Residency Exams

```
library(lme4)
Loading required package: Matrix
Warning in check_dep_version(): ABI version mismatch:
lme4 was built with Matrix ABI version 2
Current Matrix ABI version is 1
Please re-install lme4 from source or restore original 'Matrix' package
library(knitr)
library(tidyverse)
-- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
v dplyr 1.1.4 v readr
                                2.1.5
v forcats 1.0.0 v stringr
                                1.5.1
v ggplot2 3.5.1
                   v tibble
                                3.2.1
                 v tidyr
v lubridate 1.9.3
                                1.3.1
v purrr
           1.0.2
-- Conflicts ----- tidyverse_conflicts() --
x tidyr::expand() masks Matrix::expand()
x dplyr::filter() masks stats::filter()
x dplyr::lag()
              masks stats::lag()
x tidyr::pack() masks Matrix::pack()
x tidyr::unpack() masks Matrix::unpack()
i Use the conflicted package (<a href="http://conflicted.r-lib.org/">http://conflicted.r-lib.org/</a>) to force all conflicts to become
library(glmmTMB)
```

```
Warning in checkMatrixPackageVersion(getOption("TMB.check.Matrix", TRUE)): Package version is TMB was built with Matrix ABI version 2
Current Matrix ABI version is 1
Please re-install 'TMB' from source using install.packages('TMB', type = 'source') or ask CR.
```

```
library(broom)
library(emmeans)
```

Welcome to emmeans.

Caution: You lose important information if you filter this package's results. See '? untidy'

Background & Motivation

Prior to practicing full-time medicine, internal medicine students must complete a three year training known as a medical residency. At the end of their medical residency, they are required to pass an examination to obtain their MD. These residency programs are known for their extensive work weeks and rigorous workload.

In 2003 and 2011, two reform policies were passed to change the structure of internal medicine residency. The 2003 reform, passed on July 1, sought to limit the shift length and number of hours worked each week by capping it at 30 hours and 80 hours, respectively. The 2011 reform, also passed on July 1, placed stricter limits on students' shift length, capping it at 16 hours for interns and 28 hours for resident students.

These reforms were passed with the goal of improving patient care by decreasing the stress placed on internal medicine residents. In other words, with more time away from the hospital to rest and rejuvenate, students will be more successful at their jobs when they are on their shift. However, limiting the number of hours medical residents can work each week brings about several concerns regarding their performance of the examination at the end of their residency. While some believed a work time limit granted them more time to study, others thought that less hands on work would result in decreased pass rates.

These concerns bring about out motivating question: is there empirical evidence of an association between the reforms and the rate at which the medical residents passed the exam?

Data Manipulation and Exploratory Data Analysis

The provided data set contains three columns and twenty rows, containing a column for year, number of exam takers, and pass rate as a decimal. Each rows represents a single year in which internal medicine residents completed their residency and took the examination.

```
# Data Manipulation

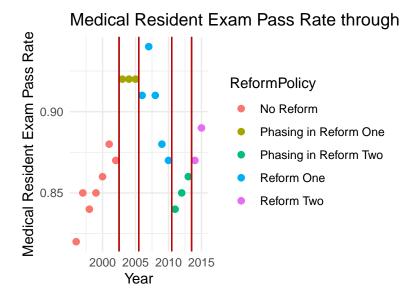
# Add Pass and Fail Count
data$Pass <- round(data$N * data$Pct)
data$Fail <- (data$N - data$Pass)

# Add Time Period Information
data$timeperiod <- rep(1, nrow(data))
data$timeperiod[data$Year > 2002] <- 2
data$timeperiod[data$Year > 2010] <- 3
data$timeperiod <- factor(data$timeperiod, levels = c(1, 2, 3), labels = c("tp1", "tp2", "tp2",
```

Year	N	Pct	Pass	Fail	timeperiod
			5710		-
1997	7173	0.85	6097	1076	tp1
1998	7348	0.84	6172	1176	$\operatorname{tp1}$

We manipulated the data by adding three additional columns for the number of students that passed the examination, the number of students that failed the examination, and the time period based on the year. The first time period is from 1996-2002, when there was no reform policy in place. The second time period is 2003-2010, when the first reform policy was in place. The third time period is 2011-2015, when the second reform policy was in place.

We further manipulated the data by releveling to time period two. Releveling to time period two enabled us to more easily compare time period one to time period two (no reform policy to reform policy one) and time period two to time period three (reform policy one to reform policy two).



Model Implementation Details

Model 1 - Binomial Mixture

We reasonably assumed that exam pass rates contain unobserved year-to-year variation, such as differences in exam difficulty or cohort quality. To capture this extra source of randomness, we fit a binomial mixture model by including a random intercept for each year. This approach allowed us to separate systematic effects of reform policies from idiosyncratic annual noise.

We fit a generalized linear mixed model (GLMM) with a logit link, specifying pass/fail outcomes as the response and reform time period as the fixed effect, with year as a random effect:

$$\begin{aligned} & \text{Pass}_{iy} \sim \text{Binomial}(n_{iy}, p_{iy}), \quad \text{logit}(p_{iy}) = \alpha + \beta \cdot \text{timeperiod}_{iy} + u_y, \quad u_y \sim N(0, \sigma^2). \\ & \#\# \text{ Model 2 - Beta-binomial} \end{aligned}$$

The binomial assumption may underestimate variability in exam outcomes because pass probabilities likely vary within each period. To allow for overdispersion, we modeled the yearly pass probabilities as Beta-distributed. This yields a hierarchical structure where exam outcomes are drawn from a binomial conditional on the latent Beta-distributed rate.

We fit a beta-binomial regression with time period as the predictor using the glmmTMB package with a logit link. The model structure was:

$$\pi_y \sim \text{Beta}(\alpha, \beta), \quad \text{Pass}_y \sim \text{Binomial}(n_y, \pi_y).$$

Here, dispersion was estimated directly from the data, capturing unmodeled heterogeneity.

Model 3 - Beta-binomial on subset

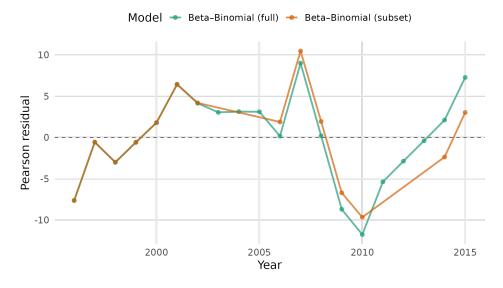
Because policy reforms were phased in gradually, exam cohorts in transition years likely had mixed exposure. To avoid contamination, we fit the beta-binomial model on a restricted dataset excluding these phase-in years. This design isolates the effect of reforms once they were fully implemented.

We restricted the dataset to years 1996–2002, 2006–2010, and 2014–2015. The same beta-binomial specification was used, with time period as the predictor and a logit link for the mean pass probability.

Model Evaluation

To check goodness of fit, we plotted Pearson residuals over time. We appear to minimize residuals over time in the mixture model, as seen below. Putting this together, we concluded that the full beta-binomial mixture model is the best fit.

Residuals Over Time: Full vs Subset Beta-Binomial



Residuals Over Time: Beta-Binomial vs GLMM Mixture



Shortcomings

An issue came with finding good evaluation metrics that compare across models. We initially attempted AIC across all models, but soon learned this was not comparable for any of the models because the mixture modeling package and beta binomial packages use different formulations of AIC. In addition, AIC is dependent on the likelihood and because of this comparing AIC across a full or subset model since they utilize variable number of observations.

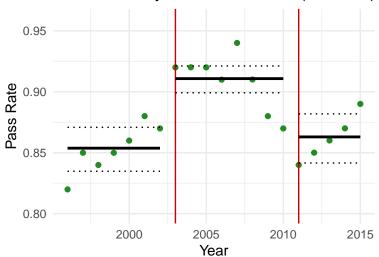
Looking ahead, a next step would be to use Bayesian models, which would allow us to incorporate auxiliary information into the analysis.

Conclusion

```
Estimate Std. Error z value Pr(>|z|)
(Intercept) 2.3234138 0.06885337 33.744374 1.292817e-249
timeperiodtp1 -0.5591105 0.10041764 -5.567851 2.579000e-08
timeperiodtp3 -0.4831369 0.11034058 -4.378597 1.194459e-05
```

We concluded that the binomial mixture model is the best fit. With that model, we found that the first reform increased the odds of passing, while the second reform was associated with a decline in performance. Both of these were confirmed by significant coefficients.

Pass Rates by Year with GLMM (binomial)



Citations

 $https://www.bmj.com/content/366/bmj.l4134\#:\sim:text=The\%20first\%20reform\%2C\%20in\%202003, the\%20perint https://jamanetwork.com/journals/jamainternalmedicine/fullarticle/1672284$