

Oman Road Traffic Accidents (RTA) Analysis in R

Using Time Series

1. Data Extraction

```
In [1]: # @author: Aamir M. Khan
# Created First: April 2 2019
# Updated Last:
```

Import Libraries

```
In [2]: library(reshape2)
library(ggplot2)
library(tidyverse)

-- Attaching packages ----- tidyverse 1.2.1 --
v tibble 2.1.1      v purrr  0.3.2
v tidyr  0.8.3      v dplyr  0.8.1
v readr  1.3.1      v stringr 1.4.0
v tibble 2.1.1      v forcats 0.4.0
-- Conflicts ----- tidyverse_conflicts() --
x dplyr::filter() masks stats::filter()
x dplyr::lag()    masks stats::lag()
```

Load Data Files

```
In [3]: data <- read.csv("../data/fatalities.csv")
head(data,3)
```

Year	January	February	March	April	May	June	July	August	September	October	November	December
2000	56	38	49	38	31	43	61	24	23	52	32	45
2001	40	35	41	27	36	47	40	40	47	60	28	58
2002	46	56	30	86	38	59	42	55	35	45	45	43

```
In [4]: dim(data)
```

19 13

```
In [5]: class(data)
```

'data.frame'

Formatting the Data

```
In [6]: fatalities = melt(data, id='Year', variable.name='Month', value.name='Value')
head(fatalities,5)
```

Year	Month	Value
2000	January	56
2001	January	40
2002	January	46
2003	January	70
2004	January	45

```
In [7]: fatalities$Month <- as.integer(factor(fatalities$Month, levels = unique(fatalities$Month)))
fatalities = na.omit(fatalities)
fatalities$Time <- as.Date(sprintf("%d-%02d-%02d", fatalities$Year, fatalities$Month,1))
fatalities <- fatalities[c('Time','Value')]
fatalities <- fatalities[order(fatalities$Time),]
rownames(fatalities) <- NULL
head(fatalities,5)
```

Time	Value
2000-01-01	56
2000-02-01	38
2000-03-01	49
2000-04-01	38
2000-05-01	31

```
In [8]: fatalities$Year <- format(fatalities$Time, "%Y")
```

```
In [9]: str(fatalities)
```

```
'data.frame':  228 obs. of  3 variables:
 $ Time : Date, format: "2000-01-01" "2000-02-01" ...
 $ Value: int  56 38 49 38 31 43 61 24 23 52 ...
 $ Year  : chr  "2000" "2000" "2000" "2000" ...
```

```
In [10]: seq(2000, 2018, by=3)
```

2000 2003 2006 2009 2012 2015 2018

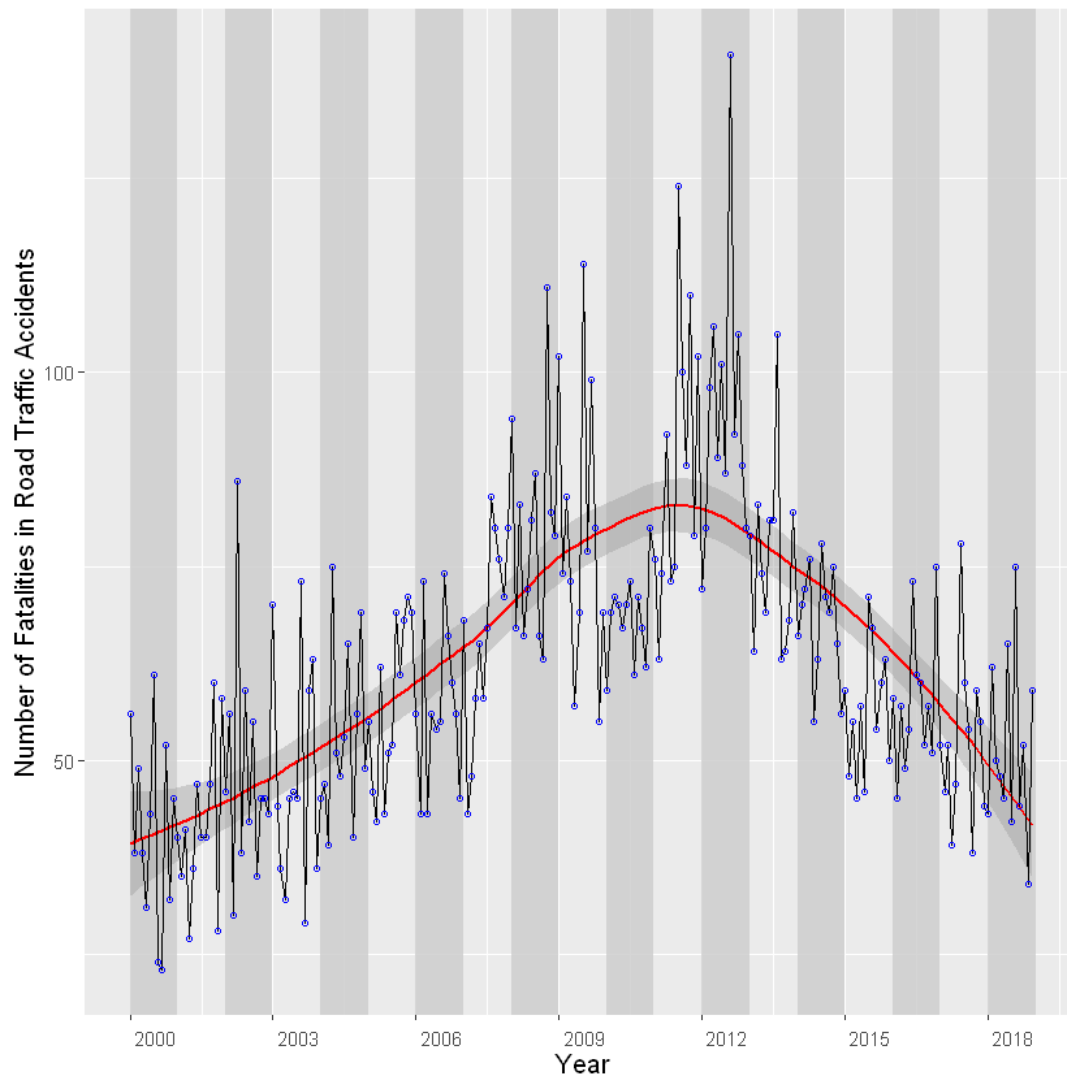
Time Series of Road Traffic Accidents

```

In [15]: xmin <- seq(2000,2018,2)
xlb11 <- seq(as.Date("2000-01-01"), as.Date("2018-12-01"), by="3 years")
xlb12 <- seq(2000, 2018, by=3)
p <- ggplot() +
  geom_rect(data=data.frame(xmin=xmin), aes(xmin=as.Date(paste0(xmin,"-01-01")),
                                           xmax=as.Date(paste0(xmin,"-12-31")),
                                           ymin=-Inf, ymax=+Inf),
            fill='gray80', alpha=0.8) +
  geom_smooth(data=fatalities, aes(x=Time, y=Value), col='red', size=0.7) +
  geom_line(data=fatalities, aes(x=Time, y=Value), size=0.5) +
  geom_point(data=fatalities, aes(x=Time, y=Value), shape=1, col='blue', size=1) +
  #ggtitle("Road Traffic Deaths in Oman from 2000 to 2018") +
  labs(x='Year', y='Number of Fatalities in Road Traffic Accidents') +
  scale_x_date(breaks = as.Date(xlb11), labels = xlb12) +
  theme_grey() +
  theme(
    #plot.title = element_text(color="blue", size=14, face="bold.italic", hjust = 0.5),
    axis.title.x = element_text(size=12),
    axis.text.x = element_text(lineheight = 0, angle = 0, vjust=0, hjust=-0.1),
    axis.title.y = element_text(size=12)
  )
p

```

`geom_smooth()` using method = 'loess' and formula 'y ~ x'



```

In [16]: ggsave('Fatalities_Ts.png', p, device='png', dpi=1200, limitsize=FALSE)

```

Saving 6.67 x 6.67 in image
 `geom_smooth()` using method = 'loess' and formula 'y ~ x'

Decomposition of Time Series Data

```
In [17]: # transpose and convert to vector form
        data_v <- as.vector(t(data[, -1]))
        data_v

56 38 49 38 31 43 61 24 23 52 32 45 40 35 41 27 36 47 40 40 47 60 28 58 46 56 30 86 38 59 42 55
35 45 45 43 70 44 36 32 45 46 45 73 29 59 63 36 45 47 39 75 51 48 53 65 40 56 69 49 55 46 42 62
43 51 52 69 61 68 71 69 56 43 73 43 56 54 55 74 66 60 56 45 68 43 48 58 65 58 67 84 80 76 71 80
94 67 83 66 72 81 87 66 63 111 82 79 102 74 84 73 57 69 114 77 99 80 55 69 59 69 71 70 67 70 73 61
71 67 62 80 76 63 74 92 73 75 124 100 88 110 79 102 72 80 98 106 89 101 87 141 92 105 88 80 79 64
83 74 69 81 81 105 63 64 68 82 66 70 72 76 55 63 78 71 69 75 65 56 59 48 55 45 57 46 71 67 54 60
63 50 58 45 57 49 54 73 61 60 52 57 51 75 52 46 52 39 47 78 60 54 38 59 55 44 43 62 50 48 45 65
42 75 44 52 34 59
```

```
In [18]: # create time series from data
        data_ts <- ts(data_v, frequency=12, start=c(2000,1))
```

```
In [19]: summary(data_ts)

      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
23.00   47.00   60.50   62.33   73.00  141.00
```

```
In [20]: print(data_ts)

      Jan Feb Mar Apr May Jun Jul Aug Sep Oct Nov Dec
2000  56  38  49  38  31  43  61  24  23  52  32  45
2001  40  35  41  27  36  47  40  40  47  60  28  58
2002  46  56  30  86  38  59  42  55  35  45  45  43
2003  70  44  36  32  45  46  45  73  29  59  63  36
2004  45  47  39  75  51  48  53  65  40  56  69  49
2005  55  46  42  62  43  51  52  69  61  68  71  69
2006  56  43  73  43  56  54  55  74  66  60  56  45
2007  68  43  48  58  65  58  67  84  80  76  71  80
2008  94  67  83  66  72  81  87  66  63 111  82  79
2009 102  74  84  73  57  69 114  77  99  80  55  69
2010  59  69  71  70  67  70  73  61  71  67  62  80
2011  76  63  74  92  73  75 124 100  88 110  79 102
2012  72  80  98 106  89 101  87 141  92 105  88  80
2013  79  64  83  74  69  81  81 105  63  64  68  82
2014  66  70  72  76  55  63  78  71  69  75  65  56
2015  59  48  55  45  57  46  71  67  54  60  63  50
2016  58  45  57  49  54  73  61  60  52  57  51  75
2017  52  46  52  39  47  78  60  54  38  59  55  44
2018  43  62  50  48  45  65  42  75  44  52  34  59
```

```
In [21]: # decompose seasonal time series data: original, seasonal, trend, random  
fitF <- decompose(data_ts)  
fitF
```

	Jan	Feb	Mar	Apr	May	Jun	Jul	Aug	Sep	Oct	Nov	Dec
2000	56	38	49	38	31	43	61	24	23	52	32	45
2001	40	35	41	27	36	47	40	40	47	60	28	58
2002	46	56	30	86	38	59	42	55	35	45	45	43
2003	70	44	36	32	45	46	45	73	29	59	63	36
2004	45	47	39	75	51	48	53	65	40	56	69	49
2005	55	46	42	62	43	51	52	69	61	68	71	69
2006	56	43	73	43	56	54	55	74	66	60	56	45
2007	68	43	48	58	65	58	67	84	80	76	71	80
2008	94	67	83	66	72	81	87	66	63	111	82	79
2009	102	74	84	73	57	69	114	77	99	80	55	69
2010	59	69	71	70	67	70	73	61	71	67	62	80
2011	76	63	74	92	73	75	124	100	88	110	79	102
2012	72	80	98	106	89	101	87	141	92	105	88	80
2013	79	64	83	74	69	81	81	105	63	64	68	82
2014	66	70	72	76	55	63	78	71	69	75	65	56
2015	59	48	55	45	57	46	71	67	54	60	63	50
2016	58	45	57	49	54	73	61	60	52	57	51	75
2017	52	46	52	39	47	78	60	54	38	59	55	44
2018	43	62	50	48	45	65	42	75	44	52	34	59
\$seasonal												
	Jan	Feb	Mar	Apr	May	Jun						
2000	0.1618441	-7.5788966	-2.9677855	-1.1830633	-6.8543596	1.2197145						
2001	0.1618441	-7.5788966	-2.9677855	-1.1830633	-6.8543596	1.2197145						
2002	0.1618441	-7.5788966	-2.9677855	-1.1830633	-6.8543596	1.2197145						
2003	0.1618441	-7.5788966	-2.9677855	-1.1830633	-6.8543596	1.2197145						
2004	0.1618441	-7.5788966	-2.9677855	-1.1830633	-6.8543596	1.2197145						
2005	0.1618441	-7.5788966	-2.9677855	-1.1830633	-6.8543596	1.2197145						
2006	0.1618441	-7.5788966	-2.9677855	-1.1830633	-6.8543596	1.2197145						
2007	0.1618441	-7.5788966	-2.9677855	-1.1830633	-6.8543596	1.2197145						
2008	0.1618441	-7.5788966	-2.9677855	-1.1830633	-6.8543596	1.2197145						
2009	0.1618441	-7.5788966	-2.9677855	-1.1830633	-6.8543596	1.2197145						
2010	0.1618441	-7.5788966	-2.9677855	-1.1830633	-6.8543596	1.2197145						
2011	0.1618441	-7.5788966	-2.9677855	-1.1830633	-6.8543596	1.2197145						
2012	0.1618441	-7.5788966	-2.9677855	-1.1830633	-6.8543596	1.2197145						
2013	0.1618441	-7.5788966	-2.9677855	-1.1830633	-6.8543596	1.2197145						
2014	0.1618441	-7.5788966	-2.9677855	-1.1830633	-6.8543596	1.2197145						
2015	0.1618441	-7.5788966	-2.9677855	-1.1830633	-6.8543596	1.2197145						
2016	0.1618441	-7.5788966	-2.9677855	-1.1830633	-6.8543596	1.2197145						
2017	0.1618441	-7.5788966	-2.9677855	-1.1830633	-6.8543596	1.2197145						
2018	0.1618441	-7.5788966	-2.9677855	-1.1830633	-6.8543596	1.2197145						
	Jul	Aug	Sep	Oct	Nov	Dec						
2000	6.5831404	8.5021219	-3.5557485	7.1965664	-1.8034336	0.2798997						
2001	6.5831404	8.5021219	-3.5557485	7.1965664	-1.8034336	0.2798997						
2002	6.5831404	8.5021219	-3.5557485	7.1965664	-1.8034336	0.2798997						
2003	6.5831404	8.5021219	-3.5557485	7.1965664	-1.8034336	0.2798997						
2004	6.5831404	8.5021219	-3.5557485	7.1965664	-1.8034336	0.2798997						
2005	6.5831404	8.5021219	-3.5557485	7.1965664	-1.8034336	0.2798997						
2006	6.5831404	8.5021219	-3.5557485	7.1965664	-1.8034336	0.2798997						
2007	6.5831404	8.5021219	-3.5557485	7.1965664	-1.8034336	0.2798997						
2008	6.5831404	8.5021219	-3.5557485	7.1965664	-1.8034336	0.2798997						
2009	6.5831404	8.5021219	-3.5557485	7.1965664	-1.8034336	0.2798997						
2010	6.5831404	8.5021219	-3.5557485	7.1965664	-1.8034336	0.2798997						
2011	6.5831404	8.5021219	-3.5557485	7.1965664	-1.8034336	0.2798997						
2012	6.5831404	8.5021219	-3.5557485	7.1965664	-1.8034336	0.2798997						
2013	6.5831404	8.5021219	-3.5557485	7.1965664	-1.8034336	0.2798997						
2014	6.5831404	8.5021219	-3.5557485	7.1965664	-1.8034336	0.2798997						
2015	6.5831404	8.5021219	-3.5557485	7.1965664	-1.8034336	0.2798997						
2016	6.5831404	8.5021219	-3.5557485	7.1965664	-1.8034336	0.2798997						
2017	6.5831404	8.5021219	-3.5557485	7.1965664	-1.8034336	0.2798997						
2018	6.5831404	8.5021219	-3.5557485	7.1965664	-1.8034336	0.2798997						
\$trend												
	Jan	Feb	Mar	Apr	May	Jun	Jul	Aug				
2000	NA	NA	NA	NA	NA	NA	40.33333	39.54167				
2001	37.70833	37.50000	39.16667	40.50000	40.66667	41.04167	41.83333	42.95833				
2002	49.08333	49.79167	49.91667	48.79167	48.87500	48.95833	49.33333	49.83333				
2003	44.95833	45.83333	46.33333	46.66667	48.00000	48.45833	47.12500	46.20833				
2004	51.16667	51.16667	51.29167	51.62500	51.75000	52.54167	53.50000	53.87500				
2005	52.54167	52.66667	53.70833	55.08333	55.66667	56.58333	57.45833	57.37500				
2006	59.70833	60.04167	60.45833	60.33333	59.37500	57.75000	57.25000	57.75000				
2007	58.50000	59.41667	60.41667	61.66667	62.95833	65.04167	67.58333	69.66667				
2008	77.58333	77.66667	76.20833	76.95833	78.87500	79.29167	79.58333	80.20833				
2009	80.04167	81.62500	83.58333	83.79167	81.37500	79.83333	77.62500	75.62500				
2010	73.29167	70.91667	69.08333	67.37500	67.12500	67.87500	69.04167	69.50000				
2011	74.37500	78.12500	80.45833	82.95833	85.45833	87.08333	87.83333	88.37500				
2012	94.20833	94.37500	96.25000	96.20833	96.37500	95.83333	95.20833	94.83333				
2013	86.66667	84.91667	82.20833	79.29167	76.75000	76.00000	75.54167	75.25000				
2014	71.95833	70.41667	69.25000	69.95833	70.29167	69.08333	67.70833	66.50000				
2015	60.04167	59.58333	58.79167	57.54167	56.83333	56.50000	56.20833	56.04167				
2016	58.00000	57.29167	56.91667	56.70833	56.08333	56.62500	57.41667	57.20833				
2017	55.79167	55.50000	54.66667	54.16667	54.41667	53.29167	51.62500	51.91667				
2018	51.16667	51.29167	52.41667	52.37500	51.20833	50.95833	NA	NA				
	Sep	Oct	Nov	Dec								
2000	39.08333	38.29167	38.04167	38.41667								
2001	43.37500	45.37500	47.91667	48.50000								
2002	49.58333	47.58333	45.62500	45.37500								
2003	46.45833	48.37500	50.41667	50.75000								
2004	53.95833	53.54167	52.66667	52.45833								
2005	58.54167	59.04167	58.79167	59.45833								
2006	56.70833	56.29167	57.29167	57.83333								
2007	72.12500	73.91667	74.54167	75.79167								

```

2008 80.54167 80.87500 80.54167 79.41667
2009 74.87500 74.20833 74.50000 74.95833
2010 69.37500 70.41667 71.58333 72.04167
2011 90.08333 91.66667 92.91667 94.66667
2012 93.54167 91.58333 89.41667 87.75000
2013 75.04167 74.66667 74.16667 72.83333
2014 64.87500 62.87500 61.66667 61.04167
2015 56.00000 56.25000 56.29167 57.29167
2016 57.04167 56.41667 55.70833 55.62500
2017 52.50000 52.79167 53.08333 52.45833
2018 NA NA NA NA

```

```
$random
```

```

      Jan      Feb      Mar      Apr      May
2000 NA      NA      NA      NA      NA
2001 2.12982253 5.07889660 4.80111883 -12.31693673 2.18769290
2002 -3.24517747 13.78722994 -16.94888117 38.39139660 -4.02064043
2003 24.87982253 5.74556327 -7.36554784 -13.48360340 3.85435957
2004 -6.32851080 3.41222994 -9.32388117 24.55806327 6.10435957
2005 2.29648920 0.91222994 -8.74054784 8.09972994 -5.81230710
2006 -3.87017747 -9.46277006 15.50945216 -16.15027006 3.47935957
2007 9.33815586 -8.83777006 -9.44888117 -2.48360340 8.89602623
2008 16.25482253 -3.08777006 9.75945216 -9.77527006 -0.02064043
2009 21.79648920 -0.04610340 3.38445216 -9.60860340 -17.52064043
2010 -14.45351080 5.66222994 4.88445216 3.80806327 6.72935957
2011 1.46315586 -7.54610340 -3.49054784 10.22472994 -5.60397377
2012 -22.37017747 -6.79610340 4.71778549 10.97472994 -0.52064043
2013 -7.82851080 -13.33777006 3.75945216 -4.10860340 -0.89564043
2014 -6.12017747 7.16222994 5.71778549 7.22472994 -8.43730710
2015 -1.20351080 -4.00443673 -0.82388117 -11.35860340 7.02102623
2016 -0.16184414 -4.71277006 3.05111883 -6.52527006 4.77102623
2017 -3.95351080 -1.92110340 0.30111883 -13.98360340 -0.56230710
2018 -8.32851080 18.28722994 0.55111883 -3.19193673 0.64602623

      Jun      Jul      Aug      Sep      Oct
2000 NA      14.08352623 -24.04378858 -12.52758488 6.51176698
2001 4.73861883 -8.41647377 -11.46045525 7.18074846 7.42843364
2002 8.82195216 -13.91647377 -3.33545525 -11.02758488 -9.77989969
2003 -3.67804784 -8.70814043 18.28954475 -13.90258488 3.42843364
2004 -5.76138117 -7.08314043 2.62287809 -10.40258488 -4.73823302
2005 -6.80304784 -12.04147377 3.12287809 6.01408179 1.76176698
2006 -4.96971451 -8.83314043 7.74787809 12.84741512 -3.48823302
2007 -8.26138117 -7.16647377 5.83121142 11.43074846 -5.11323302
2008 0.48861883 0.83352623 -22.71045525 -13.98591821 22.92843364
2009 -12.05304784 29.79185957 -7.12712191 27.68074846 -1.40489969
2010 0.90528549 -2.62480710 -17.00212191 5.18074846 -10.61323302
2011 -13.30304784 29.58352623 3.12287809 1.47241512 11.13676698
2012 3.94695216 -14.79147377 37.66454475 2.01408179 6.22010031
2013 3.78028549 -1.12480710 21.24787809 -8.48591821 -17.86323302
2014 -7.30304784 3.70852623 -4.00212191 7.68074846 4.92843364
2015 -11.71971451 8.20852623 2.45621142 1.55574846 -3.44656636
2016 15.15528549 -2.99980710 -5.71045525 -1.48591821 -6.61323302
2017 23.48861883 1.79185957 -6.41878858 -10.94425154 -0.98823302
2018 12.82195216 NA      NA      NA      NA

      Nov      Dec
2000 -4.23823302 6.30343364
2001 -18.11323302 9.22010031
2002 1.17843364 -2.65489969
2003 14.38676698 -15.02989969
2004 18.13676698 -3.73823302
2005 14.01176698 9.26176698
2006 0.51176698 -13.11323302
2007 -1.73823302 3.92843364
2008 3.26176698 -0.69656636
2009 -17.69656636 -6.23823302
2010 -7.77989969 7.67843364
2011 -12.11323302 7.05343364
2012 0.38676698 -8.02989969
2013 -4.36323302 8.88676698
2014 5.13676698 -5.32156636
2015 8.51176698 -7.57156636
2016 -2.90489969 19.09510031
2017 3.72010031 -8.73823302
2018 NA      NA

```

```
$figure
```

```

[1] 0.1618441 -7.5788966 -2.9677855 -1.1830633 -6.8543596 1.2197145
[7] 6.5831404 8.5021219 -3.5557485 7.1965664 -1.8034336 0.2798997

```

```
$type
```

```
[1] "additive"
```

```
attr("class")
```

```
[1] "decomposed.ts"
```

```

In [22]: Time = attributes(data_ts)[[1]]
         Time = seq(Time[1],Time[2], length.out=(Time[2]-Time[1])*Time[3])

```

```

In [25]: # convert to data frame
         dat1 = cbind(Time, with(fitF, data.frame(Observed=x, Trend=trend, Seasonal=seasonal, Random=random)))
         dat2 = gather(dat1, component, value, -Time)

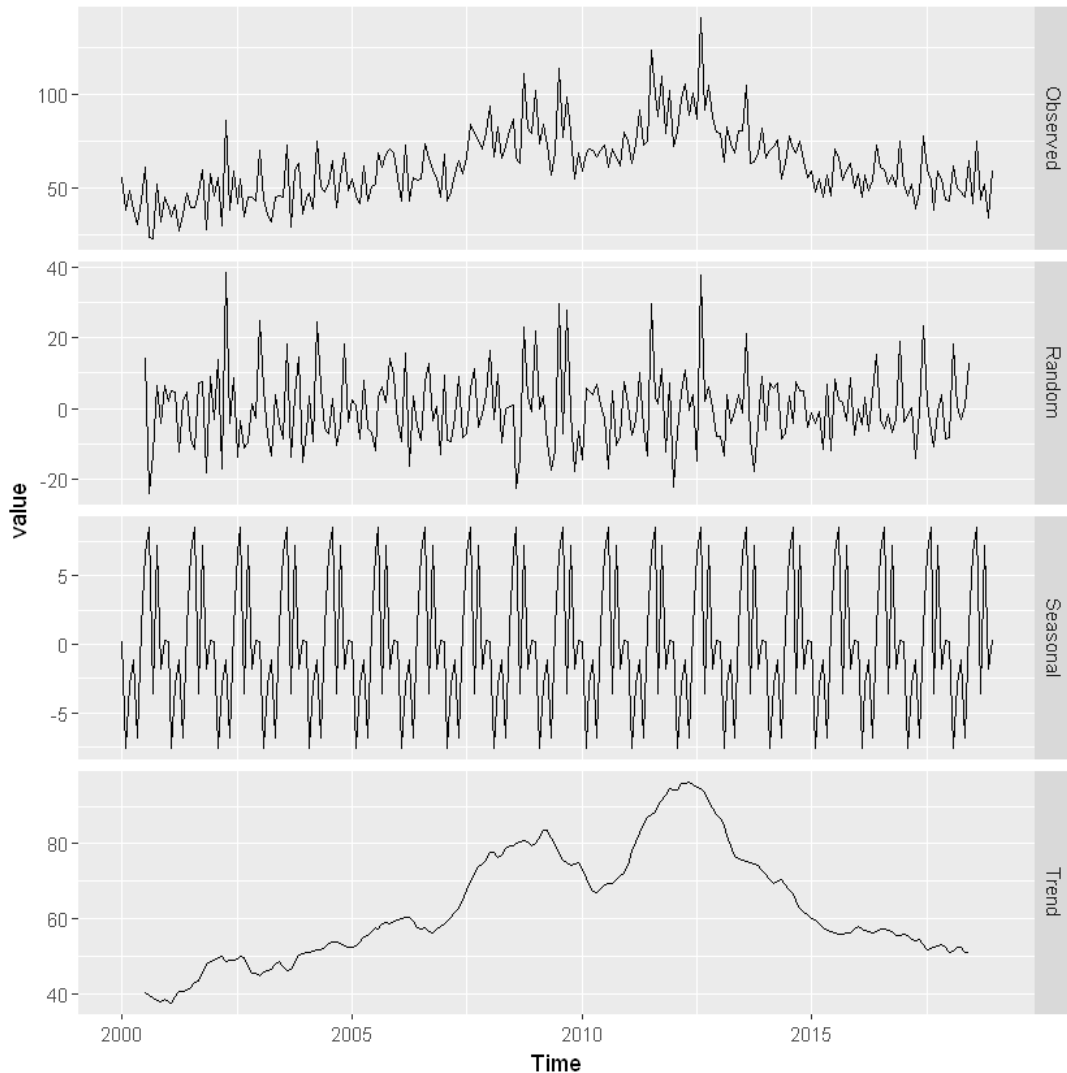
```

```
In [26]: # plot decomposed time series frame
q <- ggplot(dat2, aes(Time, value)) +
  facet_grid(component ~ ., scales="free_y") +
  geom_line(size=0.5) ##
# #labs(y=expression(Fatalities~of~Accidents), x="Time (Years)") +
# #labs(x='Year', y='Number of Fatalities in Road Traffic Accidents') +
# #ggtitle(expression(Decomposed~RTA~time~series)) +
# #theme_bw() +
# #theme(plot.title=element_text(hjust=0.5))
# theme_grey() +
# theme(
# #plot.title = element_text(color="blue", size=14, face="bold.italic", hjust = 0.5),
# #axis.title.x = element_text(size=12),
# #axis.text.x = element_text(lineheight = 0, angle = 0, vjust=0, hjust=-0.1),
# #axis.title.y = element_text(size=12)
# )
q
```

Don't know how to automatically pick scale for object of type ts. Defaulting to continuous.

Warning message:

"Removed 6 rows containing missing values (geom_path)."



```
In [27]: ggsave('Decomposed.png', q, device='png', dpi=1200, limitsize=FALSE)
```

Saving 6.67 x 6.67 in image

Don't know how to automatically pick scale for object of type ts. Defaulting to continuous.

Warning message:

"Removed 6 rows containing missing values (geom_path)."

In []:


```
In [469]: xmin <- seq(2000,2018,2)
ggplot() +
  facet_wrap(gather(dat, component, value, -Time), aes(Time, value), component ~ ., scales="free_y") +
  geom_line(gather(dat, component, value, -Time), aes(Time, value), col='blue', size=0.5) +
  #Labs(y=expression(Fatalities~of~Accidents), x="Time (Years)") +
  labs(x='Year', y='Number of Fatalities in Road Traffic Accidents') +
#   ggtitle(expression(Decomposed~RTA~time~series)) +
#   theme_bw() +
#   theme(plot.title=element_text(hjust=0.5))
theme_grey() +
theme(
  #plot.title = element_text(color="blue", size=14, face="bold.italic",hjust = 0.5),
  axis.title.x = element_text(size=12),
  axis.text.x = element_text(lineheight = 0, angle = 0, vjust=0, hjust=-0.1),
  axis.title.y = element_text(size=12)
)
```

Warning message:
"Only the first value of `nrow` will be used."
Warning message:
"Coercing `nrow` to be an integer."

Error in sanitise_dim(nrow): (list) object cannot be coerced to type 'integer'
Traceback:

1. facet_wrap(gather(dat, component, value, -Time), aes(Time, value),
 . component ~ ., scales = "free_y")
2. sanitise_dim(nrow)

```
In [ ]:
In [ ]:
In [ ]:
In [ ]:
In [ ]:
```

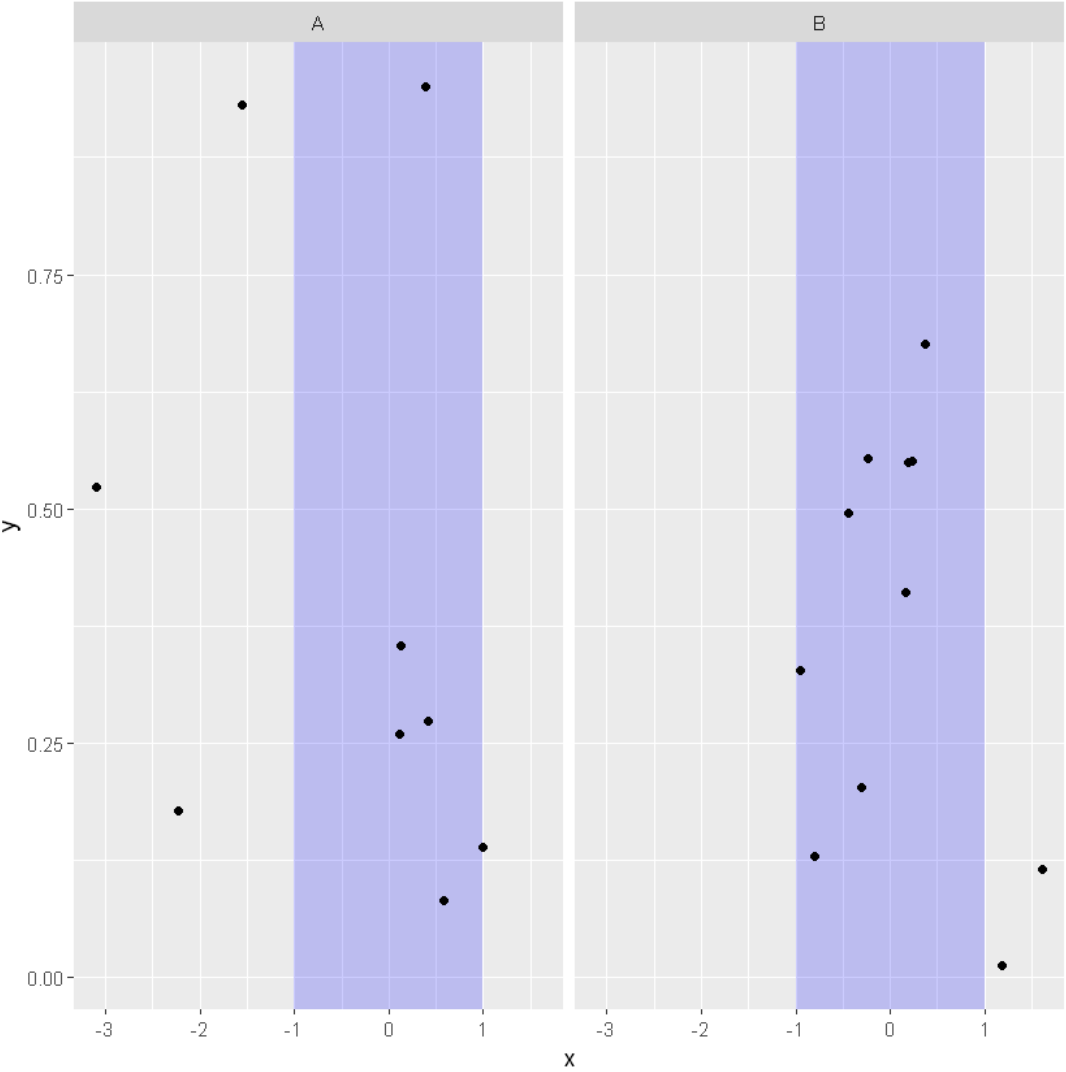
```
In [450]: df <- data.frame(x = rnorm(20),
                        y = runif(20),
                        facet = sample(c("A", "B"),
                                     20,
                                     replace = TRUE))
df
```

x	y	facet
-0.357615902	0.08661050	A
-2.249300745	0.90498928	A
0.002837861	0.19894589	B
1.283033216	0.97319802	A
-0.267299063	0.06459416	A
-0.717604576	0.45351959	B
-0.604157249	0.27298650	B
-1.394921621	0.22182555	B
-1.598624461	0.02343691	B
1.605976503	0.75214399	B
0.625106883	0.87449888	A
-0.501474286	0.02234960	B
-1.555226410	0.44241519	A
-0.157451651	0.46291342	A
-0.289198126	0.97845921	A
0.139514194	0.39979908	A
0.812509093	0.32213330	A
0.422366586	0.24405862	A
0.109740932	0.67612420	A
0.652817236	0.45258115	B

```
In [451]: rect1 <- data.frame(xmin = -1,
                        xmax = 1,
                        ymin = -Inf,
                        ymax = Inf,
                        facet = c("A", "B"))
rect1
```

xmin	xmax	ymin	ymax	facet
-1	1	-Inf	Inf	A
-1	1	-Inf	Inf	B

```
In [449]: ggplot() +  
  geom_rect(data = rect1 , aes(xmin = xmin,  
                               xmax = xmax,  
                               ymin = ymin,  
                               ymax = ymax),  
            alpha = 0.2, fill = "blue") +  
  geom_point(data = df, aes(x = x, y = y))+  
  facet_wrap(~facet)
```



```
In [ ]:   
  
In [ ]:   
  
In [ ]: 
```

```
In [458]: gather(dat, component, value, -Time)
```

Time	component	value
2000.000	Observed	56
2000.083	Observed	38
2000.167	Observed	49
2000.250	Observed	38
2000.333	Observed	31
2000.417	Observed	43
2000.500	Observed	61
2000.583	Observed	24
2000.667	Observed	23
2000.750	Observed	52
2000.833	Observed	32
2000.917	Observed	45
2001.000	Observed	40
2001.083	Observed	35
2001.167	Observed	41
2001.250	Observed	27
2001.333	Observed	36
2001.417	Observed	47
2001.500	Observed	40
2001.583	Observed	40
2001.667	Observed	47
2001.750	Observed	60
2001.833	Observed	28
2001.917	Observed	58
2002.000	Observed	46
2002.083	Observed	56
2002.167	Observed	30
2002.250	Observed	86
2002.333	Observed	38
2002.417	Observed	59
...
2016.500	Random	-2.9998071
2016.583	Random	-5.7104552
2016.667	Random	-1.4859182
2016.750	Random	-6.6132330
2016.833	Random	-2.9048997
2016.917	Random	19.0951003
2017.000	Random	-3.9535108
2017.083	Random	-1.9211034
2017.167	Random	0.3011188
2017.250	Random	-13.9836034
2017.333	Random	-0.5623071
2017.417	Random	23.4886188
2017.500	Random	1.7918596
2017.583	Random	-6.4187886
2017.667	Random	-10.9442515
2017.750	Random	-0.9882330
2017.833	Random	3.7201003
2017.917	Random	-8.7382330
2018.000	Random	-8.3285108
2018.083	Random	18.2872299
2018.167	Random	0.5511188
2018.250	Random	-3.1919367
2018.333	Random	0.6460262
2018.417	Random	12.8219522
2018.500	Random	NA
2018.583	Random	NA
2018.667	Random	NA
2018.750	Random	NA
2018.833	Random	NA

