

ITS Analysis

7/5/23

About this script

This script was crafted using Ellen's example codes. The script has three sections:

1. Data info
2. Exploration
3. Model fit

Install and/or load packages if required

Standard ITS

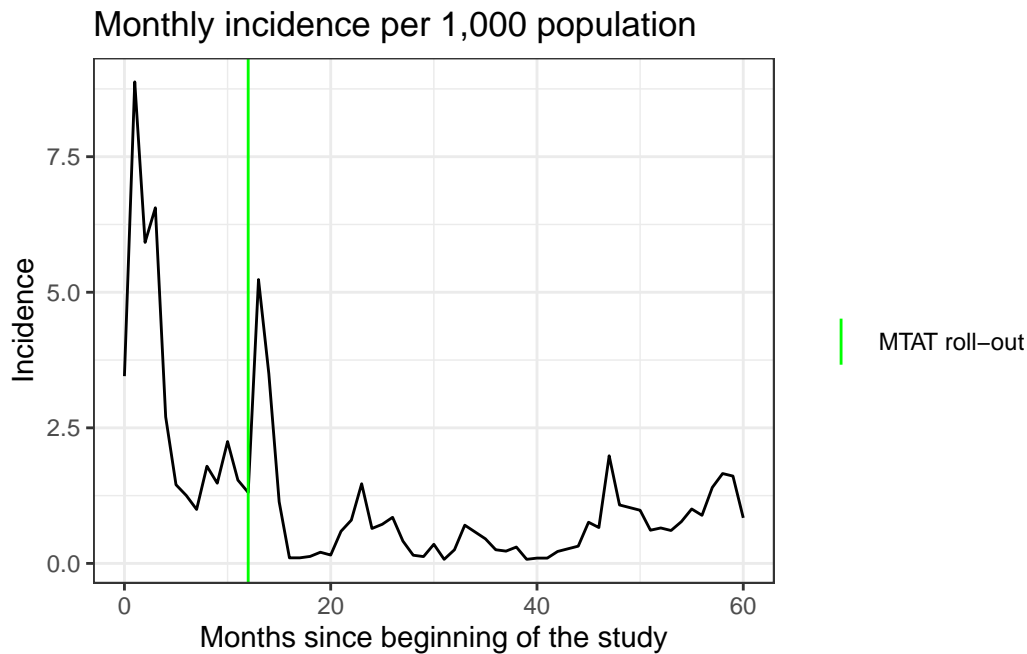
The data was aggregated monthly from 2013 to 2018 at the village level in Amhara Region, Ethiopia. The below code will read the data from the `xlsx` file. The intervention took a 2014 mass test-and-treat campaign for malaria control, followed by focal test-and-treat through 2017. Intervention impact is estimated using the INLA package.

Case data: There are 6 villages, all intervention groups. Villages are observed for 61 months and incidence cases were captured each month.

Intervention (MTAT): the intervention rolled out on month 13 (Sept 2014) in all intervention villages. Intervention villages are filtered for standard ITS.

Inspect the data structure and variable lists

Standard ITS - Exploratory



Standard ITS - Modelling

Monthly incidence rate estimates assuming a negative binomial distribution, using the log population as an offset, with stratum as a fixed effect, random intercepts for each village (IID), and the month of observation (AR-1). To incorporate lagged environmental factors, a lagged 0 - 3 was created from the original **evi**, **rainfall**, **1st_nigth**, and **1st_day** variables, using the **lag()** function. We fitted four models: covariate-only model, lag0, lag1 and lag3. The table below showed that a model with lag 2 has a low dic value.

model	dic
covariate_only	1714.670
lag0	1489.779
lag1	1493.649
lag2	1488.981
lag3	1492.203
nocov	1491.168

Standard ITS model - startum as random effect

model	dic
lag0	1490.373
lag1	1493.671
lag2	1489.182
lag3	1492.391
nocov	1491.078

Save the model results in data frame

	Incidence rate ratio	2.5th quantile, IRR	97.5th quantile, IRR
Intercept	0.000	0.000	0.003
Months since study beginning	0.851	0.686	1.055
Months since MTAT	1.146	0.910	1.442
Post_MTAT_period	0.628	0.112	3.522
2-month lagged EVI	0.060	0.001	2.718
2-month lagged rainfall	1.003	0.999	1.007
2-month lagged night temp	0.950	0.825	1.094
2-month lagged day temp	1.123	1.036	1.218
Feb	0.532	0.195	1.467
Mar	0.617	0.213	1.813
Apr	0.361	0.109	1.217
May	0.448	0.124	1.655
Jun	0.857	0.232	3.245
Jul	1.560	0.444	5.607
Aug	3.699	1.143	12.219
Sep	4.059	0.945	17.796
Oct	12.013	2.921	49.650
Nov	7.645	2.164	27.358
Dec	3.547	1.276	9.875

Model 2 result interpretation.

The intercept, which represents the estimated incidence rate ratio when all other variables are at their reference levels, does not show a significant association with the outcome. The uncertainty intervals for the intercept include the value 1, indicating a lack of evidence for a significant effect.

Regarding the months since the study began and months since MTAT, there is some evidence of a slight increase in the incidence rate with an increase in these variables. However, the uncertainty intervals include the value 1, suggesting that the associations may not be statistically significant.

During the post-MTAT period, there is evidence to suggest a lower estimated incidence rate compared to the pre-MTAT period. The uncertainty intervals for the post-MTAT period exclude the value 1, indicating a significant association.

When examining the variables Stratum2b and Stratum3, the estimated incidence rate ratios suggest different levels of association with the outcome. Stratum 2 b has a relatively smaller association, while Stratum 3 shows a significantly higher estimated incidence rate ratio. The uncertainty intervals for Stratum 3 indicate high variability, indicating the small sample size though it's statistically significant association.

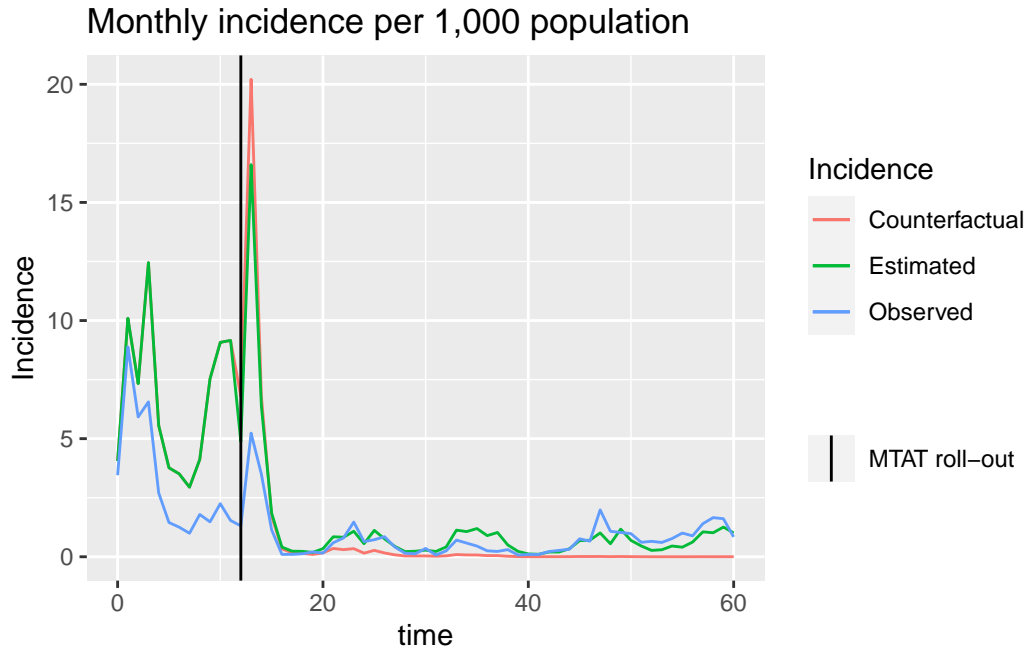
The environmental factors, such as 2-month lagged EVI, rainfall, night temperature, and day temperature, show varying levels of association with the outcome. The estimated incidence rate ratios suggest their impact on the outcome, but the uncertainty intervals indicate some uncertainty in the estimates.

Finally, for the different months or time periods, some months show lower estimated incidence rates (e.g., February, April, May, and June), while others exhibit higher estimated incidence rates (e.g., July, August, September, October, November, and December). The uncertainty intervals can provide further information on the significance of these associations.

	Incidence rate ratio	2.5th quantile, IRR	97.5th quantile, IRR
Intercept	0.000	0.000	0.001
Months since study beginning	0.857	0.690	1.062
Months since MTAT	1.139	0.904	1.433
Post_MTAT_period	0.609	0.110	3.435
Stratum2b	0.846	0.119	6.284
Stratum3	14.498	2.014	118.748
2-month lagged EVI	0.080	0.002	3.487
2-month lagged rainfall	1.003	0.999	1.006
2-month lagged night temp	0.943	0.820	1.083
2-month lagged day temp	1.122	1.035	1.217
Feb	0.540	0.198	1.490
Mar	0.636	0.219	1.874
Apr	0.378	0.114	1.275
May	0.470	0.130	1.738
Jun	0.899	0.244	3.411
Jul	1.674	0.478	6.013
Aug	3.834	1.182	12.730
Sep	4.175	0.969	18.357
Oct	12.110	2.942	50.149
Nov	7.691	2.168	27.577
Dec	3.547	1.275	9.895

A counterfactual analysis is below to estimate how many additional cases would have been observed if not for the intervention. To accomplish this, we convert the predicted case counts, which are labelled `mean` in `models$summary.fitted.values`, back to the log scale so that we can remove the effects associated with the intervention. Specifically, the regression coefficient for the post-intervention period (level change) and time since intervention roll-out (trend change), multiplied by time since roll-out, are subtracted. Now, we have counterfactual case counts in log-scale, which we can exponentiate to get counterfactual cases. In our data, cases declined by 769, or a 35% decrease, during the MTAT/FTAT period.

Counter-factual plot



Controlled ITS

The data was aggregated monthly from 2013 to 2018 at the village level in Amhara Region, Ethiopia. The code provided below will read the data from the specified `.xlsx` file (e.g., "filename.xlsx") and the corresponding sheet name (if applicable). The purpose of this analysis is to conduct a controlled interrupted time series analysis in R, focusing on a mass test-and-treat campaign in 2014 (MTAT), followed by focal test-and-treat interventions until 2017 for malaria control. The intervention impact will be estimated using the Integrated Nested Laplace Approximation (INLA) package.

Case data: The study includes a total of 12 villages, with 6 villages assigned to the intervention group and 6 to the control group. These villages were observed for 61 months, and the

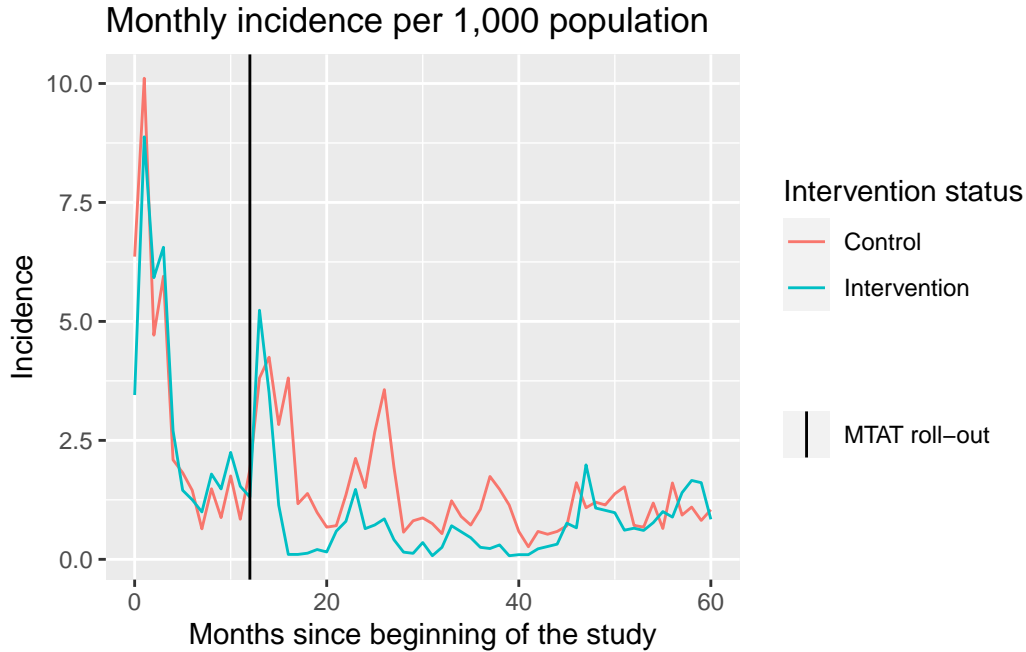
incidence cases were captured monthly.

Intervention (MTAT): The intervention was implemented in all intervention villages in September 2014 (month 13). The villages were divided into two groups: control and intervention.

Time: Time is defined in months. In the dataset, there are three-time variables: (1) time since the beginning of the study, a spline term representing months since the beginning of the intervention, (2) time_post spline representing the post-intervention period, and (3) a post-indicator to determine observations following the intervention roll-out. Our exposures of interest are the interactions of the group with (1) the post-intervention period and (2) the time since the intervention.

Covariates: The analysis includes environmental data extracted from the Malaria Atlas Project database, such as evi, daylight temp, night light temp, and rainfall. Additionally, intervention data related to vector control measures, including Long-Lasting Insecticidal Nets (LLINs) and Indoor Residual Spraying (IRS), are considered covariates.

Incidence plot: Intervention vs control



We estimate monthly incidence rates, assuming a negative binomial distribution, using log population as an offset with, **fixed effects of stratum**, random intercepts for each village (IID) and the month of observation (AR-1). Interaction terms are used to estimate the relative change in incidence rates in intervention areas compared to control areas 1) immediately following MTAT (level change) and 2) each additional month thereafter (trend change). Similar to the SITS, we fitted four different models and a model with lag 2 showed low dic.

Save the model results (stratum as fixed effect) in data frame.

	Incidence rate ratio	2.5th quantile, IRR	97.5th quantile, IRR
Intercept	0.000	0.000	0.001
Months since study beginning	0.807	0.705	0.921
Intervention	0.971	0.081	11.752
Months since MTAT	1.209	1.051	1.392
Post period	4.572	1.504	14.311
Stratum2b	1.293	0.114	14.688
Stratum3	9.641	0.849	112.730
2-month lagged EVI	0.167	0.018	1.534
2-month lagged rainfall	1.002	1.000	1.004
2-month lagged night temp	0.925	0.860	0.996
2-month lagged day temp	1.112	1.062	1.163
Feb	0.685	0.401	1.176
Mar	0.643	0.358	1.160
Apr	0.447	0.231	0.868
May	0.514	0.252	1.049
Jun	0.704	0.336	1.480
Jul	1.289	0.638	2.620
Aug	2.542	1.313	4.948
Sep	3.480	1.505	8.117
Oct	9.488	4.238	21.328
Nov	6.328	3.083	13.079
Dec	3.762	2.134	6.653
Intervention:time	1.125	1.017	1.245
Intervention:time_post	0.893	0.807	0.988
Intervention:post	0.072	0.029	0.175
Intervention:Stratum2b	0.623	0.020	19.317
Intervention:Stratum3	1.510	0.049	46.109

A random intercept for each village (IID), the month of observation (AR-1), and stratum (iid)

model	dic
lag0	3342.247
lag1	3342.214
lag2	3324.511
lag3	3339.188

Model comparison: DIC, CPO and/or PIT

The precision of stratum has been fixed with the prior information.

Deviance information criterion (DIC) between two model stratum as the fixed effect and stratum as a random effect, respectively. The smallest DIC (stratum as a random effect) indicates the best-fit model.

[1] 3324.761 3324.511

Conditional predictive ordinate - the smallest CPO indicates the best-fit model.

A histogram of PIT must resemble a uniform distribution; extreme values indicate outlying observations. Both model's outputs don't show a normal distributions - skewed to the right

Final Results - Modelc22

Over time, for every one-unit increase in months, since the study began, there is an estimated decrease in the incidence rate by approximately 19.3%. Conversely, for every one-unit increase in months since the MTAT event, there is an estimated increase in the incidence rate by approximately 21.0%. During the post-period, there is a substantial estimated increase in the incidence rate by approximately 358.1%. The intervention shows a potential decrease in the incidence rate by approximately 4.3%, although there is some uncertainty in the effects. These findings provide valuable insights for understanding the dynamics of the incidence rate and can inform targeted interventions to mitigate the risk of the outcome. The “2-month lagged EVI” suggests a potential decrease in the incidence rate by approximately 84.7% to 1.2%. Conversely, the “2-month lagged rainfall” and “2-month lagged day temp” indicate a slight increase in the incidence rate by approximately 0.2% to 11.2% and 6.3% to 16.3%, respectively. However, the wide uncertainty intervals highlight the need for further investigation and consideration of other factors to better understand the associations. The analysis uncovered significant interactions between variables, revealing the dynamic nature of interventions over time. The “time: Intervention” interaction indicated that the intervention's effect varied with time, resulting in a 12.3% increase in the incidence rate for each unit increase in time. The “Intervention:time_post” interaction showed an 11.0% decrease in the incidence rate for each unit increase in time during the post period. Moreover, the “Intervention:post” interaction demonstrated a substantial reduction of approximately 92.7% in the incidence rate during the post period when the intervention was implemented. These findings underscore the importance of considering the timing of interventions for understanding their impact on the outcome.

Among the months examined, April, May, and June showed estimated incidence rate ratios below 1, indicating a lower incidence rate compared to the reference month. Specifically, April had an estimated incidence rate ratio of approximately 44.1%, May had 50.2%, and June had 68.9% compared to the reference month. However, the uncertainty intervals for these months suggest some variability and the possibility of non-significant associations.

For the remaining months (February, March, July, August, September, October, November, and December), the estimated incidence rate ratios ranged from approximately 36.0% to 267.2% compared to the reference month. However, since the uncertainty intervals include the value 1 for most of these months, it suggests that the associations may not be statistically significant.

	Incidence rate ratio	2.5th quantile, IRR	97.5th quantile, IRR
Intercept	0.000	0.000	0.006
Months since study beginning	0.807	0.706	0.920
Intervention	0.958	0.248	3.717
Months since MTAT	1.210	1.053	1.391
Post period	4.527	1.501	13.999
2-month lagged EVI	0.152	0.017	1.383
2-month lagged rainfall	1.002	1.000	1.004
2-month lagged night temp	0.931	0.864	1.001
2-month lagged day temp	1.112	1.063	1.163
Feb	0.683	0.400	1.170
Mar	0.640	0.357	1.150
Apr	0.440	0.228	0.852
May	0.501	0.246	1.021
Jun	0.687	0.329	1.441
Jul	1.264	0.626	2.557
Aug	2.527	1.309	4.904
Sep	3.525	1.528	8.199
Oct	9.641	4.315	21.607
Nov	6.404	3.127	13.197
Dec	3.773	2.143	6.659
time: Intervention	1.123	1.015	1.242
Intervention: time__post	0.894	0.807	0.990
Intervention: post	0.073	0.030	0.178

