_grid(ADM1_EN ~ ADM2_EN) +





```
Bogia District

Madang District

Middle Ramu District

Rai Coast District

Sumkar District

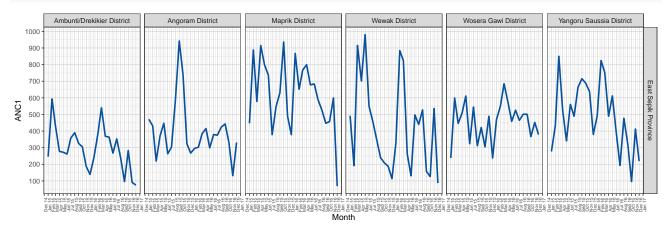
Usino Bundi District

Usino Bundi District

Usino Bundi District

Madang District

On the control of the cont
```



```
Aitape/Lumi District
                                   Nuku District
                                                         Telefomin District
                                                                             Vanimo/Green River District
 18000
 17000
 16000
15000
 14000
 13000
 12000
11000
10000
 9000
8000
7000
  6000
  5000
  4000
3000
  2000
ggplot(dist1[dist1$ADM1_PCODE == 16, ], aes(as.Date(date), anc1Std)) +
  geom_line(colour = "#08519c", size = 1) +
  scale_x_date(name = "Month", date_breaks = "1 month", date_labels = "%b %y") +
  scale_y_continuous(name = "ANC1",
                         breaks = seq(from = 0,
                                         to = max(dist1$anc1Std),
                                         by = 100) +
  facet_grid(ADM1_EN ~ ADM2_EN) +
  themeSettings
                                              Manus District
 1100
 1000
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  800
  500
  400
  300
  200
ggplot(dist1[dist1$ADM1_PCODE == 17, ], aes(as.Date(date), anc1Std)) +
  geom_line(colour = "#08519c", size = 1) +
  scale_x_date(name = "Month", date_breaks = "1 month", date_labels = "%b %y") +
  scale_y_continuous(name = "ANC1",
                         breaks = seq(from = 0,
                                         to = max(dist1$anc1Std),
```

by = 100)) +

facet_grid(ADM1_EN ~ ADM2_EN) +

```
Kavieng District
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                                            Mouth
Jan 15
Feb 15
Apr 15
ggplot(dist1[dist1$ADM1_PCODE == 18, ], aes(as.Date(date), anc1Std)) +
  geom_line(colour = "#08519c", size = 1) +
  scale_x_date(name = "Month", date_breaks = "1 month", date_labels = "%b %y") +
  scale_y_continuous(name = "ANC1",
                        breaks = seq(from = 0,
                                        to = max(dist1$anc1Std),
                                        by = 100)) +
  facet_grid(ADM1_EN ~ ADM2_EN) +
  themeSettings
           Gazelle District
                                 Kokopo District
                                                       Pomio District
                                                                            Rabaul District
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ggplot(dist1[dist1$ADM1_PCODE == 19, ], aes(as.Date(date), anc1Std)) +
  geom_line(colour = "#08519c", size = 1) +
  scale_x_date(name = "Month", date_breaks = "1 month", date_labels = "%b %y") +
  scale_y_continuous(name = "ANC1",
                        breaks = seq(from = 0,
                                        to = max(dist1$anc1Std),
                                        by = 100)) +
  facet_grid(ADM1_EN ~ ADM2_EN) +
```

```
Kandrian/Gloucester District
                                                                                                                                                                                                                                   Talasea District
    1700
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    1500
    1400
    1300
    1200
                                                                                                                                                                                                                                                                                                                           New Britain Province
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     1000
      900
      800
      700
      600
      500
      400
      300
                                                                                                                                                           Mouth

Feb 15:

Apr 15:

May 15:

Jun 15:

Aug 1
                                                                                          Jan 16
Feb 16
Mar 16
Apr 16
Jul 16
Jul 16
Sep 16
Sep 16
Nov 16
ggplot(dist1[dist1$ADM1_PCODE == 20, ], aes(as.Date(date), anc1Std)) +
        geom_line(colour = "#08519c", size = 1) +
       scale_x_date(name = "Month", date_breaks = "1 month", date_labels = "%b %y") +
        scale_y_continuous(name = "ANC1",
                                                                                     breaks = seq(from = 0,
                                                                                                                                           to = max(dist1$anc1Std),
                                                                                                                                          by = 100)) +
       facet_grid(ADM1_EN ~ ADM2_EN) +
       themeSettings
                                          Central Bougainville District
                                                                                                                                                North Bougainville District
                                                                                                                                                                                                                                                    South Bougainville District
    2200
    2100
2000
1900
1800
1700
1600
1400
1300
1200
1100
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700
600
500
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                                                                                                                                                                                                                           Jan 15

Reb 15

Mar 15

Apr 15

Jun 15

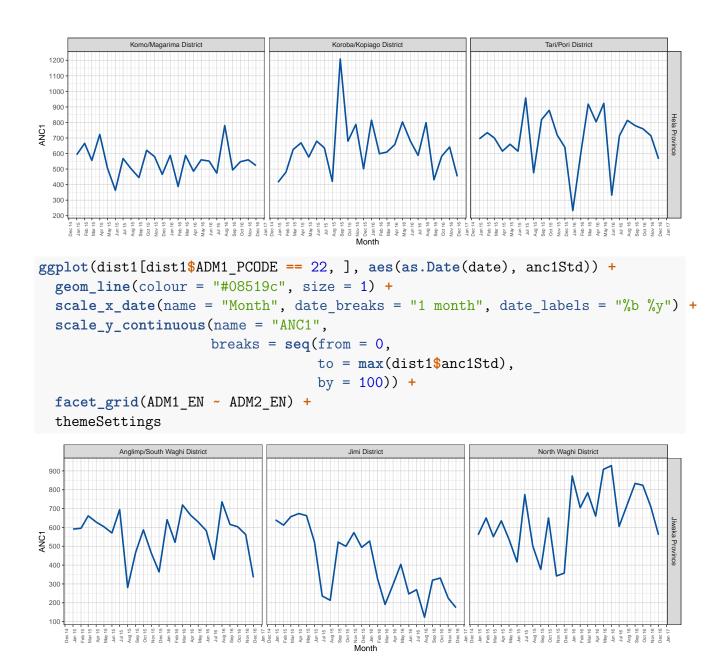
Jul 15

Sep 15

Sep 15

Nov 15
ggplot(dist1[dist1$ADM1_PCODE == 21, ], aes(as.Date(date), anc1Std)) +
       geom_line(colour = "#08519c", size = 1) +
        scale_x_date(name = "Month", date_breaks = "1 month", date_labels = "%b %y") +
        scale_y_continuous(name = "ANC1",
                                                                                     breaks = seq(from = 0,
                                                                                                                                           to = max(dist1\sanc1Std),
                                                                                                                                           by = 100)) +
       facet_grid(ADM1_EN ~ ADM2_EN) +
```

themeSettings

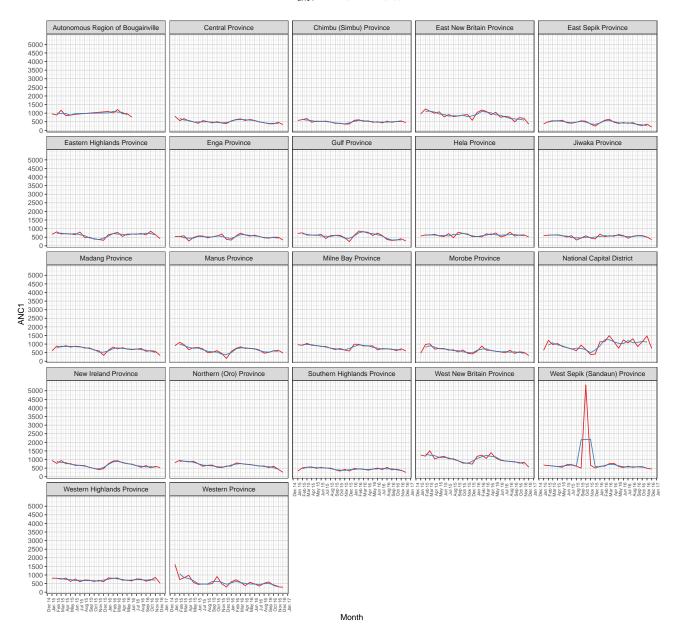


Smoothing time-series data

In the line charts above, we will notice that it is not very easy to see the trend of the indicators over time. Smoothing the time-series data is usually done to address this. Smoothing is usually performed using a rolling/running averages. This can be done in R using the function rollmean() from the zoo package and is implemented as shown below:

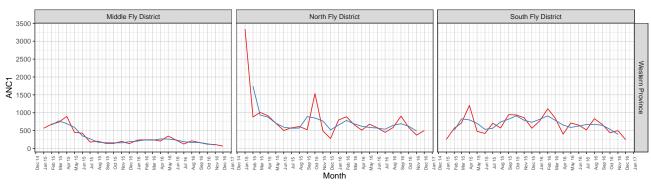
```
value = "value",
anc1Std, anc1Sm,
factor_key = TRUE)
```

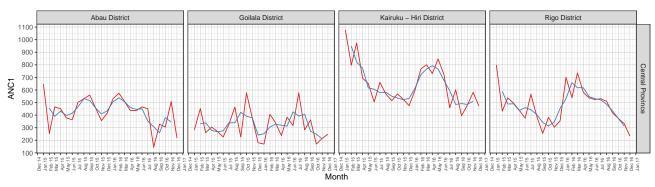
The smoothed data can then be plotted alongside the raw data as follows (for province data):



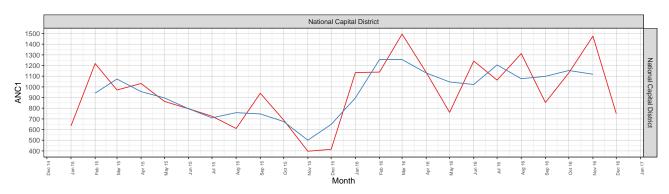
We can do the same for the district data.

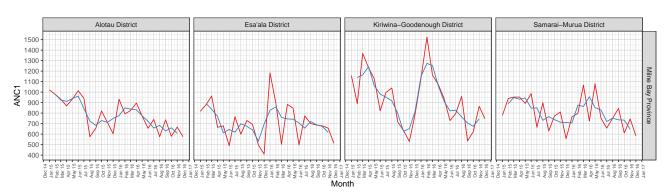
anc1 --- raw --- smooth



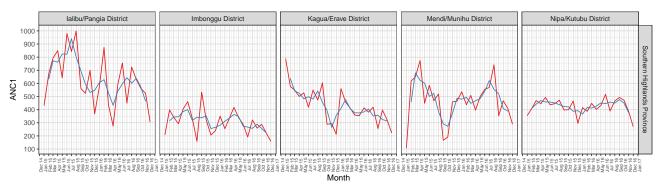


anc1 ·

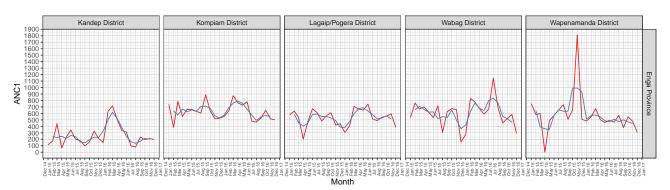


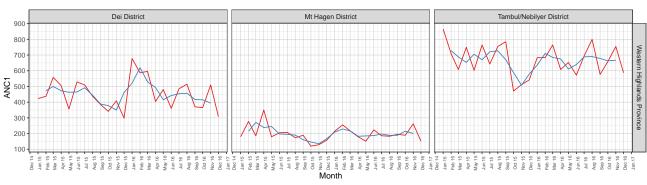


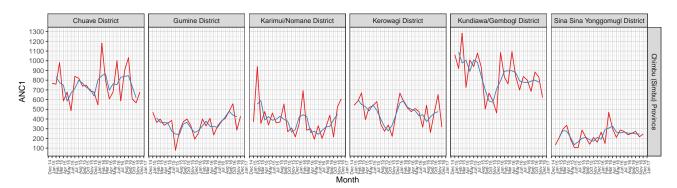
anc1 ·

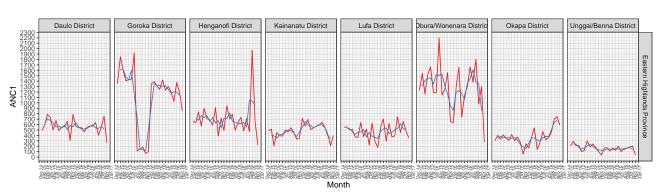


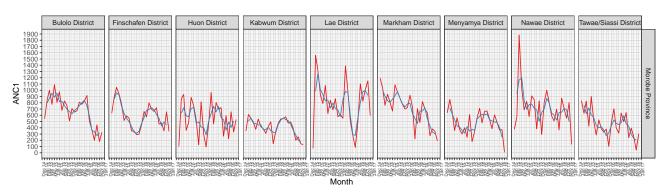
anc1

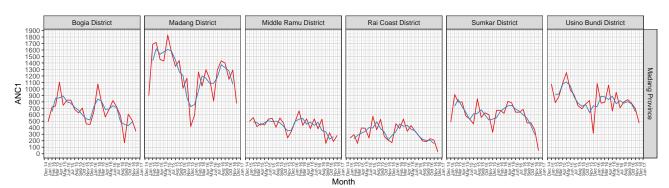


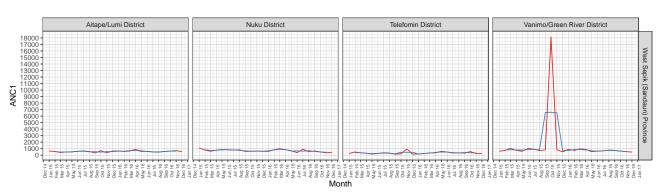


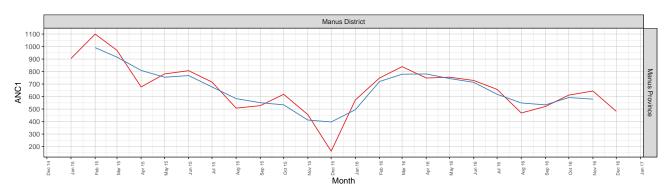


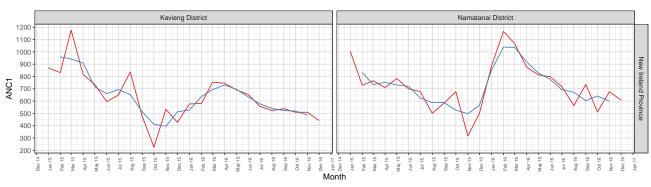








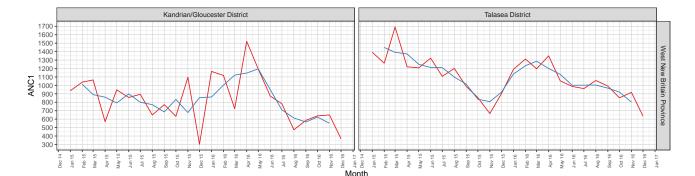




smooth

```
Gazelle District

Fomio Distri
```

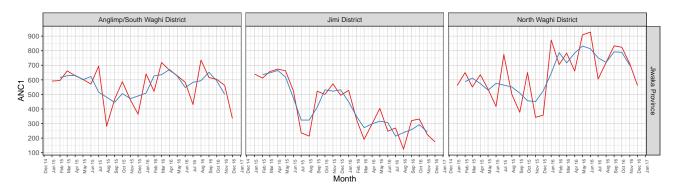


anc1 -

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anc1 ·



These approaches can be used for all the indicators.

3. Mapping of spatial distribution of indicators

The provincedata and the districtdata data frame objects are to be used for mapping. In addition, the province and district map data from the papuanewguinea package is for plotting the boundaries.

First, let us inspect the province and the district map objects.

	ADM1_EN	ADM1_PCODE	ADM0_EN	ADM0_PCODE
0	Autonomous Region of Bougainville	20	Papua New Guinea	PG
1	Central Province	03	Papua New Guinea	PG
2	Chimbu (Simbu) Province	10	Papua New Guinea	PG
3	East New Britain Province	18	Papua New Guinea	PG
4	East Sepik Province	14	Papua New Guinea	PG
5	Eastern Highlands Province	11	Papua New Guinea	PG
6	Enga Province	08	Papua New Guinea	PG
7	Gulf Province	02	Papua New Guinea	PG
8	Hela Province	21	Papua New Guinea	PG
9	Jiwaka Province	22	Papua New Guinea	PG
10	Madang Province	13	Papua New Guinea	PG
11	Manus Province	16	Papua New Guinea	PG
12	Milne Bay Province	05	Papua New Guinea	PG
13	Morobe Province	12	Papua New Guinea	PG
14	National Capital District	04	Papua New Guinea	PG
15	New Ireland Province	17	Papua New Guinea	PG
16	Northern (Oro) Province	06	Papua New Guinea	PG
17	Southern Highlands Province	07	Papua New Guinea	PG
18	West New Britain Province	19	Papua New Guinea	PG
19	West Sepik (Sandaun) Province	15	Papua New Guinea	PG
20	Western Highlands Province	09	Papua New Guinea	PG
21	Western Province	01	Papua New Guinea	PG

We notice that the order of the provinces are not sequential based on administrative code. We will need to re-order this in such a way that the admin codes are sequential. This can be done as follows:

```
province@data <- province@data[order(province@data$ADM1_PCODE), ]</pre>
```

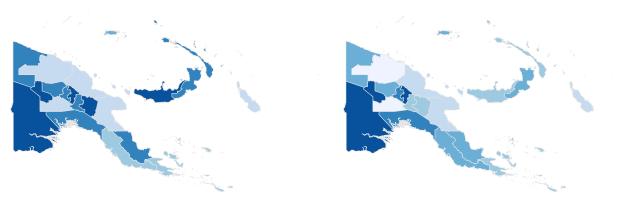
We can now map the provincedata. For this, we will use anc1 as our index indicator.

The first thing we need to do is to standardise the anc1 indicator in the same way we did earlier in the time-series analysis. This can be done as follows:

We then need to classify the standardised anc1 values into groups/classes that will allow us to colour the province based on their anc1 classification. We can use an approach in which the standardised anc1 counts are grouped into meaningful classes. A useful approach will be using quantiles. For this, we can use the R package called classInt which has a function called classIntervals(). The anc1 can be classified as follows:

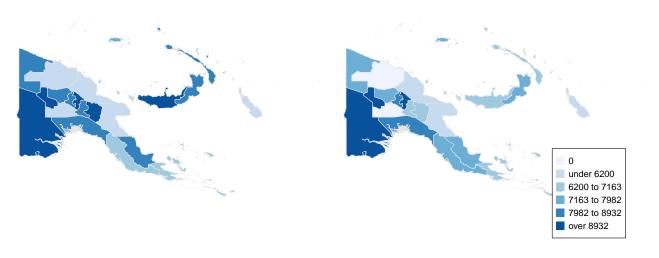
We can now map the anc1 indicator for year 2015 and year 2016 as follows:

```
col = colourscheme[anc1Province$class[anc1Province$year == 2016] + 1],
border = "gray90",
lwd = 0.5)
```



Now, it will be useful to add a title to each plot to identify which map is for which year and to add a legend to show what the colours refer to. This can be done as follows:

```
par(mar = c(0, 0, 0, 0), mfrow = c(1, 2))
plot(province,
     col = colourscheme[anc1Province$class[anc1Province$year == 2015] + 1],
     border = "gray90",
     1wd = 0.5)
title(main = "At least one antenatal care visits in 2015", line = -1, adj = 1)
plot(province,
     col = colourscheme[anc1Province$class[anc1Province$year == 2016] + 1],
     border = "gray90",
     1wd = 0.5)
title(main = "At least one antenatal care visits in 2016", line = -1, adj = 1)
legend(x = "bottomright",
       inset = 0.1,
       y.intersp = 1.2,
       legend = c("0", names(print(classIntervals(anc1Province$anc1Std,
                                                  n = 5,
                                                   style = "quantile",
                                                  dataPrecision = 0),
                                   between = "to",
                                   cutlabels = FALSE))),
       pch = 15, pt.cex = 2,
       col = colourscheme)
```



style: quantile
one of 12,650 possible partitions of this variable into 5 classes
under 6200 6200 to 7163 7163 to 7982 7982 to 8932 over 8932
5 5 5 6