

Week 10: Temporal data

27/03/23

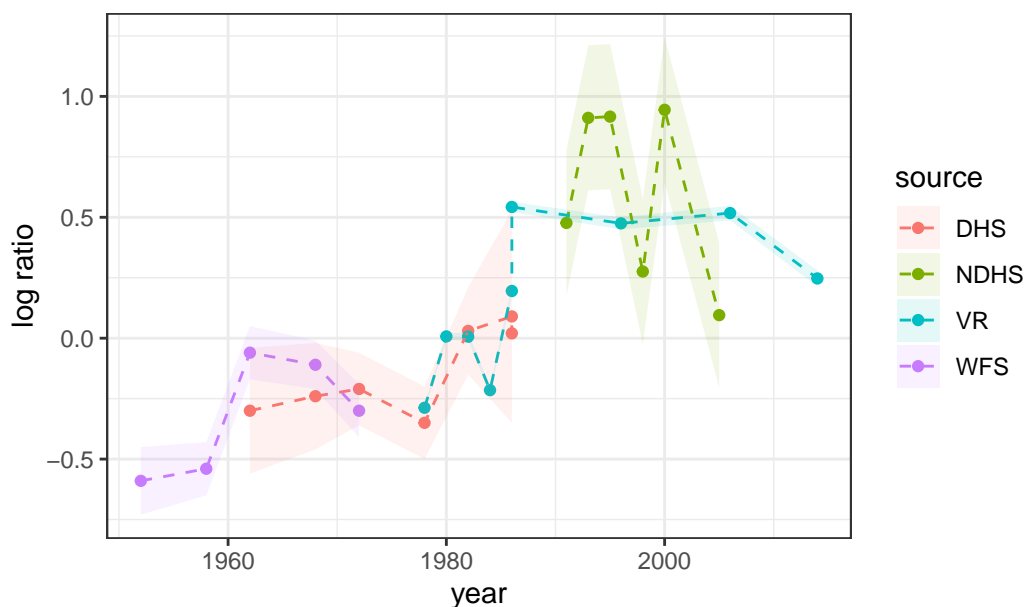
Child mortality in Sri Lanka

In this lab you will be fitting a couple of different models to the data about child mortality in Sri Lanka, which was used in the lecture. Here's the data and the plot from the lecture:

```
library(tidyverse)
library(here)
library(rstan)
library(tidybayes)

lka <- read_csv("lka.csv")
ggplot(lka, aes(year, logit_ratio)) +
  geom_point(aes( color = source)) +
  geom_line(aes( color = source), lty = 2) +
  geom_ribbon(aes(ymin = logit_ratio - se,
                 ymax = logit_ratio + se,
                 fill = source), alpha = 0.1) +
  theme_bw()+
  labs(title = "Ratio of neonatal to other child mortality (logged), Sri Lanka", y = "log
```

Ratio of neonatal to other child mortality (logged), Sri Lanka



Fitting a linear model

Let's firstly fit a linear model in time to these data. Here's the code to do this:

```
observed_years <- lka$year
years <- min(observed_years):max(observed_years)
nyears <- length(years)

stan_data <- list(y = lka$logit_ratio, year_i = observed_years - years[1]+1,
                 T = nyears, years = years, N = length(observed_years),
                 mid_year = mean(years), se = lka$se)

mod <- stan(data = stan_data,
            file = "lka_linear_me.stan",
            refresh = 0)
```

Running /Library/Frameworks/R.framework/Resources/bin/R CMD SHLIB foo.c

clang -arch arm64 -I"/Library/Frameworks/R.framework/Resources/include" -DNDEBUG -I"/Library/Frameworks/R.framework/Versions/4.2-arm64/Resources/library/S

In file included from <built-in>:1:

In file included from /Library/Frameworks/R.framework/Versions/4.2-arm64/Resources/library/S

In file included from /Library/Frameworks/R.framework/Versions/4.2-arm64/Resources/library/R

```

In file included from /Library/Frameworks/R.framework/Versions/4.2-arm64/Resources/library/R
/Library/Frameworks/R.framework/Versions/4.2-arm64/Resources/library/RcppEigen/include/Eigen,
namespace Eigen {
~

/Library/Frameworks/R.framework/Versions/4.2-arm64/Resources/library/RcppEigen/include/Eigen,
namespace Eigen {
~

;
In file included from <built-in>:1:
In file included from /Library/Frameworks/R.framework/Versions/4.2-arm64/Resources/library/S
In file included from /Library/Frameworks/R.framework/Versions/4.2-arm64/Resources/library/R
/Library/Frameworks/R.framework/Versions/4.2-arm64/Resources/library/RcppEigen/include/Eigen,
#include <complex>
~~~~~~

3 errors generated.
make: *** [foo.o] Error 1

```

Extract the results:

```

res <- mod %>%
  gather_draws(mu[t]) %>%
  median_qi() %>%
  mutate(year = years[t])

```

Plot the results:

```

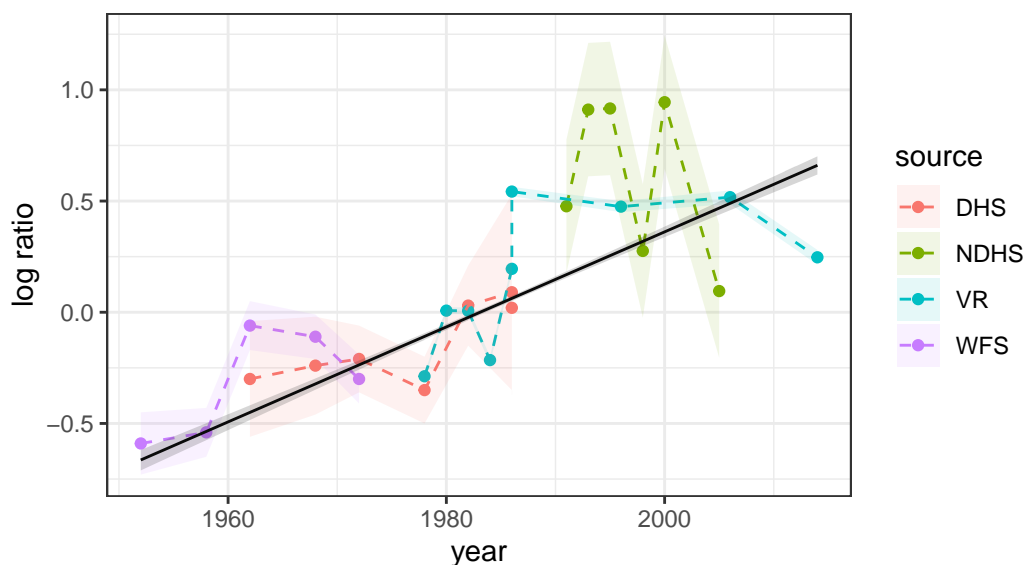
ggplot(lka, aes(year, logit_ratio)) +
  geom_point(aes( color = source)) +
  geom_line(aes( color = source), lty = 2) +
  geom_ribbon(aes(ymin = logit_ratio - se,
                  ymax = logit_ratio + se,
                  fill = source), alpha = 0.1) +

  theme_bw()+
  geom_line(data = res, aes(year, .value)) +
  geom_ribbon(data = res, aes(y = .value, ymin = .lower, ymax = .upper), alpha = 0.2)+
  theme_bw()+
  labs(title = "Ratio of neonatal to other child mortality (logged), Sri Lanka",
       y = "log ratio", subtitle = "Linear fit shown in black")

```

Ratio of neonatal to other child mortality (logged), Sri Lanka

Linear fit shown in black



Question 1

Project the linear model above out to 2023 by adding a `generated quantities` block in Stan (do the projections based on the expected value μ). Plot the resulting projections on a graph similar to that above.

```
stan_data <- list(y = lka$logit_ratio, year_i = observed_years - years[1]+1,
                 T = nyears, years = years, N = length(observed_years),
                 mid_year = mean(years), se = lka$se, P = 9)

mod2 <- stan(data = stan_data,
             file = "l10q1.stan",
             refresh = 0)
```

Running /Library/Frameworks/R.framework/Resources/bin/R CMD SHLIB foo.c

clang -arch arm64 -I"/Library/Frameworks/R.framework/Resources/include" -DNDEBUG -I"/Library/Frameworks/R.framework/Versions/4.2-arm64/Resources/library/S

In file included from <built-in>:1:

In file included from /Library/Frameworks/R.framework/Versions/4.2-arm64/Resources/library/S

In file included from /Library/Frameworks/R.framework/Versions/4.2-arm64/Resources/library/R

In file included from /Library/Frameworks/R.framework/Versions/4.2-arm64/Resources/library/R

/Library/Frameworks/R.framework/Versions/4.2-arm64/Resources/library/RcppEigen/include/Eigen

namespace Eigen {

```

~
/Library/Frameworks/R.framework/Versions/4.2-arm64/Resources/library/RcppEigen/include/Eigen,
namespace Eigen {
    ~
    ;
In file included from <built-in>:1:
In file included from /Library/Frameworks/R.framework/Versions/4.2-arm64/Resources/library/S:
In file included from /Library/Frameworks/R.framework/Versions/4.2-arm64/Resources/library/R:
/Library/Frameworks/R.framework/Versions/4.2-arm64/Resources/library/RcppEigen/include/Eigen,
#include <complex>
    ~~~~~~
3 errors generated.
make: *** [foo.o] Error 1

```

```

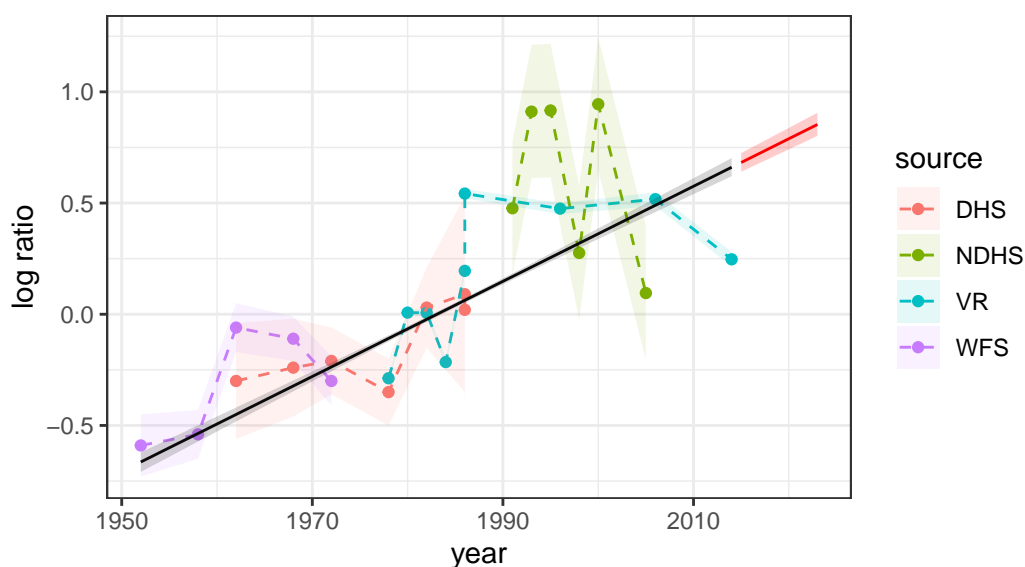
res <- mod2 |>
  gather_draws(mu[t]) |>
  median_qi() |>
  mutate(year = years[t])

res_p <- mod2 |>
  gather_draws(mu_p[p]) |>
  median_qi() |>
  mutate(year = nyears+years[p])

ggplot(lka, aes(year, logit_ratio)) +
  geom_point(aes( color = source)) +
  geom_line(aes( color = source), lty = 2) +
  geom_ribbon(aes(ymin = logit_ratio - se,
                  ymax = logit_ratio + se,
                  fill = source), alpha = 0.1) +
  theme_bw()+
  geom_line(data = res, aes(year, .value)) +
  geom_ribbon(data = res, aes(y = .value, ymin = .lower, ymax = .upper), alpha = 0.2)+
  geom_line(data = res_p, aes(year, .value), col = 'red') +
  geom_ribbon(data = res_p, aes(y = .value, ymin = .lower, ymax = .upper), alpha = 0.2, fi
  theme_bw()+
  labs(title = "Ratio of neonatal to other child mortality (logged), Sri Lanka",
        y = "log ratio", subtitle = "Linear fit shown in black")

```

Ratio of neonatal to other child mortality (logged), Sri Lanka
Linear fit shown in black



Random walks

Question 2

Code up and estimate a first order random walk model to fit to the Sri Lankan data, taking into account measurement error, and project out to 2023.

```
mod3 <- stan(data = stan_data,
             file = "l10q2.stan",
             refresh = 0)
```

```
Running /Library/Frameworks/R.framework/Resources/bin/R CMD SHLIB foo.c
clang -arch arm64 -I"/Library/Frameworks/R.framework/Resources/include" -DNDEBUG -I"/Library/Frameworks/R.framework/Versions/4.2-arm64/Resources/library/S
In file included from <built-in>:1:
In file included from /Library/Frameworks/R.framework/Versions/4.2-arm64/Resources/library/S
In file included from /Library/Frameworks/R.framework/Versions/4.2-arm64/Resources/library/R
In file included from /Library/Frameworks/R.framework/Versions/4.2-arm64/Resources/library/R
/Library/Frameworks/R.framework/Versions/4.2-arm64/Resources/library/RcppEigen/include/Eigen,
namespace Eigen {
^
/Library/Frameworks/R.framework/Versions/4.2-arm64/Resources/library/RcppEigen/include/Eigen,
```

```

namespace Eigen {
    ^
    ;
In file included from <built-in>:1:
In file included from /Library/Frameworks/R.framework/Versions/4.2-arm64/Resources/library/S:
In file included from /Library/Frameworks/R.framework/Versions/4.2-arm64/Resources/library/R:
/Library/Frameworks/R.framework/Versions/4.2-arm64/Resources/library/RcppEigen/include/Eigen:
#include <complex>
    ^~~~~~
3 errors generated.
make: *** [foo.o] Error 1

```

mod3

Inference for Stan model: l10q2.
 4 chains, each with iter=2000; warmup=1000; thin=1;
 post-warmup draws per chain=1000, total post-warmup draws=4000.

	mean	se_mean	sd	2.5%	25%	50%	75%	97.5%	n_eff	Rhat
mu[1]	-0.57	0.00	0.13	-0.83	-0.66	-0.57	-0.48	-0.30	4496	1.00
mu[2]	-0.55	0.00	0.19	-0.95	-0.68	-0.55	-0.43	-0.18	3929	1.00
mu[3]	-0.54	0.00	0.22	-1.00	-0.69	-0.54	-0.39	-0.10	3462	1.00
mu[4]	-0.53	0.00	0.23	-0.98	-0.68	-0.53	-0.38	-0.07	3298	1.00
mu[5]	-0.52	0.00	0.22	-0.95	-0.66	-0.52	-0.38	-0.09	3348	1.00
mu[6]	-0.51	0.00	0.18	-0.87	-0.63	-0.51	-0.39	-0.16	3667	1.00
mu[7]	-0.50	0.00	0.10	-0.71	-0.57	-0.50	-0.43	-0.30	3948	1.00
mu[8]	-0.41	0.00	0.17	-0.74	-0.53	-0.41	-0.30	-0.07	4083	1.00
mu[9]	-0.32	0.00	0.19	-0.70	-0.44	-0.32	-0.20	0.06	4098	1.00
mu[10]	-0.23	0.00	0.17	-0.56	-0.34	-0.23	-0.11	0.13	3848	1.00
mu[11]	-0.14	0.00	0.10	-0.32	-0.20	-0.14	-0.07	0.06	3731	1.00
mu[12]	-0.14	0.00	0.18	-0.48	-0.26	-0.14	-0.03	0.22	3713	1.00
mu[13]	-0.14	0.00	0.21	-0.56	-0.28	-0.14	0.00	0.29	3607	1.00
mu[14]	-0.14	0.00	0.22	-0.57	-0.29	-0.15	0.00	0.31	3719	1.00
mu[15]	-0.14	0.00	0.21	-0.55	-0.27	-0.14	-0.01	0.27	3960	1.00
mu[16]	-0.14	0.00	0.17	-0.48	-0.25	-0.14	-0.03	0.21	4392	1.00
mu[17]	-0.14	0.00	0.08	-0.31	-0.20	-0.14	-0.08	0.02	4407	1.00
mu[18]	-0.17	0.00	0.16	-0.50	-0.28	-0.17	-0.07	0.15	4272	1.00
mu[19]	-0.21	0.00	0.19	-0.58	-0.33	-0.21	-0.09	0.15	4102	1.00
mu[20]	-0.24	0.00	0.17	-0.58	-0.34	-0.23	-0.13	0.09	3877	1.00
mu[21]	-0.26	0.00	0.08	-0.43	-0.32	-0.26	-0.21	-0.10	4154	1.00
mu[22]	-0.27	0.00	0.17	-0.61	-0.38	-0.27	-0.16	0.06	2911	1.00

mu[23]	-0.27	0.00	0.21	-0.69	-0.40	-0.27	-0.14	0.13	3439	1.00
mu[24]	-0.27	0.00	0.21	-0.70	-0.42	-0.28	-0.13	0.16	3271	1.00
mu[25]	-0.28	0.00	0.20	-0.67	-0.41	-0.28	-0.15	0.13	3642	1.00
mu[26]	-0.28	0.00	0.16	-0.59	-0.39	-0.28	-0.18	0.04	4483	1.00
mu[27]	-0.29	0.00	0.01	-0.31	-0.30	-0.29	-0.28	-0.26	6381	1.00
mu[28]	-0.14	0.00	0.12	-0.38	-0.21	-0.14	-0.07	0.10	6363	1.00
mu[29]	0.01	0.00	0.02	-0.02	0.00	0.01	0.02	0.03	5131	1.00
mu[30]	0.00	0.00	0.12	-0.24	-0.08	0.00	0.08	0.25	5647	1.00
mu[31]	0.01	0.00	0.02	-0.03	-0.01	0.01	0.02	0.04	6111	1.00
mu[32]	-0.10	0.00	0.12	-0.34	-0.18	-0.10	-0.02	0.14	6532	1.00
mu[33]	-0.21	0.00	0.02	-0.24	-0.22	-0.21	-0.20	-0.18	6244	1.00
mu[34]	0.07	0.00	0.12	-0.16	-0.01	0.07	0.15	0.31	6509	1.00
mu[35]	0.34	0.00	0.01	0.32	0.33	0.34	0.35	0.37	5651	1.00
mu[36]	0.38	0.00	0.16	0.06	0.27	0.38	0.48	0.71	4088	1.00
mu[37]	0.42	0.00	0.21	0.02	0.29	0.41	0.55	0.82	3513	1.00
mu[38]	0.45	0.00	0.22	0.01	0.31	0.45	0.59	0.89	3207	1.00
mu[39]	0.49	0.00	0.21	0.08	0.35	0.49	0.63	0.92	3078	1.00
mu[40]	0.53	0.00	0.19	0.16	0.40	0.53	0.65	0.91	2825	1.00
mu[41]	0.59	0.00	0.21	0.19	0.45	0.59	0.72	1.00	2594	1.00
mu[42]	0.65	0.00	0.18	0.30	0.52	0.64	0.77	1.03	2457	1.00
mu[43]	0.62	0.00	0.18	0.27	0.50	0.62	0.74	1.00	2809	1.00
mu[44]	0.60	0.00	0.14	0.34	0.51	0.60	0.69	0.89	2795	1.00
mu[45]	0.48	0.00	0.02	0.43	0.46	0.48	0.49	0.53	4695	1.00
mu[46]	0.48	0.00	0.15	0.18	0.38	0.48	0.57	0.77	4830	1.00
mu[47]	0.48	0.00	0.16	0.16	0.37	0.48	0.59	0.80	4183	1.00
mu[48]	0.54	0.00	0.19	0.16	0.42	0.54	0.67	0.92	3872	1.00
mu[49]	0.61	0.00	0.19	0.24	0.49	0.61	0.74	0.98	3007	1.00
mu[50]	0.58	0.00	0.22	0.15	0.43	0.57	0.72	1.04	3201	1.00
mu[51]	0.55	0.00	0.23	0.10	0.39	0.54	0.70	1.02	3249	1.00
mu[52]	0.51	0.00	0.23	0.06	0.36	0.51	0.65	0.97	3119	1.00
mu[53]	0.47	0.00	0.20	0.07	0.34	0.48	0.61	0.85	3156	1.00
mu[54]	0.44	0.00	0.14	0.14	0.35	0.44	0.53	0.70	3792	1.00
mu[55]	0.51	0.00	0.03	0.45	0.49	0.51	0.53	0.58	5753	1.00
mu[56]	0.48	0.00	0.17	0.15	0.37	0.48	0.59	0.81	4555	1.00
mu[57]	0.45	0.00	0.21	0.04	0.31	0.45	0.59	0.89	3501	1.00
mu[58]	0.42	0.00	0.24	-0.05	0.26	0.41	0.57	0.91	2635	1.00
mu[59]	0.39	0.00	0.24	-0.08	0.23	0.38	0.54	0.87	2738	1.00
mu[60]	0.35	0.00	0.24	-0.10	0.20	0.35	0.50	0.83	2976	1.00
mu[61]	0.32	0.00	0.21	-0.09	0.18	0.31	0.45	0.74	3220	1.00
mu[62]	0.28	0.00	0.16	-0.04	0.18	0.28	0.38	0.61	3590	1.00
mu[63]	0.25	0.00	0.03	0.18	0.23	0.25	0.27	0.31	5949	1.00
sigma	0.17	0.00	0.04	0.11	0.14	0.16	0.19	0.26	491	1.00
mu_p[1]	0.24	0.00	0.18	-0.11	0.13	0.25	0.36	0.59	3898	1.00

mu_p[2]	0.25	0.00	0.25	-0.24	0.09	0.24	0.40	0.73	3846	1.00
mu_p[3]	0.25	0.00	0.30	-0.34	0.06	0.25	0.44	0.83	3896	1.00
mu_p[4]	0.25	0.01	0.34	-0.43	0.02	0.25	0.47	0.91	3872	1.00
mu_p[5]	0.25	0.01	0.39	-0.49	-0.01	0.24	0.50	1.00	3913	1.00
mu_p[6]	0.25	0.01	0.42	-0.56	-0.02	0.24	0.51	1.08	3779	1.00
mu_p[7]	0.25	0.01	0.45	-0.64	-0.04	0.25	0.53	1.14	3840	1.00
mu_p[8]	0.25	0.01	0.49	-0.71	-0.07	0.25	0.55	1.22	3779	1.00
mu_p[9]	0.25	0.01	0.52	-0.77	-0.09	0.24	0.57	1.28	3726	1.00
lp__	-7.09	0.60	11.86	-31.17	-14.85	-6.60	1.55	14.75	392	1.01

Samples were drawn using NUTS(diag_e) at Mon Mar 27 12:29:55 2023.
 For each parameter, n_eff is a crude measure of effective sample size,
 and Rhat is the potential scale reduction factor on split chains (at
 convergence, Rhat=1).

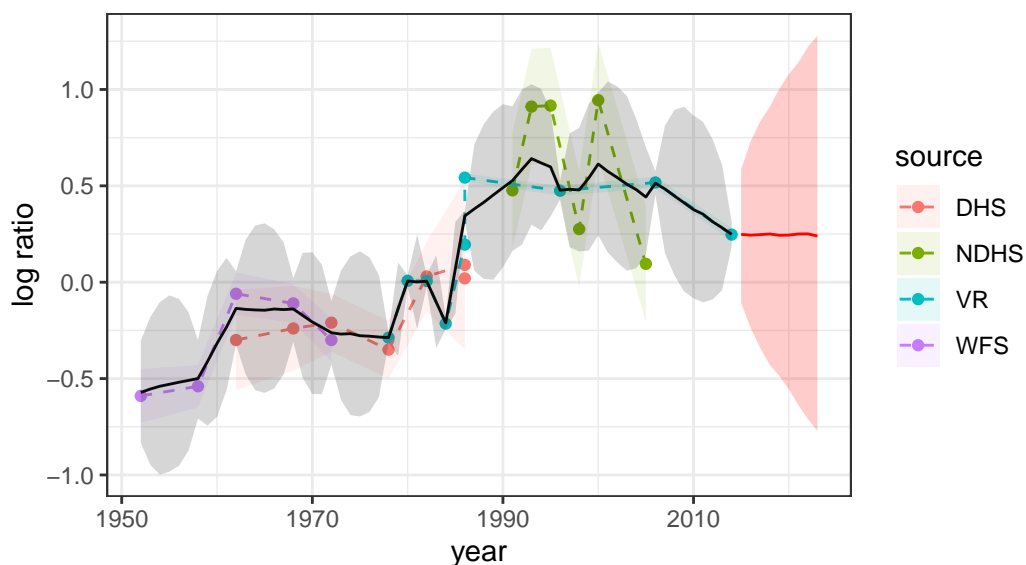
```
res <- mod3 |>
  gather_draws(mu[t]) |>
  median_qi() |>
  mutate(year = years[t])

res_p <- mod3 |>
  gather_draws(mu_p[p]) |>
  median_qi() |>
  mutate(year = nyears+ years[p])

ggplot(lka, aes(year, logit_ratio)) +
  geom_point(aes( color = source)) +
  geom_line(aes( color = source), lty = 2) +
  geom_ribbon(aes(ymin = logit_ratio - se,
                 ymax = logit_ratio + se,
                 fill = source), alpha = 0.1) +
  theme_bw()+
  geom_line(data = res, aes(year, .value)) +
  geom_ribbon(data = res, aes(y = .value, ymin = .lower, ymax = .upper), alpha = 0.2)+
  geom_line(data = res_p, aes(year, .value), col = 'red') +
  geom_ribbon(data = res_p, aes(y = .value, ymin = .lower, ymax = .upper), alpha = 0.2, fill = 'red') +
  theme_bw()+
  labs(title = "Ratio of neonatal to other child mortality (logged), Sri Lanka",
       y = "log ratio", subtitle = "Linear fit shown in black")
```

Ratio of neonatal to other child mortality (logged), Sri Lanka

Linear fit shown in black



Question 3

Now alter your model above to estimate and project a second-order random walk model (RW2).

```
mod4 <- stan(data = stan_data,
             file = "l10q3.stan",
             refresh = 0)
```

```
Running /Library/Frameworks/R.framework/Resources/bin/R CMD SHLIB foo.c
clang -arch arm64 -I"/Library/Frameworks/R.framework/Resources/include" -DNDEBUG -I"/Library/Frameworks/R.framework/Versions/4.2-arm64/Resources/library/S
In file included from <built-in>:1:
In file included from /Library/Frameworks/R.framework/Versions/4.2-arm64/Resources/library/S
In file included from /Library/Frameworks/R.framework/Versions/4.2-arm64/Resources/library/R
In file included from /Library/Frameworks/R.framework/Versions/4.2-arm64/Resources/library/R
/Library/Frameworks/R.framework/Versions/4.2-arm64/Resources/library/RcppEigen/include/Eigen,
namespace Eigen {
~
/Library/Frameworks/R.framework/Versions/4.2-arm64/Resources/library/RcppEigen/include/Eigen,
namespace Eigen {
~
;
```

```

In file included from <built-in>:1:
In file included from /Library/Frameworks/R.framework/Versions/4.2-arm64/Resources/library/S:
In file included from /Library/Frameworks/R.framework/Versions/4.2-arm64/Resources/library/R:
/Library/Frameworks/R.framework/Versions/4.2-arm64/Resources/library/RcppEigen/include/Eigen.
#include <complex>
    ^~~~~~
3 errors generated.
make: *** [foo.o] Error 1

```

mod4

```

Inference for Stan model: l10q3.
4 chains, each with iter=2000; warmup=1000; thin=1;
post-warmup draws per chain=1000, total post-warmup draws=4000.

```

	mean	se_mean	sd	2.5%	25%	50%	75%	97.5%	n_eff	Rhat
mu[1]	-0.58	0.00	0.14	-0.84	-0.67	-0.58	-0.48	-0.31	4646	1.00
mu[2]	-0.59	0.00	0.19	-0.96	-0.71	-0.59	-0.46	-0.21	2782	1.00
mu[3]	-0.60	0.01	0.26	-1.11	-0.76	-0.60	-0.43	-0.10	2375	1.00
mu[4]	-0.61	0.01	0.27	-1.14	-0.79	-0.61	-0.43	-0.07	2290	1.00
mu[5]	-0.60	0.00	0.24	-1.07	-0.76	-0.60	-0.45	-0.14	2395	1.00
mu[6]	-0.58	0.00	0.17	-0.92	-0.69	-0.58	-0.47	-0.25	2484	1.00
mu[7]	-0.52	0.00	0.11	-0.73	-0.59	-0.52	-0.45	-0.32	3113	1.00
mu[8]	-0.43	0.00	0.14	-0.71	-0.52	-0.42	-0.33	-0.15	3140	1.00
mu[9]	-0.32	0.00	0.17	-0.63	-0.43	-0.32	-0.21	0.01	3074	1.00
mu[10]	-0.20	0.00	0.14	-0.47	-0.30	-0.21	-0.11	0.08	3297	1.00
mu[11]	-0.11	0.00	0.10	-0.30	-0.18	-0.11	-0.04	0.08	3623	1.00
mu[12]	-0.06	0.00	0.16	-0.37	-0.16	-0.06	0.05	0.26	2830	1.00
mu[13]	-0.04	0.00	0.22	-0.45	-0.18	-0.04	0.10	0.41	2435	1.00
mu[14]	-0.04	0.00	0.24	-0.50	-0.20	-0.04	0.11	0.43	2335	1.00
mu[15]	-0.06	0.00	0.21	-0.48	-0.21	-0.06	0.07	0.37	2559	1.00
mu[16]	-0.09	0.00	0.15	-0.39	-0.20	-0.09	0.00	0.20	2917	1.00
mu[17]	-0.13	0.00	0.09	-0.30	-0.19	-0.13	-0.07	0.04	4082	1.00
mu[18]	-0.17	0.00	0.14	-0.44	-0.26	-0.17	-0.08	0.10	3004	1.00
mu[19]	-0.20	0.00	0.16	-0.53	-0.30	-0.20	-0.09	0.12	3021	1.00
mu[20]	-0.24	0.00	0.13	-0.50	-0.32	-0.23	-0.15	0.02	3395	1.00
mu[21]	-0.28	0.00	0.09	-0.45	-0.34	-0.27	-0.22	-0.11	4032	1.00
mu[22]	-0.33	0.00	0.15	-0.61	-0.42	-0.33	-0.23	-0.02	3031	1.00
mu[23]	-0.38	0.00	0.21	-0.78	-0.51	-0.38	-0.25	0.04	2585	1.00
mu[24]	-0.41	0.00	0.22	-0.85	-0.55	-0.41	-0.27	0.04	2486	1.00
mu[25]	-0.41	0.00	0.19	-0.79	-0.54	-0.41	-0.30	-0.03	2748	1.00

mu[26]	-0.38	0.00	0.12	-0.62	-0.45	-0.38	-0.30	-0.14	3103	1.00
mu[27]	-0.29	0.00	0.01	-0.31	-0.30	-0.29	-0.28	-0.26	7838	1.00
mu[28]	-0.13	0.00	0.07	-0.27	-0.18	-0.13	-0.09	0.00	4808	1.00
mu[29]	0.01	0.00	0.01	-0.02	0.00	0.01	0.02	0.04	6783	1.00
mu[30]	0.05	0.00	0.06	-0.07	0.01	0.05	0.09	0.18	5978	1.00
mu[31]	0.00	0.00	0.02	-0.03	-0.01	0.00	0.01	0.04	6693	1.00
mu[32]	-0.15	0.00	0.06	-0.27	-0.19	-0.15	-0.11	-0.01	6734	1.00
mu[33]	-0.21	0.00	0.02	-0.24	-0.22	-0.21	-0.20	-0.18	6388	1.00
mu[34]	0.02	0.00	0.07	-0.10	-0.02	0.02	0.06	0.16	6489	1.00
mu[35]	0.34	0.00	0.01	0.32	0.33	0.34	0.35	0.37	6834	1.00
mu[36]	0.55	0.00	0.12	0.31	0.47	0.55	0.62	0.79	2898	1.00
mu[37]	0.67	0.00	0.20	0.27	0.54	0.67	0.79	1.05	2698	1.00
mu[38]	0.72	0.00	0.24	0.24	0.57	0.72	0.87	1.17	2515	1.00
mu[39]	0.73	0.00	0.24	0.23	0.58	0.73	0.89	1.19	2425	1.00
mu[40]	0.73	0.00	0.22	0.27	0.58	0.73	0.88	1.14	2393	1.00
mu[41]	0.74	0.00	0.21	0.33	0.60	0.74	0.88	1.12	2572	1.00
mu[42]	0.73	0.00	0.19	0.35	0.60	0.73	0.86	1.09	2594	1.00
mu[43]	0.67	0.00	0.16	0.35	0.56	0.67	0.78	1.00	2548	1.00
mu[44]	0.59	0.00	0.11	0.38	0.51	0.58	0.65	0.81	2982	1.00
mu[45]	0.48	0.00	0.02	0.43	0.46	0.48	0.49	0.53	6802	1.00
mu[46]	0.44	0.00	0.11	0.22	0.37	0.44	0.51	0.66	3421	1.00
mu[47]	0.47	0.00	0.16	0.14	0.36	0.47	0.58	0.78	2835	1.00
mu[48]	0.54	0.00	0.20	0.16	0.40	0.54	0.67	0.93	1966	1.00
mu[49]	0.61	0.01	0.22	0.17	0.45	0.60	0.76	1.05	1428	1.00
mu[50]	0.61	0.01	0.26	0.12	0.44	0.61	0.78	1.14	1324	1.00
mu[51]	0.58	0.01	0.28	0.04	0.39	0.58	0.77	1.15	1350	1.01
mu[52]	0.54	0.01	0.28	0.00	0.35	0.53	0.71	1.10	1416	1.01
mu[53]	0.49	0.01	0.23	0.04	0.34	0.50	0.64	0.95	1600	1.00
mu[54]	0.48	0.00	0.14	0.19	0.39	0.48	0.57	0.74	1624	1.00
mu[55]	0.51	0.00	0.03	0.45	0.49	0.51	0.53	0.57	7209	1.00
mu[56]	0.53	0.00	0.16	0.22	0.43	0.52	0.63	0.84	1656	1.00
mu[57]	0.53	0.01	0.27	0.00	0.35	0.52	0.69	1.08	1475	1.00
mu[58]	0.51	0.01	0.35	-0.16	0.28	0.50	0.72	1.26	1470	1.00
mu[59]	0.47	0.01	0.39	-0.29	0.22	0.45	0.71	1.30	1491	1.00
mu[60]	0.43	0.01	0.38	-0.31	0.19	0.41	0.66	1.23	1624	1.00
mu[61]	0.37	0.01	0.31	-0.23	0.17	0.36	0.57	1.02	1811	1.00
mu[62]	0.31	0.00	0.19	-0.07	0.19	0.31	0.43	0.70	2008	1.00
mu[63]	0.25	0.00	0.03	0.18	0.22	0.25	0.27	0.31	7547	1.00
sigma	0.14	0.00	0.04	0.09	0.12	0.14	0.16	0.23	471	1.00
mu_p[1]	0.31	0.00	0.24	-0.17	0.16	0.31	0.46	0.79	2829	1.00
mu_p[2]	0.25	0.00	0.15	-0.06	0.16	0.25	0.35	0.55	3461	1.00
mu_p[3]	0.19	0.01	0.41	-0.65	-0.07	0.19	0.44	1.02	3603	1.00
mu_p[4]	0.13	0.01	0.73	-1.35	-0.33	0.14	0.60	1.57	3499	1.00

mu_p[5]	0.07	0.02	1.08	-2.08	-0.61	0.09	0.76	2.21	3555	1.00
mu_p[6]	0.02	0.02	1.45	-2.85	-0.90	0.03	0.93	2.88	3611	1.00
mu_p[7]	-0.04	0.03	1.83	-3.66	-1.21	-0.03	1.11	3.56	3649	1.00
mu_p[8]	-0.09	0.04	2.24	-4.51	-1.52	-0.08	1.33	4.36	3720	1.00
mu_p[9]	-0.14	0.04	2.66	-5.41	-1.84	-0.13	1.53	5.09	3770	1.00
lp__	1.86	0.64	13.04	-25.30	-6.72	2.92	11.20	24.58	416	1.00

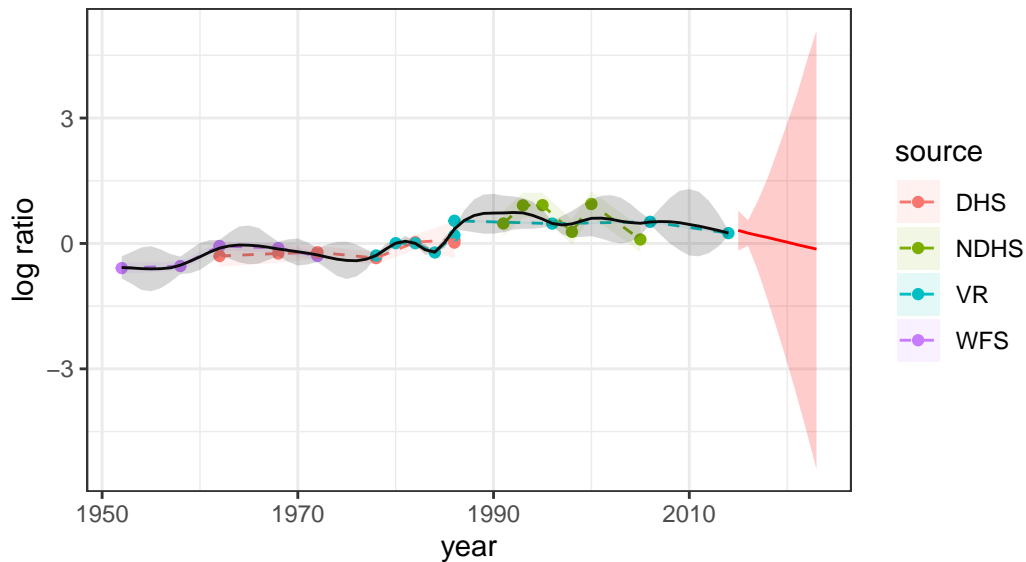
Samples were drawn using NUTS(diag_e) at Mon Mar 27 12:30:07 2023.
 For each parameter, n_eff is a crude measure of effective sample size,
 and Rhat is the potential scale reduction factor on split chains (at
 convergence, Rhat=1).

```
res <- mod4 |>
  gather_draws(mu[t]) |>
  median_qi() |>
  mutate(year = years[t])

res_p <- mod4 |>
  gather_draws(mu_p[p]) |>
  median_qi() |>
  mutate(year = nyears+ years[p])

ggplot(lka, aes(year, logit_ratio)) +
  geom_point(aes( color = source)) +
  geom_line(aes( color = source), lty = 2) +
  geom_ribbon(aes(ymin = logit_ratio - se,
                 ymax = logit_ratio + se,
                 fill = source), alpha = 0.1) +
  theme_bw()+
  geom_line(data = res, aes(year, .value)) +
  geom_ribbon(data = res, aes(y = .value, ymin = .lower, ymax = .upper), alpha = 0.2)+
  geom_line(data = res_p, aes(year, .value), col = 'red') +
  geom_ribbon(data = res_p, aes(y = .value, ymin = .lower, ymax = .upper), alpha = 0.2, fill = 'red') +
  theme_bw()+
  labs(title = "Ratio of neonatal to other child mortality (logged), Sri Lanka",
       y = "log ratio", subtitle = "Linear fit shown in black")
```

Ratio of neonatal to other child mortality (logged), Sri Lanka
Linear fit shown in black



as we can see, the second order projection credible interval is not revealing anything, as discussed during class.

Question 4

Run the first order and second order random walk models, including projections out to 2023. Compare these estimates with the linear fit by plotting everything on the same graph.

let's plot them together:

```
res2 <- mod2 |>
  gather_draws(mu[t]) |>
  median_qi() |>
  mutate(year = years[t])

res3 <- mod3 |>
  gather_draws(mu[t]) |>
  median_qi() |>
  mutate(year = years[t])

res4 <- mod4 |>
  gather_draws(mu[t]) |>
```

```

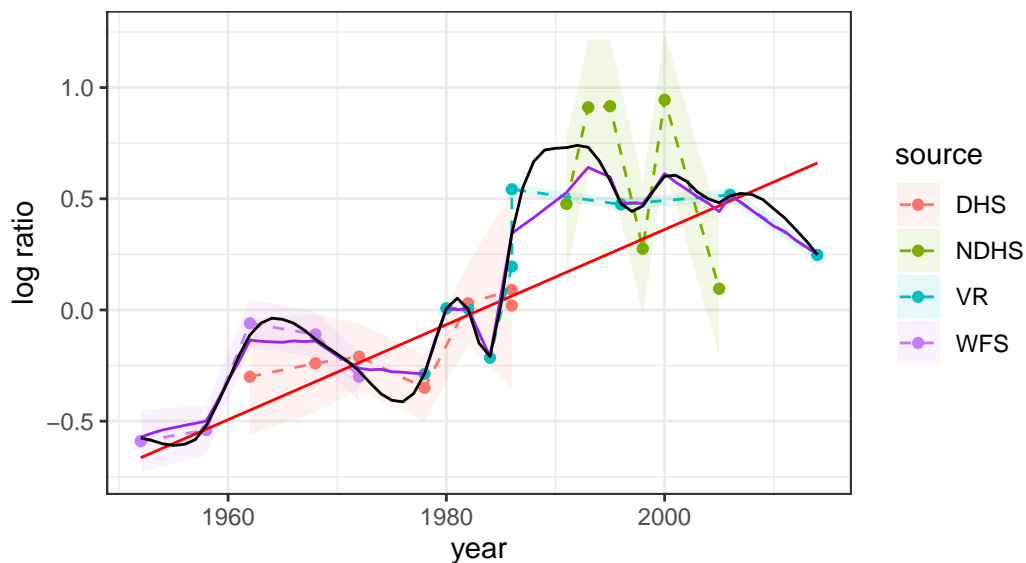
median_qi() |>
mutate(year = years[t])

ggplot(lka, aes(year, logit_ratio)) +
  geom_point(aes( color = source)) +
  geom_line(aes( color = source), lty = 2) +
  geom_ribbon(aes(ymin = logit_ratio - se,
                 ymax = logit_ratio + se,
                 fill = source), alpha = 0.1) +

  theme_bw()+
  geom_line(data = res2, aes(year, .value), col = 'red') +
  geom_line(data = res3, aes(year, .value), col = 'purple') +
  geom_line(data = res4, aes(year, .value), col = 'black') +
  theme_bw()+
  labs(title = "Ratio of neonatal to other child mortality (logged), Sri Lanka",
       y = "log ratio", subtitle = "Linear fit shown in black")

```

Ratio of neonatal to other child mortality (logged), Sri Lanka
Linear fit shown in black



as we can see from the plot, the black line is the fit and prediction of 2nd order RW, the purple line is the fit and prediction of 1st order RW, and the red line is the fit and prediction of linear mode. The linear model is forecasting an increment in children mortality for the incoming 9 years, due to the fluctuation of the dataset. However, the RW model both fits the model more adaptively, and forecasts a decrement in mortality rate. ## Question 5

Rerun the RW2 model excluding the VR data. Briefly comment on the differences between the two data situations.

let's remove the VR data first:

```
df <- filter(lka, source != 'VR')
observed_years <- df$year
years <- min(observed_years):max(observed_years)
nyears <- length(years)
stan_data <- list(y = df$logit_ratio, year_i = observed_years - years[1]+1,
                 T = nyears, years = years, N = length(observed_years),
                 mid_year = mean(years), se = df$se, P = 9)

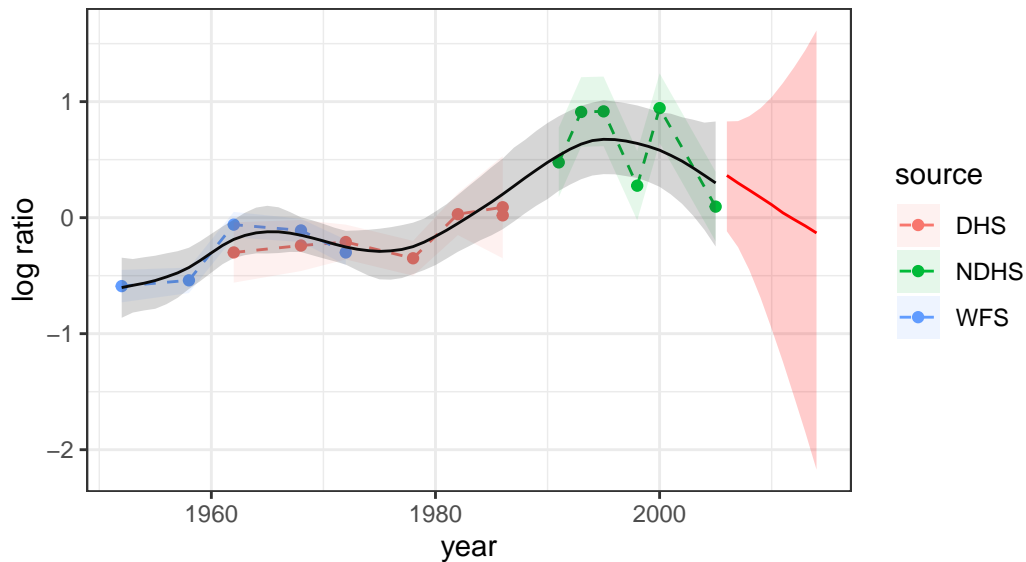
mod5 <- stan(data = stan_data,
            file = "l10q3.stan",
            refresh = 0)

res <- mod5 |>
  gather_draws(mu[t]) |>
  median_qi() |>
  mutate(year = years[t])

res_p <- mod5 |>
  gather_draws(mu_p[p]) |>
  median_qi() |>
  mutate(year = nyears+ years[p])

ggplot(df, aes(year, logit_ratio)) +
  geom_point(aes( color = source)) +
  geom_line(aes( color = source), lty = 2) +
  geom_ribbon(aes(ymin = logit_ratio - se,
                ymax = logit_ratio + se,
                fill = source), alpha = 0.1) +
  theme_bw()+
  geom_line(data = res, aes(year, .value)) +
  geom_ribbon(data = res, aes(y = .value, ymin = .lower, ymax = .upper), alpha = 0.2)+
  geom_line(data = res_p, aes(year, .value), col = 'red') +
  geom_ribbon(data = res_p, aes(y = .value, ymin = .lower, ymax = .upper), alpha = 0.2, fill = 'red') +
  theme_bw()+
  labs(title = "Ratio of neonatal to other child mortality (logged), Sri Lanka",
       y = "log ratio", subtitle = "Linear fit shown in black")
```


Ratio of neonatal to other child mortality (logged), Sri Lanka
 Linear fit shown in black



as we can see, the fitted curve is more smooth in the `geom_ribbon` area, and the prediction error bar is narrower.

Question 6

Briefly comment on which model you think is most appropriate, or an alternative model that would be more appropriate in this context.

I think the second order Random walk model so far reflects the model changes the best, it gives a smooth estimation to each of the point. I am not so sure about the accuracy of each data source, but it seems like removing VR makes the prediction more accurate, but it only makes sense if VR dataset is poorly collected or measured.