

MEDI504B_ML_Project

Jacob and Hanwei

2023-02-15

1. Introduction and Background

1.1 Introduction

Our overarching goal for this project is to produce a machine learning (ML) model which can accurately predict polycystic ovary syndrome (PCOS) status (presence or absence of disease). PCOS is an endocrine (hormonal) disorder that affects females of a reproductive age. Given the widespread nature of this condition and the troubling symptoms which accompany it, including infertility, it would be helpful for physicians to be able to predict individuals more likely to experience PCOS thereby enabling them to therapeutically intervene and provide care and support in a timely manner.

1.2 Background

Polycystic ovarian syndrome (PCOS) is a common endocrine disorder affecting approximately 10-15% of reproductive-age women worldwide (Teede, H., Misso, M., Tassone, E. C., Dewailly, D., Ng, E. H., Azziz, R., ... & Norman, R. J. (2018). Recommendations from the international evidence-based guideline for the assessment and management of polycystic ovary syndrome. *Human Reproduction*, 33(9), 1602-1618. <https://doi.org/10.1093/humrep/dey256>). The condition is characterized by a complex set of symptoms, including hyperandrogenism, menstrual irregularities, and polycystic ovaries (Bozdag et al., 2016). The diagnosis of PCOS is typically based on clinical and biochemical assessments, as well as ultrasound imaging of the ovaries (Teede et al., 2018). However, the diagnosis of PCOS can be challenging due to the heterogeneous presentation of symptoms and the lack of a single diagnostic criterion (Bozdag, G., Mumusoglu, S., Zengin, D., & Karabulut, E. (2016). The prevalence and phenotypic features of polycystic ovary syndrome: a systematic review and meta-analysis. *Human Reproduction*, 31(12), 2841–2855. <https://doi.org/10.1093/humrep/dew218>).

Machine learning models have shown promise as a potential tool for the accurate prediction of PCOS. Compared to traditional diagnostic methods, machine learning models can utilize large amounts of data from various sources and provide more accurate predictions. This is particularly beneficial in the case of PCOS, as traditional diagnostic methods such as tissue biopsy can be expensive and invasive. Furthermore, machine learning models can assist clinicians in identifying patients who may benefit from early intervention, which can improve long-term health outcomes (Teede et al., 2018).

Overall, the use of machine learning models to predict PCOS has the potential to improve the accuracy of diagnosis and reduce the cost and invasiveness of traditional diagnostic methods. We sought to evaluate the comparative accuracy of multiple machine learning classifiers to select the optimal model for predicting PCOS, given considerations surrounding false positive and false negatives.

Ethics

** include something about the target population and who is at risk ** what to do if someone is pregnant
-> follow up with study doctor* **what to do if you have a suspicion that the patient is at risk for**

a life threatening medical disorder? people taking hormone therapy ** identification of patients given the data

Labels and predictors

1.3 Data

The dataset consists of physical and clinical parameters collected from 10 hospitals across Kerala, India, to determine PCOS and infertility-related issues. The dataset contains information that can be used to analyze and understand the diagnosis and treatment of PCOS and infertility.

2. Objectives

Our objectives were twofold:

- 1: Develop a simple model that can predict PCOS status using clinical and physiologic data that can be acquired using a routine blood test and assessment by a general practitioner clinician.
- 2: Optimize the model for specificity, thus minimizing the risk of a false positive in model predications.

The rationale for objective one was based on the motivation to provide a data-based method of diagnosis that is cheaper and less invasive than current methods. As discussed in the background section of this report, current diagnosis of PCOS can be time-consuming and invasive, and replacing these methods with a model is advantageous for reasons related to clinical care and resource utilization. This influenced our variable selection, as we did not consider the inclusion of predictors that can not be measured in the above stated context in model development.

Optimizing model specificity for the prediction of PCOS is important because PCOS is associated with several negative health outcomes, including infertility, insulin resistance, and metabolic disorders. Early diagnosis and treatment of PCOS can help prevent or manage these conditions, which can ultimately improve the overall health and quality of life of those affected. However, given that PCOS is not a life-threatening condition in and of itself, it is important to balance the trade-off between maximizing sensitivity and specificity in order to minimize the number of false positives and prevent unnecessary and potentially invasive follow-up testing. By optimizing model specificity, we can ensure that those who are diagnosed with PCOS are more likely to truly have the condition, while also reducing the risk of unnecessary medical interventions for those who do not have PCOS.

3. Methods

3.1 Splitting our dataset into training and testing datasets

Following our EDA (please refer to EDA section), we start by splitting our dataset into the training set and the validation set, followed by building a simple model aimed at using our data to find factors that predict PCOS status (our outcome variable). We chose to partition our data into only two sets based on the relatively small overall sample size of our data and small effective sample size of our data. Further, to help determine generalizability, we believe it would be ideal for our predictive model to be validated on an external dataset derived from a different population than our training and test data.

3.2 Building logistic regression models

We first attempted to model our data with PCOS status as the outcome variable using three different logistic regression models. At this stage of variable selection, we relied on the Akaike Information Criteria (AIC) to assess model fit. Of the tested logistic regression models, we prespecified that we would select the model with the lowest AIC. The three tested models utilized the following construction:

- (1) Model 1 included all the features in our dataset
- (2) Model 2 only includes physiological variables we think are of interest in PCOS, namely the hormone measurements (n=5 different types of hormones)
- (3) Model 3 includes the n=5 the hormone measurement levels we in our dataset, and in addition to these features, the clinical symptoms that have been known to be associated with PCOS: weight gain, hair growth + skin darkening, hair loss and pimples.

Of our three logistic regression models, Model 3 had the lowest AIC score of 566.45. We also assessed autocorrelation and collinearity among the variables in model three. *NEED TO ADD*

Results

4. Results

4a. Exploratory Data Analysis

This section contains the steps and output for the EDA performed on the PCOS dataset. The section proceeds sequentially with each step of EDA. Please refer to bullet points, figure titles and figure captions for more details. The first steps in our EDA involved getting a high-level overview of our dataset and determining the dimension of our data: n=541 rows and n=45 columns. Further checking of the dataset reveals an additional column (column 45). This column does not contain any useful information and is not one of our 44 features, therefore the column was removed. We find that our dataset has the following variables: Sl. No, Patient File No., PCOS (Y/N), Age (yrs), Weight (Kg), Height(Cm), BMI, Blood Group, Pulse rate(bpm), RR(breaths/min), Hb(g/dl), Cycle(R/I), Cycle length(days), Marriage Status (Yrs), Pregnant(Y/N), No. of abortions, I beta-HCG(mIU/mL), II beta-HCG(mIU/mL), FSH(mIU/mL), LH(mIU/mL), FSH/LH, Hip(inch), Waist(inch), Waist:Hip Ratio, TSH (mIU/L), AMH(ng/mL), PRL(ng/mL), Vit D3 (ng/mL), PRG(ng/mL), RBS(mg/dl), Weight gain(Y/N), hair growth(Y/N), Skin darkening (Y/N), Hair loss(Y/N), Pimples(Y/N), Fast food (Y/N), Reg.Exercise(Y/N), BP _Systolic (mmHg), BP _Diastolic (mmHg), Follicle No. (L), Follicle No. (R), Avg. F size (L) (mm), Avg. F size (R) (mm), Endometrium (mm).

Data Wrangling

Next, we formatted the data to ensure the variables are the appropriate class type, this will enable us to perform our EDA. Specifically we converted binary variables (1 or 0) into the character class type.

Bivariate associations

We examined associations for categorical predictors and the outcome (PCOS) by constructing bivariate plots for the predictors stratified by PCOS status.



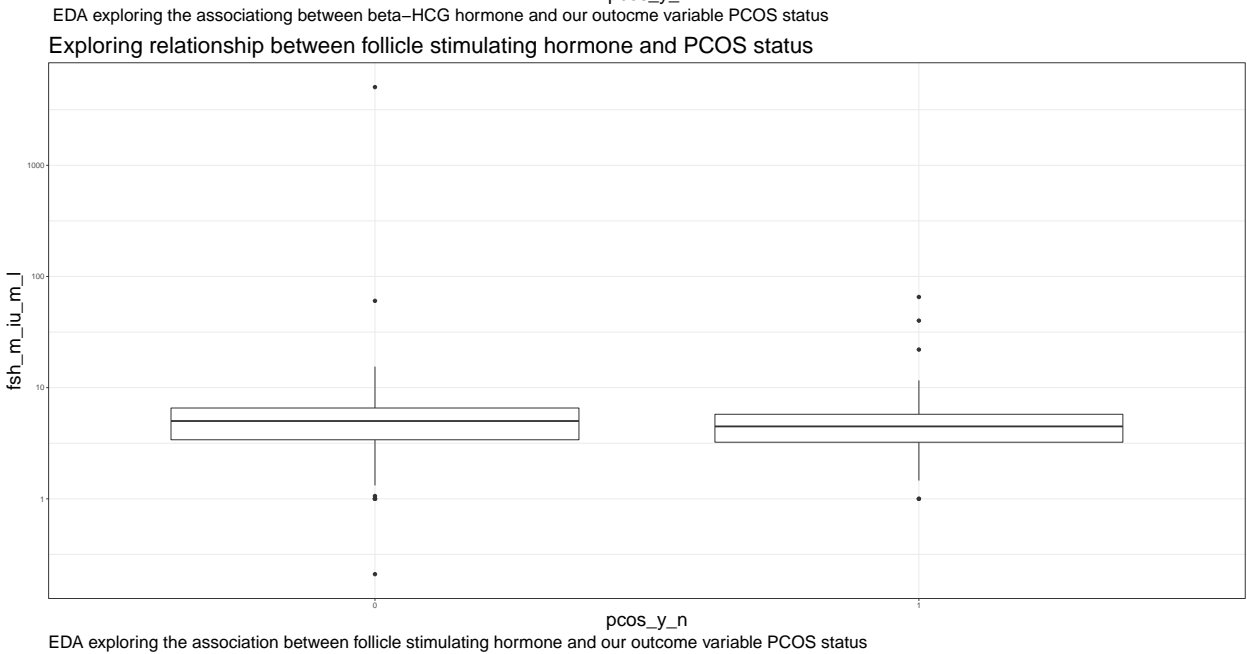
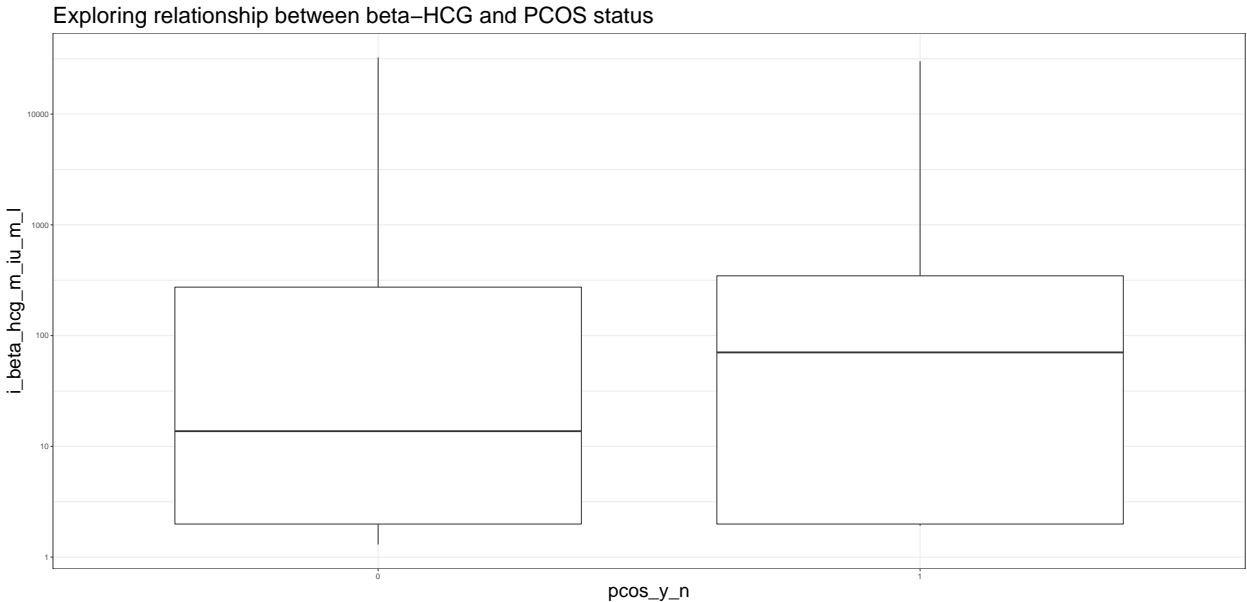
From the figure, we can see that the proportion of individuals with weight gain and a positive PCOS status is greater than the proportion of individuals with no weight gain and a positive PCOS status. We observe similar associations for hair growth and skin darkening. We also observe a similar but much weaker association for the categorical variables of hair loss and pimples. Interestingly, we observe the inverse trend for the categorical variable for fast food consumption. As these predictors appear to be visually associated with the outcome, as well as physiologically feasible to be associated with PCOS, they should be considered for inclusion in the predictive model.

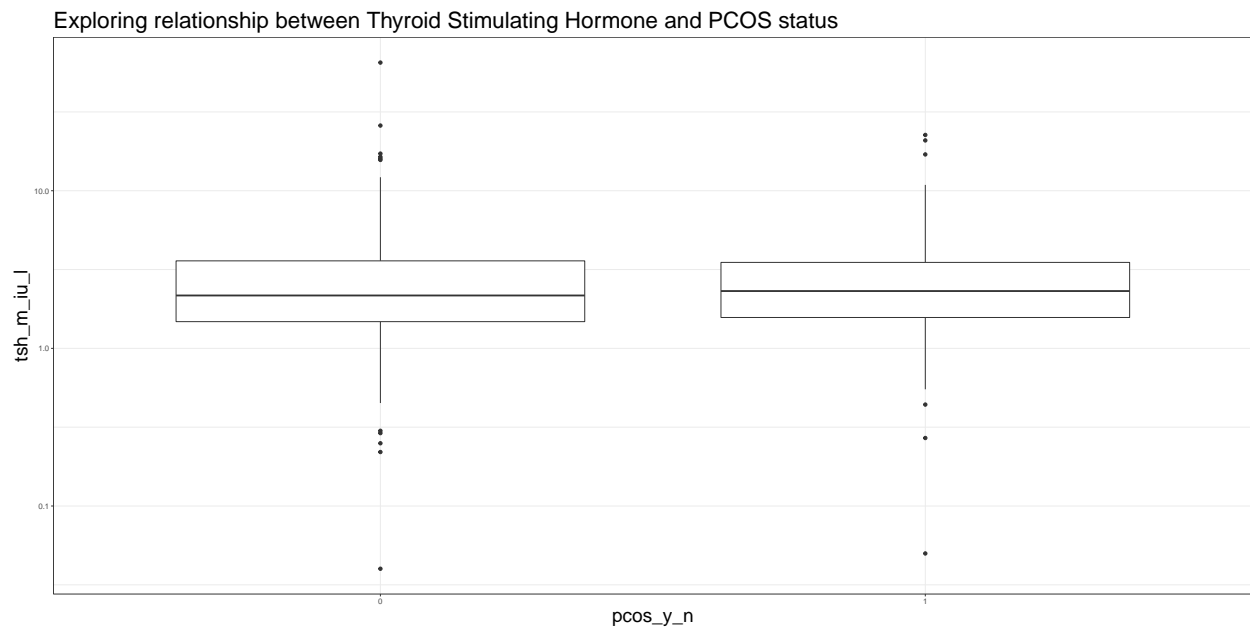
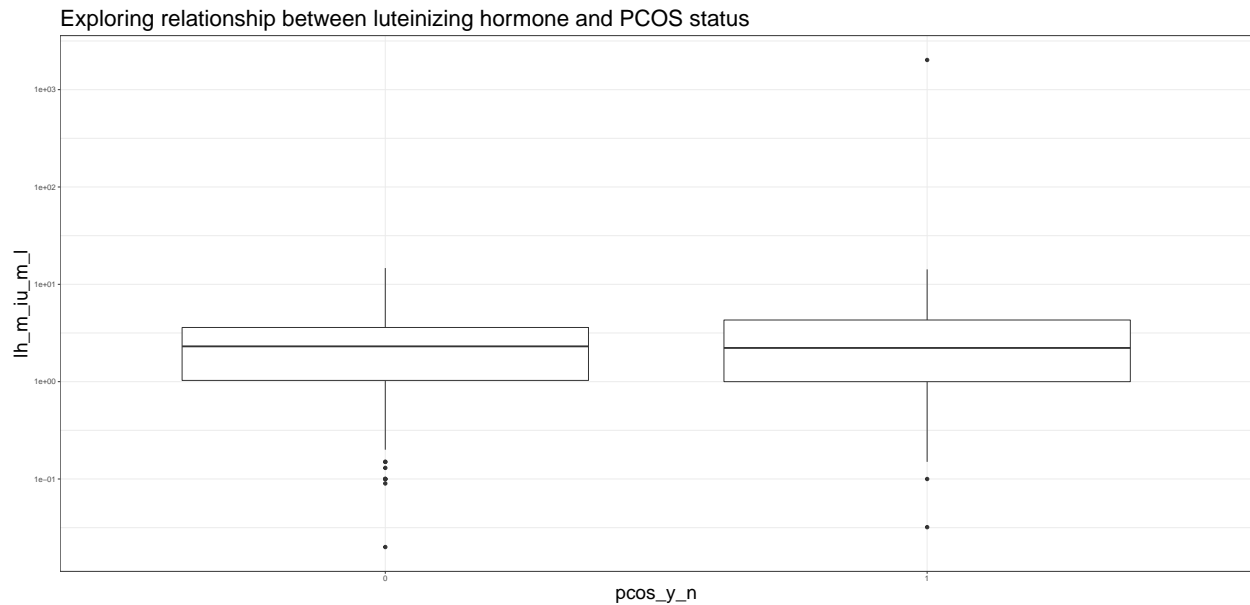
The bar plots for regular exercise and pregnancy do not appear to show a significant difference in proportion of positive PCOS cases, but we would need to perform a statistical analyses (such as an unpaired t-test) to know for sure.

To ascertain the relationship between PCOS status and continuous predictors, we also plotted boxplots to visualize any associations between PCOS status and a continuous variable. Please refer to the appendix for these boxplots.

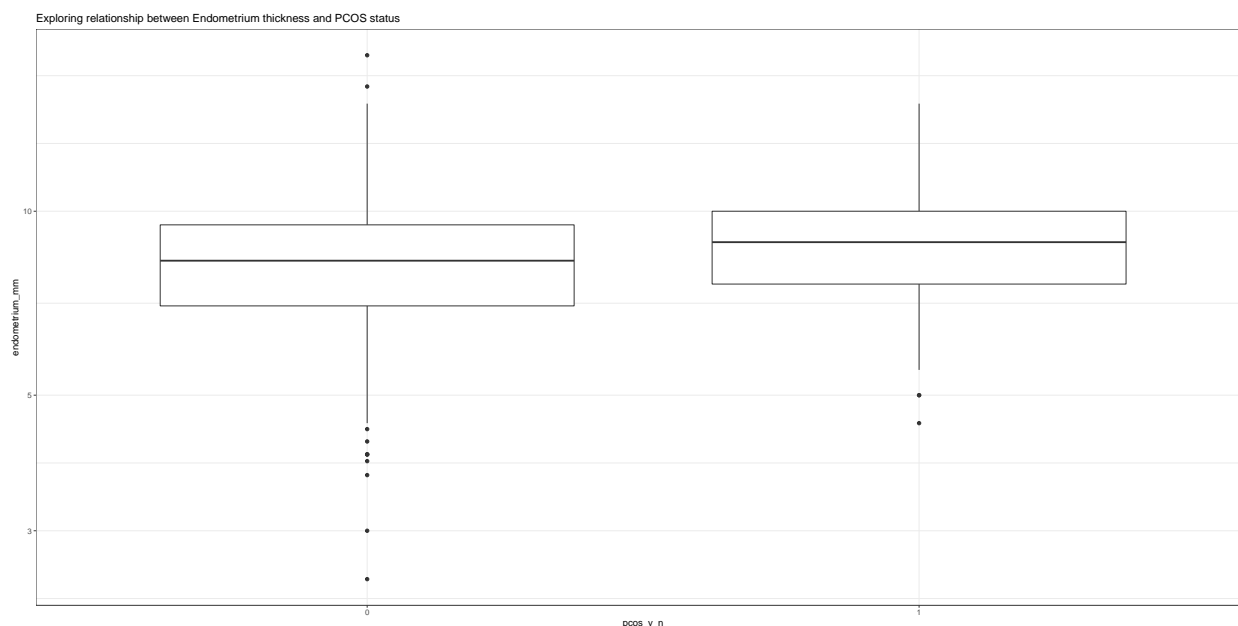
- For this section of the EDA, we looked more closely at the relationships between selected variables and our outcome variable of interest (for this project: yes or no for PCOS), based on how our boxplots looked initially (please refer to Appendix for all boxplots) and on some knowledge gained via our literature survey, we will highlight the relationship of select variables with our outcome variable (PCOS status).
- While the etiology for PCOS is not known, my literature survey suggests PCOS is associated with abnormal hormone levels. Thus, to look into this further as part of the EDA, we compared measurements of hormone levels and PCOS status to get a sense of the relationships between these variables.

Boxplots looking at relationship of hormones and PCOS status





- From the literature, we know that PCOS diagnosis often involves an ultrasound and endometrium thickness is measured as part of this process. Therefore, as part of our EDA, we also sought to look at the relationship between endometrium thickness and PCOS status.



- From this series of boxplots, we did not observe (visually) any significant difference in the association of PCOS status and hormone levels. However, given what we know about the association of PCOS and hormone imbalances from the literature, I think it would still be useful to include these variables when we build our model.
- From the literature², there is also evidence that miscarriages are associated with PCOS. Therefore, as part of our EDA, we sought to look at the relationship between PCOS status and number of abortions.

Modelling

```
#####
## setting the seed (enables randomization in a reproducible way)
set.seed(504)

data_split <- caret::createDataPartition(PCOS_data_without_infertility_formatted_final$pcos_y_n, p=0.7,
train.data <- PCOS_data_without_infertility_formatted_final[data_split,]
dim(train.data)

## [1] 379 44

test.data <- PCOS_data_without_infertility_formatted_final[-data_split,]
dim(test.data)

## [1] 160 44

table(train.data$pcos_y_n)

##
## 0 1
## 255 124
```

```
# the first model we will build will include all the variables
#library(glmnet)
```

```
logistic_model1 <- glm(pcos_y_n ~ .,
                        data = train.data,
                        family = binomial(link="logit"))
```

```
## Warning: glm.fit: algorithm did not converge
```

```
summary(logistic_model1)
```

```
##
## Call:
## glm(formula = pcos_y_n ~ ., family = binomial(link = "logit"),
##      data = train.data)
##
## Deviance Residuals:
##      [1]  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0
##     [26]  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0
##     [51]  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0
##     [76]  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0
##    [101]  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0
##    [126]  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0
##    [151]  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0
##    [176]  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0
##    [201]  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0
##    [226]  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0
##    [251]  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0
##    [276]  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0
##    [301]  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0
##    [326]  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0
##    [351]  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0
##    [376]  0  0  0  0
##
## Coefficients: (805 not defined because of singularities)
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    -2.657e+01  3.561e+05      0      1
## sl_no4         -2.990e-12  5.036e+05      0      1
## sl_no7         -2.925e-12  5.036e+05      0      1
## sl_no8         -2.674e-12  5.036e+05      0      1
## sl_no10        -2.722e-12  5.036e+05      0      1
## sl_no11        -2.968e-12  5.036e+05      0      1
## sl_no12        -3.104e-12  5.036e+05      0      1
## sl_no13         5.313e+01  5.036e+05      0      1
## sl_no15        -3.046e-12  5.036e+05      0      1
## sl_no16        -3.085e-12  5.036e+05      0      1
## sl_no18        -3.135e-12  5.036e+05      0      1
## sl_no19        -3.119e-12  5.036e+05      0      1
## sl_no20         5.313e+01  5.036e+05      0      1
## sl_no21        -3.267e-12  5.036e+05      0      1
## sl_no22        -3.403e-12  5.036e+05      0      1
## sl_no23        -3.200e-12  5.036e+05      0      1
```


## sl_no24	-3.267e-12	5.036e+05	0	1
## sl_no28	-3.326e-12	5.036e+05	0	1
## sl_no30	5.313e+01	5.036e+05	0	1
## sl_no31	-3.225e-12	5.036e+05	0	1
## sl_no32	5.313e+01	5.036e+05	0	1
## sl_no33	-3.138e-12	5.036e+05	0	1
## sl_no34	-3.237e-12	5.036e+05	0	1
## sl_no36	-3.351e-12	5.036e+05	0	1
## sl_no37	-3.335e-12	5.036e+05	0	1
## sl_no39	-3.060e-12	5.036e+05	0	1
## sl_no41	-3.175e-12	5.036e+05	0	1
## sl_no42	-3.248e-12	5.036e+05	0	1
## sl_no43	-3.152e-12	5.036e+05	0	1
## sl_no45	5.313e+01	5.036e+05	0	1
## sl_no46	-3.138e-12	5.036e+05	0	1
## sl_no48	-3.420e-12	5.036e+05	0	1
## sl_no49	-3.120e-12	5.036e+05	0	1
## sl_no50	5.313e+01	5.036e+05	0	1
## sl_no52	-3.290e-12	5.036e+05	0	1
## sl_no54	-3.071e-12	5.036e+05	0	1
## sl_no56	-2.986e-12	5.036e+05	0	1
## sl_no57	-3.190e-12	5.036e+05	0	1
## sl_no58	-3.554e-12	5.036e+05	0	1
## sl_no59	-3.602e-12	5.036e+05	0	1
## sl_no60	-3.102e-12	5.036e+05	0	1
## sl_no64	-2.869e-12	5.036e+05	0	1
## sl_no66	-3.000e-12	5.036e+05	0	1
## sl_no67	-2.704e-12	5.036e+05	0	1
## sl_no68	-3.334e-12	5.036e+05	0	1
## sl_no69	5.313e+01	5.036e+05	0	1
## sl_no70	5.313e+01	5.036e+05	0	1
## sl_no71	-3.124e-12	5.036e+05	0	1
## sl_no72	-3.272e-12	5.036e+05	0	1
## sl_no73	-3.203e-12	5.036e+05	0	1
## sl_no75	-3.414e-12	5.036e+05	0	1
## sl_no76	-2.961e-12	5.036e+05	0	1
## sl_no77	5.313e+01	5.036e+05	0	1
## sl_no80	-2.929e-12	5.036e+05	0	1
## sl_no82	-2.795e-12	5.036e+05	0	1
## sl_no83	-3.197e-12	5.036e+05	0	1
## sl_no84	-2.858e-12	5.036e+05	0	1
## sl_no87	-2.792e-12	5.036e+05	0	1
## sl_no88	-3.212e-12	5.036e+05	0	1
## sl_no89	5.313e+01	5.036e+05	0	1
## sl_no91	-3.181e-12	5.036e+05	0	1
## sl_no92	-3.085e-12	5.036e+05	0	1
## sl_no94	-2.994e-12	5.036e+05	0	1
## sl_no97	5.313e+01	5.036e+05	0	1
## sl_no99	-2.984e-12	5.036e+05	0	1
## sl_no100	-3.027e-12	5.036e+05	0	1
## sl_no101	-3.090e-12	5.036e+05	0	1
## sl_no102	-2.986e-12	5.036e+05	0	1
## sl_no103	5.313e+01	5.036e+05	0	1
## sl_no104	-3.056e-12	5.036e+05	0	1

## sl_no105	-2.861e-12	5.036e+05	0	1
## sl_no106	5.313e+01	5.036e+05	0	1
## sl_no107	-3.016e-12	5.036e+05	0	1
## sl_no108	5.313e+01	5.036e+05	0	1
## sl_no110	-3.385e-12	5.036e+05	0	1
## sl_no111	-3.096e-12	5.036e+05	0	1
## sl_no112	-2.939e-12	5.036e+05	0	1
## sl_no114	5.313e+01	5.036e+05	0	1
## sl_no115	5.313e+01	5.036e+05	0	1
## sl_no116	-3.091e-12	5.036e+05	0	1
## sl_no117	-3.195e-12	5.036e+05	0	1
## sl_no118	5.313e+01	5.036e+05	0	1
## sl_no121	-2.786e-12	5.036e+05	0	1
## sl_no122	-3.191e-12	5.036e+05	0	1
## sl_no123	5.313e+01	5.036e+05	0	1
## sl_no124	5.313e+01	5.036e+05	0	1
## sl_no125	5.313e+01	5.036e+05	0	1
## sl_no126	-3.122e-12	5.036e+05	0	1
## sl_no127	-3.253e-12	5.036e+05	0	1
## sl_no129	-3.162e-12	5.036e+05	0	1
## sl_no130	-3.263e-12	5.036e+05	0	1
## sl_no132	-3.265e-12	5.036e+05	0	1
## sl_no133	5.313e+01	5.036e+05	0	1
## sl_no134	5.313e+01	5.036e+05	0	1
## sl_no135	-2.981e-12	5.036e+05	0	1
## sl_no136	-3.039e-12	5.036e+05	0	1
## sl_no138	-3.413e-12	5.036e+05	0	1
## sl_no139	-3.168e-12	5.036e+05	0	1
## sl_no141	-3.411e-12	5.036e+05	0	1
## sl_no142	5.313e+01	5.036e+05	0	1
## sl_no143	5.313e+01	5.036e+05	0	1
## sl_no144	-3.231e-12	5.036e+05	0	1
## sl_no145	-3.471e-12	5.036e+05	0	1
## sl_no146	-3.464e-12	5.036e+05	0	1
## sl_no150	-3.079e-12	5.036e+05	0	1
## sl_no151	5.313e+01	5.036e+05	0	1
## sl_no152	5.313e+01	5.036e+05	0	1
## sl_no153	5.313e+01	5.036e+05	0	1
## sl_no158	-3.145e-12	5.036e+05	0	1
## sl_no159	5.313e+01	5.036e+05	0	1
## sl_no160	-3.163e-12	5.036e+05	0	1
## sl_no162	-3.245e-12	5.036e+05	0	1
## sl_no163	5.313e+01	5.036e+05	0	1
## sl_no164	-2.950e-12	5.036e+05	0	1
## sl_no166	-2.935e-12	5.036e+05	0	1
## sl_no168	5.313e+01	5.036e+05	0	1
## sl_no169	5.313e+01	5.036e+05	0	1
## sl_no170	5.313e+01	5.036e+05	0	1
## sl_no171	-3.138e-12	5.036e+05	0	1
## sl_no172	5.313e+01	5.036e+05	0	1
## sl_no173	-3.043e-12	5.036e+05	0	1
## sl_no174	5.313e+01	5.036e+05	0	1
## sl_no176	5.313e+01	5.036e+05	0	1
## sl_no177	5.313e+01	5.036e+05	0	1

## sl_no178	5.313e+01	5.036e+05	0	1
## sl_no180	-3.291e-12	5.036e+05	0	1
## sl_no181	5.313e+01	5.036e+05	0	1
## sl_no182	-3.107e-12	5.036e+05	0	1
## sl_no184	-3.188e-12	5.036e+05	0	1
## sl_no185	-2.889e-12	5.036e+05	0	1
## sl_no186	-3.121e-12	5.036e+05	0	1
## sl_no188	-3.082e-12	5.036e+05	0	1
## sl_no189	-3.069e-12	5.036e+05	0	1
## sl_no190	5.313e+01	5.036e+05	0	1
## sl_no192	5.313e+01	5.036e+05	0	1
## sl_no193	-3.021e-12	5.036e+05	0	1
## sl_no195	-3.378e-12	5.036e+05	0	1
## sl_no199	-3.039e-12	5.036e+05	0	1
## sl_no200	5.313e+01	5.036e+05	0	1
## sl_no203	-3.144e-12	5.036e+05	0	1
## sl_no206	5.313e+01	5.036e+05	0	1
## sl_no209	-3.067e-12	5.036e+05	0	1
## sl_no214	5.313e+01	5.036e+05	0	1
## sl_no216	-3.011e-12	5.036e+05	0	1
## sl_no217	5.313e+01	5.036e+05	0	1
## sl_no218	5.313e+01	5.036e+05	0	1
## sl_no219	-3.086e-12	5.036e+05	0	1
## sl_no220	-3.212e-12	5.036e+05	0	1
## sl_no221	-3.151e-12	5.036e+05	0	1
## sl_no222	5.313e+01	5.036e+05	0	1
## sl_no223	5.313e+01	5.036e+05	0	1
## sl_no224	-3.051e-12	5.036e+05	0	1
## sl_no225	-3.011e-12	5.036e+05	0	1
## sl_no226	-3.255e-12	5.036e+05	0	1
## sl_no227	-3.348e-12	5.036e+05	0	1
## sl_no228	5.313e+01	5.036e+05	0	1
## sl_no229	-3.122e-12	5.036e+05	0	1
## sl_no230	-3.185e-12	5.036e+05	0	1
## sl_no231	5.313e+01	5.036e+05	0	1
## sl_no232	-3.164e-12	5.036e+05	0	1
## sl_no233	5.313e+01	5.036e+05	0	1
## sl_no234	-3.279e-12	5.036e+05	0	1
## sl_no235	-3.086e-12	5.036e+05	0	1
## sl_no237	-3.249e-12	5.036e+05	0	1
## sl_no238	5.313e+01	5.036e+05	0	1
## sl_no239	-3.220e-12	5.036e+05	0	1
## sl_no240	-3.153e-12	5.036e+05	0	1
## sl_no241	5.313e+01	5.036e+05	0	1
## sl_no242	-3.098e-12	5.036e+05	0	1
## sl_no243	-3.206e-12	5.036e+05	0	1
## sl_no246	5.313e+01	5.036e+05	0	1
## sl_no247	-3.049e-12	5.036e+05	0	1
## sl_no248	5.313e+01	5.036e+05	0	1
## sl_no251	5.313e+01	5.036e+05	0	1
## sl_no252	-3.003e-12	5.036e+05	0	1
## sl_no253	5.313e+01	5.036e+05	0	1
## sl_no254	-3.174e-12	5.036e+05	0	1
## sl_no255	-2.956e-12	5.036e+05	0	1

## sl_no256	-2.998e-12	5.036e+05	0	1
## sl_no257	-3.042e-12	5.036e+05	0	1
## sl_no258	-3.351e-12	5.036e+05	0	1
## sl_no259	-3.041e-12	5.036e+05	0	1
## sl_no263	-3.087e-12	5.036e+05	0	1
## sl_no264	-3.135e-12	5.036e+05	0	1
## sl_no266	-3.129e-12	5.036e+05	0	1
## sl_no267	-3.142e-12	5.036e+05	0	1
## sl_no268	5.313e+01	5.036e+05	0	1
## sl_no269	-3.202e-12	5.036e+05	0	1
## sl_no270	5.313e+01	5.036e+05	0	1
## sl_no271	-3.201e-12	5.036e+05	0	1
## sl_no272	-3.084e-12	5.036e+05	0	1
## sl_no275	-3.022e-12	5.036e+05	0	1
## sl_no277	5.313e+01	5.036e+05	0	1
## sl_no278	5.313e+01	5.036e+05	0	1
## sl_no280	-3.160e-12	5.036e+05	0	1
## sl_no281	-3.037e-12	5.036e+05	0	1
## sl_no282	-3.113e-12	5.036e+05	0	1
## sl_no284	-3.201e-12	5.036e+05	0	1
## sl_no286	-3.166e-12	5.036e+05	0	1
## sl_no287	-3.057e-12	5.036e+05	0	1
## sl_no288	5.313e+01	5.036e+05	0	1
## sl_no289	-3.204e-12	5.036e+05	0	1
## sl_no291	5.313e+01	5.036e+05	0	1
## sl_no292	-3.025e-12	5.036e+05	0	1
## sl_no293	-3.193e-12	5.036e+05	0	1
## sl_no294	5.313e+01	5.036e+05	0	1
## sl_no295	5.313e+01	5.036e+05	0	1
## sl_no297	-3.018e-12	5.036e+05	0	1
## sl_no298	-3.056e-12	5.036e+05	0	1
## sl_no299	-3.024e-12	5.036e+05	0	1
## sl_no300	-3.145e-12	5.036e+05	0	1
## sl_no301	5.313e+01	5.036e+05	0	1
## sl_no304	5.313e+01	5.036e+05	0	1
## sl_no305	-3.202e-12	5.036e+05	0	1
## sl_no306	-2.905e-12	5.036e+05	0	1
## sl_no308	-3.005e-12	5.036e+05	0	1
## sl_no309	5.313e+01	5.036e+05	0	1
## sl_no310	5.313e+01	5.036e+05	0	1
## sl_no311	-3.074e-12	5.036e+05	0	1
## sl_no312	5.313e+01	5.036e+05	0	1
## sl_no313	5.313e+01	5.036e+05	0	1
## sl_no315	5.313e+01	5.036e+05	0	1
## sl_no316	-3.023e-12	5.036e+05	0	1
## sl_no317	-3.288e-12	5.036e+05	0	1
## sl_no318	-3.197e-12	5.036e+05	0	1
## sl_no321	5.313e+01	5.036e+05	0	1
## sl_no322	-3.218e-12	5.036e+05	0	1
## sl_no323	-3.436e-12	5.036e+05	0	1
## sl_no324	5.313e+01	5.036e+05	0	1
## sl_no326	-3.254e-12	5.036e+05	0	1
## sl_no328	-3.393e-12	5.036e+05	0	1
## sl_no329	5.313e+01	5.036e+05	0	1

## sl_no330	-3.169e-12	5.036e+05	0	1
## sl_no333	-3.150e-12	5.036e+05	0	1
## sl_no336	-3.169e-12	5.036e+05	0	1
## sl_no337	5.313e+01	5.036e+05	0	1
## sl_no338	5.313e+01	5.036e+05	0	1
## sl_no339	5.313e+01	5.036e+05	0	1
## sl_no340	-3.266e-12	5.036e+05	0	1
## sl_no341	5.313e+01	5.036e+05	0	1
## sl_no342	5.313e+01	5.036e+05	0	1
## sl_no343	5.313e+01	5.036e+05	0	1
## sl_no344	-3.135e-12	5.036e+05	0	1
## sl_no345	-2.967e-12	5.036e+05	0	1
## sl_no346	-3.113e-12	5.036e+05	0	1
## sl_no349	-3.068e-12	5.036e+05	0	1
## sl_no350	-3.029e-12	5.036e+05	0	1
## sl_no352	5.313e+01	5.036e+05	0	1
## sl_no353	5.313e+01	5.036e+05	0	1
## sl_no354	-3.016e-12	5.036e+05	0	1
## sl_no355	-3.152e-12	5.036e+05	0	1
## sl_no356	-3.038e-12	5.036e+05	0	1
## sl_no357	5.313e+01	5.036e+05	0	1
## sl_no358	-2.987e-12	5.036e+05	0	1
## sl_no359	-3.237e-12	5.036e+05	0	1
## sl_no361	-3.014e-12	5.036e+05	0	1
## sl_no362	-3.193e-12	5.036e+05	0	1
## sl_no363	-3.135e-12	5.036e+05	0	1
## sl_no366	-3.085e-12	5.036e+05	0	1
## sl_no369	5.313e+01	5.036e+05	0	1
## sl_no370	-3.182e-12	5.036e+05	0	1
## sl_no371	5.313e+01	5.036e+05	0	1
## sl_no373	5.313e+01	5.036e+05	0	1
## sl_no375	-2.925e-12	5.036e+05	0	1
## sl_no376	5.313e+01	5.036e+05	0	1
## sl_no378	-2.976e-12	5.036e+05	0	1
## sl_no379	-3.058e-12	5.036e+05	0	1
## sl_no380	-3.122e-12	5.036e+05	0	1
## sl_no381	-3.065e-12	5.036e+05	0	1
## sl_no382	-3.054e-12	5.036e+05	0	1
## sl_no383	-3.084e-12	5.036e+05	0	1
## sl_no384	-3.137e-12	5.036e+05	0	1
## sl_no385	-3.115e-12	5.036e+05	0	1
## sl_no386	-2.886e-12	5.036e+05	0	1
## sl_no387	-3.111e-12	5.036e+05	0	1
## sl_no388	5.313e+01	5.036e+05	0	1
## sl_no389	-3.075e-12	5.036e+05	0	1
## sl_no392	-3.053e-12	5.036e+05	0	1
## sl_no393	5.313e+01	5.036e+05	0	1
## sl_no394	-3.120e-12	5.036e+05	0	1
## sl_no395	-3.018e-12	5.036e+05	0	1
## sl_no397	-3.107e-12	5.036e+05	0	1
## sl_no398	-3.129e-12	5.036e+05	0	1
## sl_no399	5.313e+01	5.036e+05	0	1
## sl_no400	-3.046e-12	5.036e+05	0	1
## sl_no402	-3.269e-12	5.036e+05	0	1

## sl_no403	5.313e+01	5.036e+05	0	1
## sl_no404	5.313e+01	5.036e+05	0	1
## sl_no405	-2.978e-12	5.036e+05	0	1
## sl_no407	-3.039e-12	5.036e+05	0	1
## sl_no408	-2.913e-12	5.036e+05	0	1
## sl_no409	-3.075e-12	5.036e+05	0	1
## sl_no410	-3.008e-12	5.036e+05	0	1
## sl_no412	-3.190e-12	5.036e+05	0	1
## sl_no413	5.313e+01	5.036e+05	0	1
## sl_no414	-3.249e-12	5.036e+05	0	1
## sl_no417	-3.149e-12	5.036e+05	0	1
## sl_no418	-3.214e-12	5.036e+05	0	1
## sl_no420	-3.158e-12	5.036e+05	0	1
## sl_no421	5.313e+01	5.036e+05	0	1
## sl_no423	5.313e+01	5.036e+05	0	1
## sl_no425	-3.104e-12	5.036e+05	0	1
## sl_no426	5.313e+01	5.036e+05	0	1
## sl_no428	5.313e+01	5.036e+05	0	1
## sl_no429	5.313e+01	5.036e+05	0	1
## sl_no430	5.313e+01	5.036e+05	0	1
## sl_no431	5.313e+01	5.036e+05	0	1
## sl_no433	-3.053e-12	5.036e+05	0	1
## sl_no436	-3.138e-12	5.036e+05	0	1
## sl_no437	-3.043e-12	5.036e+05	0	1
## sl_no439	5.313e+01	5.036e+05	0	1
## sl_no441	-3.105e-12	5.036e+05	0	1
## sl_no442	-3.233e-12	5.036e+05	0	1
## sl_no444	5.313e+01	5.036e+05	0	1
## sl_no446	5.313e+01	5.036e+05	0	1
## sl_no447	-2.966e-12	5.036e+05	0	1
## sl_no449	-3.179e-12	5.036e+05	0	1
## sl_no450	5.313e+01	5.036e+05	0	1
## sl_no451	5.313e+01	5.036e+05	0	1
## sl_no453	5.313e+01	5.036e+05	0	1
## sl_no454	5.313e+01	5.036e+05	0	1
## sl_no455	5.313e+01	5.036e+05	0	1
## sl_no456	5.313e+01	5.036e+05	0	1
## sl_no457	5.313e+01	5.036e+05	0	1
## sl_no461	-3.222e-12	5.036e+05	0	1
## sl_no463	5.313e+01	5.036e+05	0	1
## sl_no464	-3.116e-12	5.036e+05	0	1
## sl_no465	5.313e+01	5.036e+05	0	1
## sl_no466	-3.121e-12	5.036e+05	0	1
## sl_no467	-3.113e-12	5.036e+05	0	1
## sl_no468	5.313e+01	5.036e+05	0	1
## sl_no469	-3.003e-12	5.036e+05	0	1
## sl_no471	5.313e+01	5.036e+05	0	1
## sl_no472	5.313e+01	5.036e+05	0	1
## sl_no474	-3.041e-12	5.036e+05	0	1
## sl_no475	-3.108e-12	5.036e+05	0	1
## sl_no477	-3.100e-12	5.036e+05	0	1
## sl_no479	-3.065e-12	5.036e+05	0	1
## sl_no481	-3.099e-12	5.036e+05	0	1
## sl_no482	-3.080e-12	5.036e+05	0	1

## sl_no483	-3.101e-12	5.036e+05	0	1
## sl_no484	5.313e+01	5.036e+05	0	1
## sl_no485	-3.011e-12	5.036e+05	0	1
## sl_no486	-3.053e-12	5.036e+05	0	1
## sl_no487	5.313e+01	5.036e+05	0	1
## sl_no488	-3.095e-12	5.036e+05	0	1
## sl_no490	5.313e+01	5.036e+05	0	1
## sl_no491	-3.119e-12	5.036e+05	0	1
## sl_no492	-3.099e-12	5.036e+05	0	1
## sl_no494	5.313e+01	5.036e+05	0	1
## sl_no495	-3.109e-12	5.036e+05	0	1
## sl_no496	5.313e+01	5.036e+05	0	1
## sl_no500	-3.062e-12	5.036e+05	0	1
## sl_no501	-3.105e-12	5.036e+05	0	1
## sl_no502	-3.050e-12	5.036e+05	0	1
## sl_no504	-3.077e-12	5.036e+05	0	1
## sl_no506	-3.099e-12	5.036e+05	0	1
## sl_no509	-3.087e-12	5.036e+05	0	1
## sl_no510	5.313e+01	5.036e+05	0	1
## sl_no512	-3.030e-12	5.036e+05	0	1
## sl_no514	-3.051e-12	5.036e+05	0	1
## sl_no518	-3.076e-12	5.036e+05	0	1
## sl_no519	-2.993e-12	5.036e+05	0	1
## sl_no520	-3.103e-12	5.036e+05	0	1
## sl_no522	-2.956e-12	5.036e+05	0	1
## sl_no524	5.313e+01	5.036e+05	0	1
## sl_no526	-3.015e-12	5.036e+05	0	1
## sl_no527	-3.059e-12	5.036e+05	0	1
## sl_no528	-3.101e-12	5.036e+05	0	1
## sl_no529	-3.057e-12	5.036e+05	0	1
## sl_no530	-3.062e-12	5.036e+05	0	1
## sl_no531	-3.067e-12	5.036e+05	0	1
## sl_no533	-3.043e-12	5.036e+05	0	1
## sl_no534	5.313e+01	5.036e+05	0	1
## sl_no535	-3.067e-12	5.036e+05	0	1
## sl_no536	-3.076e-12	5.036e+05	0	1
## sl_no539	-3.072e-12	5.036e+05	0	1
## sl_no540	-3.015e-12	5.036e+05	0	1
## sl_no541	5.313e+01	5.036e+05	0	1
## patient_file_no4	NA	NA	NA	NA
## patient_file_no7	NA	NA	NA	NA
## patient_file_no8	NA	NA	NA	NA
## patient_file_no10	NA	NA	NA	NA
## patient_file_no11	NA	NA	NA	NA
## patient_file_no12	NA	NA	NA	NA
## patient_file_no13	NA	NA	NA	NA
## patient_file_no15	NA	NA	NA	NA
## patient_file_no16	NA	NA	NA	NA
## patient_file_no18	NA	NA	NA	NA
## patient_file_no19	NA	NA	NA	NA
## patient_file_no20	NA	NA	NA	NA
## patient_file_no21	NA	NA	NA	NA
## patient_file_no22	NA	NA	NA	NA
## patient_file_no23	NA	NA	NA	NA

## patient_file_no24	NA	NA	NA	NA
## patient_file_no28	NA	NA	NA	NA
## patient_file_no30	NA	NA	NA	NA
## patient_file_no31	NA	NA	NA	NA
## patient_file_no32	NA	NA	NA	NA
## patient_file_no33	NA	NA	NA	NA
## patient_file_no34	NA	NA	NA	NA
## patient_file_no36	NA	NA	NA	NA
## patient_file_no37	NA	NA	NA	NA
## patient_file_no39	NA	NA	NA	NA
## patient_file_no41	NA	NA	NA	NA
## patient_file_no42	NA	NA	NA	NA
## patient_file_no43	NA	NA	NA	NA
## patient_file_no45	NA	NA	NA	NA
## patient_file_no46	NA	NA	NA	NA
## patient_file_no48	NA	NA	NA	NA
## patient_file_no49	NA	NA	NA	NA
## patient_file_no50	NA	NA	NA	NA
## patient_file_no52	NA	NA	NA	NA
## patient_file_no54	NA	NA	NA	NA
## patient_file_no56	NA	NA	NA	NA
## patient_file_no57	NA	NA	NA	NA
## patient_file_no58	NA	NA	NA	NA
## patient_file_no59	NA	NA	NA	NA
## patient_file_no60	NA	NA	NA	NA
## patient_file_no64	NA	NA	NA	NA
## patient_file_no66	NA	NA	NA	NA
## patient_file_no67	NA	NA	NA	NA
## patient_file_no68	NA	NA	NA	NA
## patient_file_no69	NA	NA	NA	NA
## patient_file_no70	NA	NA	NA	NA
## patient_file_no71	NA	NA	NA	NA
## patient_file_no72	NA	NA	NA	NA
## patient_file_no73	NA	NA	NA	NA
## patient_file_no75	NA	NA	NA	NA
## patient_file_no76	NA	NA	NA	NA
## patient_file_no77	NA	NA	NA	NA
## patient_file_no80	NA	NA	NA	NA
## patient_file_no82	NA	NA	NA	NA
## patient_file_no83	NA	NA	NA	NA
## patient_file_no84	NA	NA	NA	NA
## patient_file_no87	NA	NA	NA	NA
## patient_file_no88	NA	NA	NA	NA
## patient_file_no89	NA	NA	NA	NA
## patient_file_no91	NA	NA	NA	NA
## patient_file_no92	NA	NA	NA	NA
## patient_file_no94	NA	NA	NA	NA
## patient_file_no97	NA	NA	NA	NA
## patient_file_no99	NA	NA	NA	NA
## patient_file_no100	NA	NA	NA	NA
## patient_file_no101	NA	NA	NA	NA
## patient_file_no102	NA	NA	NA	NA
## patient_file_no103	NA	NA	NA	NA
## patient_file_no104	NA	NA	NA	NA

## patient_file_no105	NA	NA	NA	NA
## patient_file_no106	NA	NA	NA	NA
## patient_file_no107	NA	NA	NA	NA
## patient_file_no108	NA	NA	NA	NA
## patient_file_no110	NA	NA	NA	NA
## patient_file_no111	NA	NA	NA	NA
## patient_file_no112	NA	NA	NA	NA
## patient_file_no114	NA	NA	NA	NA
## patient_file_no115	NA	NA	NA	NA
## patient_file_no116	NA	NA	NA	NA
## patient_file_no117	NA	NA	NA	NA
## patient_file_no118	NA	NA	NA	NA
## patient_file_no121	NA	NA	NA	NA
## patient_file_no122	NA	NA	NA	NA
## patient_file_no123	NA	NA	NA	NA
## patient_file_no124	NA	NA	NA	NA
## patient_file_no125	NA	NA	NA	NA
## patient_file_no126	NA	NA	NA	NA
## patient_file_no127	NA	NA	NA	NA
## patient_file_no129	NA	NA	NA	NA
## patient_file_no130	NA	NA	NA	NA
## patient_file_no132	NA	NA	NA	NA
## patient_file_no133	NA	NA	NA	NA
## patient_file_no134	NA	NA	NA	NA
## patient_file_no135	NA	NA	NA	NA
## patient_file_no136	NA	NA	NA	NA
## patient_file_no138	NA	NA	NA	NA
## patient_file_no139	NA	NA	NA	NA
## patient_file_no141	NA	NA	NA	NA
## patient_file_no142	NA	NA	NA	NA
## patient_file_no143	NA	NA	NA	NA
## patient_file_no144	NA	NA	NA	NA
## patient_file_no145	NA	NA	NA	NA
## patient_file_no146	NA	NA	NA	NA
## patient_file_no150	NA	NA	NA	NA
## patient_file_no151	NA	NA	NA	NA
## patient_file_no152	NA	NA	NA	NA
## patient_file_no153	NA	NA	NA	NA
## patient_file_no158	NA	NA	NA	NA
## patient_file_no159	NA	NA	NA	NA
## patient_file_no160	NA	NA	NA	NA
## patient_file_no162	NA	NA	NA	NA
## patient_file_no163	NA	NA	NA	NA
## patient_file_no164	NA	NA	NA	NA
## patient_file_no166	NA	NA	NA	NA
## patient_file_no168	NA	NA	NA	NA
## patient_file_no169	NA	NA	NA	NA
## patient_file_no170	NA	NA	NA	NA
## patient_file_no171	NA	NA	NA	NA
## patient_file_no172	NA	NA	NA	NA
## patient_file_no173	NA	NA	NA	NA
## patient_file_no174	NA	NA	NA	NA
## patient_file_no176	NA	NA	NA	NA
## patient_file_no177	NA	NA	NA	NA

## patient_file_no178	NA	NA	NA	NA
## patient_file_no180	NA	NA	NA	NA
## patient_file_no181	NA	NA	NA	NA
## patient_file_no182	NA	NA	NA	NA
## patient_file_no184	NA	NA	NA	NA
## patient_file_no185	NA	NA	NA	NA
## patient_file_no186	NA	NA	NA	NA
## patient_file_no188	NA	NA	NA	NA
## patient_file_no189	NA	NA	NA	NA
## patient_file_no190	NA	NA	NA	NA
## patient_file_no192	NA	NA	NA	NA
## patient_file_no193	NA	NA	NA	NA
## patient_file_no195	NA	NA	NA	NA
## patient_file_no199	NA	NA	NA	NA
## patient_file_no200	NA	NA	NA	NA
## patient_file_no203	NA	NA	NA	NA
## patient_file_no206	NA	NA	NA	NA
## patient_file_no209	NA	NA	NA	NA
## patient_file_no214	NA	NA	NA	NA
## patient_file_no216	NA	NA	NA	NA
## patient_file_no217	NA	NA	NA	NA
## patient_file_no218	NA	NA	NA	NA
## patient_file_no219	NA	NA	NA	NA
## patient_file_no220	NA	NA	NA	NA
## patient_file_no221	NA	NA	NA	NA
## patient_file_no222	NA	NA	NA	NA
## patient_file_no223	NA	NA	NA	NA
## patient_file_no224	NA	NA	NA	NA
## patient_file_no225	NA	NA	NA	NA
## patient_file_no226	NA	NA	NA	NA
## patient_file_no227	NA	NA	NA	NA
## patient_file_no228	NA	NA	NA	NA
## patient_file_no229	NA	NA	NA	NA
## patient_file_no230	NA	NA	NA	NA
## patient_file_no231	NA	NA	NA	NA
## patient_file_no232	NA	NA	NA	NA
## patient_file_no233	NA	NA	NA	NA
## patient_file_no234	NA	NA	NA	NA
## patient_file_no235	NA	NA	NA	NA
## patient_file_no237	NA	NA	NA	NA
## patient_file_no238	NA	NA	NA	NA
## patient_file_no239	NA	NA	NA	NA
## patient_file_no240	NA	NA	NA	NA
## patient_file_no241	NA	NA	NA	NA
## patient_file_no242	NA	NA	NA	NA
## patient_file_no243	NA	NA	NA	NA
## patient_file_no246	NA	NA	NA	NA
## patient_file_no247	NA	NA	NA	NA
## patient_file_no248	NA	NA	NA	NA
## patient_file_no251	NA	NA	NA	NA
## patient_file_no252	NA	NA	NA	NA
## patient_file_no253	NA	NA	NA	NA
## patient_file_no254	NA	NA	NA	NA
## patient_file_no255	NA	NA	NA	NA

## patient_file_no256	NA	NA	NA	NA
## patient_file_no257	NA	NA	NA	NA
## patient_file_no258	NA	NA	NA	NA
## patient_file_no259	NA	NA	NA	NA
## patient_file_no263	NA	NA	NA	NA
## patient_file_no264	NA	NA	NA	NA
## patient_file_no266	NA	NA	NA	NA
## patient_file_no267	NA	NA	NA	NA
## patient_file_no268	NA	NA	NA	NA
## patient_file_no269	NA	NA	NA	NA
## patient_file_no270	NA	NA	NA	NA
## patient_file_no271	NA	NA	NA	NA
## patient_file_no272	NA	NA	NA	NA
## patient_file_no275	NA	NA	NA	NA
## patient_file_no277	NA	NA	NA	NA
## patient_file_no278	NA	NA	NA	NA
## patient_file_no280	NA	NA	NA	NA
## patient_file_no281	NA	NA	NA	NA
## patient_file_no282	NA	NA	NA	NA
## patient_file_no284	NA	NA	NA	NA
## patient_file_no286	NA	NA	NA	NA
## patient_file_no287	NA	NA	NA	NA
## patient_file_no288	NA	NA	NA	NA
## patient_file_no289	NA	NA	NA	NA
## patient_file_no291	NA	NA	NA	NA
## patient_file_no292	NA	NA	NA	NA
## patient_file_no293	NA	NA	NA	NA
## patient_file_no294	NA	NA	NA	NA
## patient_file_no295	NA	NA	NA	NA
## patient_file_no297	NA	NA	NA	NA
## patient_file_no298	NA	NA	NA	NA
## patient_file_no299	NA	NA	NA	NA
## patient_file_no300	NA	NA	NA	NA
## patient_file_no301	NA	NA	NA	NA
## patient_file_no304	NA	NA	NA	NA
## patient_file_no305	NA	NA	NA	NA
## patient_file_no306	NA	NA	NA	NA
## patient_file_no308	NA	NA	NA	NA
## patient_file_no309	NA	NA	NA	NA
## patient_file_no310	NA	NA	NA	NA
## patient_file_no311	NA	NA	NA	NA
## patient_file_no312	NA	NA	NA	NA
## patient_file_no313	NA	NA	NA	NA
## patient_file_no315	NA	NA	NA	NA
## patient_file_no316	NA	NA	NA	NA
## patient_file_no317	NA	NA	NA	NA
## patient_file_no318	NA	NA	NA	NA
## patient_file_no321	NA	NA	NA	NA
## patient_file_no322	NA	NA	NA	NA
## patient_file_no323	NA	NA	NA	NA
## patient_file_no324	NA	NA	NA	NA
## patient_file_no326	NA	NA	NA	NA
## patient_file_no328	NA	NA	NA	NA
## patient_file_no329	NA	NA	NA	NA

## patient_file_no330	NA	NA	NA	NA
## patient_file_no333	NA	NA	NA	NA
## patient_file_no336	NA	NA	NA	NA
## patient_file_no337	NA	NA	NA	NA
## patient_file_no338	NA	NA	NA	NA
## patient_file_no339	NA	NA	NA	NA
## patient_file_no340	NA	NA	NA	NA
## patient_file_no341	NA	NA	NA	NA
## patient_file_no342	NA	NA	NA	NA
## patient_file_no343	NA	NA	NA	NA
## patient_file_no344	NA	NA	NA	NA
## patient_file_no345	NA	NA	NA	NA
## patient_file_no346	NA	NA	NA	NA
## patient_file_no349	NA	NA	NA	NA
## patient_file_no350	NA	NA	NA	NA
## patient_file_no352	NA	NA	NA	NA
## patient_file_no353	NA	NA	NA	NA
## patient_file_no354	NA	NA	NA	NA
## patient_file_no355	NA	NA	NA	NA
## patient_file_no356	NA	NA	NA	NA
## patient_file_no357	NA	NA	NA	NA
## patient_file_no358	NA	NA	NA	NA
## patient_file_no359	NA	NA	NA	NA
## patient_file_no361	NA	NA	NA	NA
## patient_file_no362	NA	NA	NA	NA
## patient_file_no363	NA	NA	NA	NA
## patient_file_no366	NA	NA	NA	NA
## patient_file_no369	NA	NA	NA	NA
## patient_file_no370	NA	NA	NA	NA
## patient_file_no371	NA	NA	NA	NA
## patient_file_no373	NA	NA	NA	NA
## patient_file_no375	NA	NA	NA	NA
## patient_file_no376	NA	NA	NA	NA
## patient_file_no378	NA	NA	NA	NA
## patient_file_no379	NA	NA	NA	NA
## patient_file_no380	NA	NA	NA	NA
## patient_file_no381	NA	NA	NA	NA
## patient_file_no382	NA	NA	NA	NA
## patient_file_no383	NA	NA	NA	NA
## patient_file_no384	NA	NA	NA	NA
## patient_file_no385	NA	NA	NA	NA
## patient_file_no386	NA	NA	NA	NA
## patient_file_no387	NA	NA	NA	NA
## patient_file_no388	NA	NA	NA	NA
## patient_file_no389	NA	NA	NA	NA
## patient_file_no392	NA	NA	NA	NA
## patient_file_no393	NA	NA	NA	NA
## patient_file_no394	NA	NA	NA	NA
## patient_file_no395	NA	NA	NA	NA
## patient_file_no397	NA	NA	NA	NA
## patient_file_no398	NA	NA	NA	NA
## patient_file_no399	NA	NA	NA	NA
## patient_file_no400	NA	NA	NA	NA
## patient_file_no402	NA	NA	NA	NA

## patient_file_no403	NA	NA	NA	NA
## patient_file_no404	NA	NA	NA	NA
## patient_file_no405	NA	NA	NA	NA
## patient_file_no407	NA	NA	NA	NA
## patient_file_no408	NA	NA	NA	NA
## patient_file_no409	NA	NA	NA	NA
## patient_file_no410	NA	NA	NA	NA
## patient_file_no412	NA	NA	NA	NA
## patient_file_no413	NA	NA	NA	NA
## patient_file_no414	NA	NA	NA	NA
## patient_file_no417	NA	NA	NA	NA
## patient_file_no418	NA	NA	NA	NA
## patient_file_no420	NA	NA	NA	NA
## patient_file_no421	NA	NA	NA	NA
## patient_file_no423	NA	NA	NA	NA
## patient_file_no425	NA	NA	NA	NA
## patient_file_no426	NA	NA	NA	NA
## patient_file_no428	NA	NA	NA	NA
## patient_file_no429	NA	NA	NA	NA
## patient_file_no430	NA	NA	NA	NA
## patient_file_no431	NA	NA	NA	NA
## patient_file_no433	NA	NA	NA	NA
## patient_file_no436	NA	NA	NA	NA
## patient_file_no437	NA	NA	NA	NA
## patient_file_no439	NA	NA	NA	NA
## patient_file_no441	NA	NA	NA	NA
## patient_file_no442	NA	NA	NA	NA
## patient_file_no444	NA	NA	NA	NA
## patient_file_no446	NA	NA	NA	NA
## patient_file_no447	NA	NA	NA	NA
## patient_file_no449	NA	NA	NA	NA
## patient_file_no450	NA	NA	NA	NA
## patient_file_no451	NA	NA	NA	NA
## patient_file_no453	NA	NA	NA	NA
## patient_file_no454	NA	NA	NA	NA
## patient_file_no455	NA	NA	NA	NA
## patient_file_no456	NA	NA	NA	NA
## patient_file_no457	NA	NA	NA	NA
## patient_file_no461	NA	NA	NA	NA
## patient_file_no463	NA	NA	NA	NA
## patient_file_no464	NA	NA	NA	NA
## patient_file_no465	NA	NA	NA	NA
## patient_file_no466	NA	NA	NA	NA
## patient_file_no467	NA	NA	NA	NA
## patient_file_no468	NA	NA	NA	NA
## patient_file_no469	NA	NA	NA	NA
## patient_file_no471	NA	NA	NA	NA
## patient_file_no472	NA	NA	NA	NA
## patient_file_no474	NA	NA	NA	NA
## patient_file_no475	NA	NA	NA	NA
## patient_file_no477	NA	NA	NA	NA
## patient_file_no479	NA	NA	NA	NA
## patient_file_no481	NA	NA	NA	NA
## patient_file_no482	NA	NA	NA	NA

## patient_file_no483	NA	NA	NA	NA
## patient_file_no484	NA	NA	NA	NA
## patient_file_no485	NA	NA	NA	NA
## patient_file_no486	NA	NA	NA	NA
## patient_file_no487	NA	NA	NA	NA
## patient_file_no488	NA	NA	NA	NA
## patient_file_no490	NA	NA	NA	NA
## patient_file_no491	NA	NA	NA	NA
## patient_file_no492	NA	NA	NA	NA
## patient_file_no494	NA	NA	NA	NA
## patient_file_no495	NA	NA	NA	NA
## patient_file_no496	NA	NA	NA	NA
## patient_file_no500	NA	NA	NA	NA
## patient_file_no501	NA	NA	NA	NA
## patient_file_no502	NA	NA	NA	NA
## patient_file_no504	NA	NA	NA	NA
## patient_file_no506	NA	NA	NA	NA
## patient_file_no509	NA	NA	NA	NA
## patient_file_no510	NA	NA	NA	NA
## patient_file_no512	NA	NA	NA	NA
## patient_file_no514	NA	NA	NA	NA
## patient_file_no518	NA	NA	NA	NA
## patient_file_no519	NA	NA	NA	NA
## patient_file_no520	NA	NA	NA	NA
## patient_file_no522	NA	NA	NA	NA
## patient_file_no524	NA	NA	NA	NA
## patient_file_no526	NA	NA	NA	NA
## patient_file_no527	NA	NA	NA	NA
## patient_file_no528	NA	NA	NA	NA
## patient_file_no529	NA	NA	NA	NA
## patient_file_no530	NA	NA	NA	NA
## patient_file_no531	NA	NA	NA	NA
## patient_file_no533	NA	NA	NA	NA
## patient_file_no534	NA	NA	NA	NA
## patient_file_no535	NA	NA	NA	NA
## patient_file_no536	NA	NA	NA	NA
## patient_file_no539	NA	NA	NA	NA
## patient_file_no540	NA	NA	NA	NA
## patient_file_no541	NA	NA	NA	NA
## age_yrs	NA	NA	NA	NA
## weight_kg	NA	NA	NA	NA
## height_cm	NA	NA	NA	NA
## bmi	NA	NA	NA	NA
## blood_group	NA	NA	NA	NA
## pulse_rate_bpm	NA	NA	NA	NA
## rr_breaths_min	NA	NA	NA	NA
## hb_g_dl	NA	NA	NA	NA
## cycle_r_i	NA	NA	NA	NA
## cycle_length_days	NA	NA	NA	NA
## marriage_status_yrs	NA	NA	NA	NA
## pregnant_y_n	NA	NA	NA	NA
## no_of_aborptions	NA	NA	NA	NA
## i_beta_hcg_m_iu_m_l	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l1.65	NA	NA	NA	NA

## ii_beta_hcg_m_iu_m_l11.9	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l11.99	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l11.99.	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l110.0	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l110.4	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l110.84	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l1100.09	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l1100.51	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l1102.3	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l1102.87	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l1108.66	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l11082.82	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l1109.06	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l111.24	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l1110.17	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l11134.4	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l1116.31	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l112.0	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l112.17	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l1121.8	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l11214.23	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l1122.58	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l1127.2	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l113.0	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l113.12	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l11390.58	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l1144.63	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l1145.89	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l11455.0	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l1147.6	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l1148.52	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l115.0	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l1150.91	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l1154.48	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l1155.3	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l1158.51	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l1159.71	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l116069.69	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l1161.49	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l1161.77	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l1167.41	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l1168.99	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l1169.33	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l1173.66	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l1174.37	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l1177.57	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l1177.58	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l118.13	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l118.36	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l118.49	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l1180.3	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l1181.23	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l1183.06	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_l1184.42	NA	NA	NA	NA

## ii_beta_hcg_m_iu_m_1187.79	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_119.44	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_12.0	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_12.8	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_120.0	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1204.69	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_121084.21	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1213.14	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1218.65	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1229.86	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1230.5	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1230.53	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1232.64	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1232.71	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1236.5	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1237.5	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1238.14	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_125000.0	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1255.02	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1268.37	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1272.78	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1277.28	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1278.32	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1278.52	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1288.72	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1297.21	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_13.01	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_13.05	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_13.888	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_13.9	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_13.98	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_13.99	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1300.13	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1320.49	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_13350.19	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_134.65	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1342.91	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1346.59	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1355.28	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1355.51	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1375.18	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1382.36	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1391.46	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_14.2	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_14.32	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_14.76	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_14.96	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1403.85	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1409.85	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_141.75	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_141.77	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1410.13	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_14176.0	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_145.9	NA	NA	NA	NA

## ii_beta_hcg_m_iu_m_1459.63	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1464.12	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1475.04	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1479.66	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_148.86	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1482.21	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1497.41	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_15.39	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_15.81	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1515.53	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1528.5	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_153.82	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_154.08	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1569.1	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1569.3	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_157.08	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1586.06	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1596.2	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_16.19	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_16.921	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1600.23	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_161.98	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1638.52	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_167.2	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_170.42	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1728.01	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1739.13	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1749.98	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_175.51	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_175.62	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1756.61	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1764.83	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1768.03	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1785.95	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_18.0	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_189.34	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1896.6	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_19.83	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_190.67	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1900.6	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_197.63	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_1970.75	NA	NA	NA	NA
## ii_beta_hcg_m_iu_m_199.69	NA	NA	NA	NA
## fsh_m_iu_m_1	NA	NA	NA	NA
## lh_m_iu_m_1	NA	NA	NA	NA
## fsh_lh	NA	NA	NA	NA
## hip_inch	NA	NA	NA	NA
## waist_inch	NA	NA	NA	NA
## waist_hip_ratio	NA	NA	NA	NA
## tsh_m_iu_l	NA	NA	NA	NA
## amh_ng_m_10.2	NA	NA	NA	NA
## amh_ng_m_10.28	NA	NA	NA	NA
## amh_ng_m_10.29	NA	NA	NA	NA
## amh_ng_m_10.3	NA	NA	NA	NA

## amh_ng_m_10.37	NA	NA	NA	NA
## amh_ng_m_10.45	NA	NA	NA	NA
## amh_ng_m_10.5	NA	NA	NA	NA
## amh_ng_m_10.56	NA	NA	NA	NA
## amh_ng_m_10.6	NA	NA	NA	NA
## amh_ng_m_10.69	NA	NA	NA	NA
## amh_ng_m_10.7	NA	NA	NA	NA
## amh_ng_m_10.71	NA	NA	NA	NA
## amh_ng_m_10.72	NA	NA	NA	NA
## amh_ng_m_10.78	NA	NA	NA	NA
## amh_ng_m_10.8	NA	NA	NA	NA
## amh_ng_m_10.85	NA	NA	NA	NA
## amh_ng_m_10.86	NA	NA	NA	NA
## amh_ng_m_10.87	NA	NA	NA	NA
## amh_ng_m_10.88	NA	NA	NA	NA
## amh_ng_m_10.89	NA	NA	NA	NA
## amh_ng_m_10.9	NA	NA	NA	NA
## amh_ng_m_10.91	NA	NA	NA	NA
## amh_ng_m_10.98	NA	NA	NA	NA
## amh_ng_m_10.99	NA	NA	NA	NA
## amh_ng_m_11.0	NA	NA	NA	NA
## amh_ng_m_11.01	NA	NA	NA	NA
## amh_ng_m_11.03	NA	NA	NA	NA
## amh_ng_m_11.04	NA	NA	NA	NA
## amh_ng_m_11.06	NA	NA	NA	NA
## amh_ng_m_11.1	NA	NA	NA	NA
## amh_ng_m_11.15	NA	NA	NA	NA
## amh_ng_m_11.2	NA	NA	NA	NA
## amh_ng_m_11.22	NA	NA	NA	NA
## amh_ng_m_11.25	NA	NA	NA	NA
## amh_ng_m_11.28	NA	NA	NA	NA
## amh_ng_m_11.3	NA	NA	NA	NA
## amh_ng_m_11.4	NA	NA	NA	NA
## amh_ng_m_11.42	NA	NA	NA	NA
## amh_ng_m_11.5	NA	NA	NA	NA
## amh_ng_m_11.54	NA	NA	NA	NA
## amh_ng_m_11.56	NA	NA	NA	NA
## amh_ng_m_11.58	NA	NA	NA	NA
## amh_ng_m_11.6	NA	NA	NA	NA
## amh_ng_m_11.61	NA	NA	NA	NA
## amh_ng_m_11.65	NA	NA	NA	NA
## amh_ng_m_11.67	NA	NA	NA	NA
## amh_ng_m_11.68	NA	NA	NA	NA
## amh_ng_m_11.69	NA	NA	NA	NA
## amh_ng_m_11.7	NA	NA	NA	NA
## amh_ng_m_11.8	NA	NA	NA	NA
## amh_ng_m_11.89	NA	NA	NA	NA
## amh_ng_m_11.9	NA	NA	NA	NA
## amh_ng_m_11.92	NA	NA	NA	NA
## amh_ng_m_110.0	NA	NA	NA	NA
## amh_ng_m_110.04	NA	NA	NA	NA
## amh_ng_m_110.07	NA	NA	NA	NA
## amh_ng_m_110.32	NA	NA	NA	NA
## amh_ng_m_110.53	NA	NA	NA	NA

## amh_ng_m_110.7	NA	NA	NA	NA
## amh_ng_m_110.8	NA	NA	NA	NA
## amh_ng_m_110.9	NA	NA	NA	NA
## amh_ng_m_111.0	NA	NA	NA	NA
## amh_ng_m_111.1	NA	NA	NA	NA
## amh_ng_m_111.2	NA	NA	NA	NA
## amh_ng_m_111.4	NA	NA	NA	NA
## amh_ng_m_111.48	NA	NA	NA	NA
## amh_ng_m_111.6	NA	NA	NA	NA
## amh_ng_m_111.9	NA	NA	NA	NA
## amh_ng_m_112.0	NA	NA	NA	NA
## amh_ng_m_112.7	NA	NA	NA	NA
## amh_ng_m_113.6	NA	NA	NA	NA
## amh_ng_m_114.6	NA	NA	NA	NA
## amh_ng_m_114.7	NA	NA	NA	NA
## amh_ng_m_115.0	NA	NA	NA	NA
## amh_ng_m_115.3	NA	NA	NA	NA
## amh_ng_m_115.7	NA	NA	NA	NA
## amh_ng_m_115.9	NA	NA	NA	NA
## amh_ng_m_116.0	NA	NA	NA	NA
## amh_ng_m_116.6	NA	NA	NA	NA
## amh_ng_m_116.7	NA	NA	NA	NA
## amh_ng_m_116.8	NA	NA	NA	NA
## amh_ng_m_117.0	NA	NA	NA	NA
## amh_ng_m_117.5	NA	NA	NA	NA
## amh_ng_m_117.6	NA	NA	NA	NA
## amh_ng_m_117.9	NA	NA	NA	NA
## amh_ng_m_118.0	NA	NA	NA	NA
## amh_ng_m_118.2	NA	NA	NA	NA
## amh_ng_m_118.4	NA	NA	NA	NA
## amh_ng_m_118.5	NA	NA	NA	NA
## amh_ng_m_118.9	NA	NA	NA	NA
## amh_ng_m_119.0	NA	NA	NA	NA
## amh_ng_m_119.3	NA	NA	NA	NA
## amh_ng_m_119.6	NA	NA	NA	NA
## amh_ng_m_119.8	NA	NA	NA	NA
## amh_ng_m_12.0	NA	NA	NA	NA
## amh_ng_m_12.01	NA	NA	NA	NA
## amh_ng_m_12.04	NA	NA	NA	NA
## amh_ng_m_12.06	NA	NA	NA	NA
## amh_ng_m_12.07	NA	NA	NA	NA
## amh_ng_m_12.1	NA	NA	NA	NA
## amh_ng_m_12.13	NA	NA	NA	NA
## amh_ng_m_12.14	NA	NA	NA	NA
## amh_ng_m_12.2	NA	NA	NA	NA
## amh_ng_m_12.23	NA	NA	NA	NA
## amh_ng_m_12.25	NA	NA	NA	NA
## amh_ng_m_12.26	NA	NA	NA	NA
## amh_ng_m_12.28	NA	NA	NA	NA
## amh_ng_m_12.3	NA	NA	NA	NA
## amh_ng_m_12.31	NA	NA	NA	NA
## amh_ng_m_12.33	NA	NA	NA	NA
## amh_ng_m_12.333	NA	NA	NA	NA
## amh_ng_m_12.34	NA	NA	NA	NA

## amh_ng_m_12.35	NA	NA	NA	NA
## amh_ng_m_12.36	NA	NA	NA	NA
## amh_ng_m_12.38	NA	NA	NA	NA
## amh_ng_m_12.4	NA	NA	NA	NA
## amh_ng_m_12.5	NA	NA	NA	NA
## amh_ng_m_12.53	NA	NA	NA	NA
## amh_ng_m_12.58	NA	NA	NA	NA
## amh_ng_m_12.6	NA	NA	NA	NA
## amh_ng_m_12.65	NA	NA	NA	NA
## amh_ng_m_12.69	NA	NA	NA	NA
## amh_ng_m_12.7	NA	NA	NA	NA
## amh_ng_m_12.78	NA	NA	NA	NA
## amh_ng_m_12.8	NA	NA	NA	NA
## amh_ng_m_12.83	NA	NA	NA	NA
## amh_ng_m_12.85	NA	NA	NA	NA
## amh_ng_m_12.9	NA	NA	NA	NA
## amh_ng_m_120.0	NA	NA	NA	NA
## amh_ng_m_121.0	NA	NA	NA	NA
## amh_ng_m_121.8	NA	NA	NA	NA
## amh_ng_m_121.9	NA	NA	NA	NA
## amh_ng_m_122.0	NA	NA	NA	NA
## amh_ng_m_126.4	NA	NA	NA	NA
## amh_ng_m_126.8	NA	NA	NA	NA
## amh_ng_m_128.6	NA	NA	NA	NA
## amh_ng_m_13.0	NA	NA	NA	NA
## amh_ng_m_13.02	NA	NA	NA	NA
## amh_ng_m_13.03	NA	NA	NA	NA
## amh_ng_m_13.05	NA	NA	NA	NA
## amh_ng_m_13.1	NA	NA	NA	NA
## amh_ng_m_13.14	NA	NA	NA	NA
## amh_ng_m_13.17	NA	NA	NA	NA
## amh_ng_m_13.18	NA	NA	NA	NA
## amh_ng_m_13.2	NA	NA	NA	NA
## amh_ng_m_13.22	NA	NA	NA	NA
## amh_ng_m_13.29	NA	NA	NA	NA
## amh_ng_m_13.3	NA	NA	NA	NA
## amh_ng_m_13.38	NA	NA	NA	NA
## amh_ng_m_13.4	NA	NA	NA	NA
## amh_ng_m_13.49	NA	NA	NA	NA
## amh_ng_m_13.5	NA	NA	NA	NA
## amh_ng_m_13.55	NA	NA	NA	NA
## amh_ng_m_13.56	NA	NA	NA	NA
## amh_ng_m_13.6	NA	NA	NA	NA
## amh_ng_m_13.63	NA	NA	NA	NA
## amh_ng_m_13.64	NA	NA	NA	NA
## amh_ng_m_13.65	NA	NA	NA	NA
## amh_ng_m_13.68	NA	NA	NA	NA
## amh_ng_m_13.7	NA	NA	NA	NA
## amh_ng_m_13.8	NA	NA	NA	NA
## amh_ng_m_13.81	NA	NA	NA	NA
## amh_ng_m_13.86	NA	NA	NA	NA
## amh_ng_m_13.88	NA	NA	NA	NA
## amh_ng_m_13.9	NA	NA	NA	NA
## amh_ng_m_13.91	NA	NA	NA	NA

## amh_ng_m_13.99	NA	NA	NA	NA
## amh_ng_m_132.0	NA	NA	NA	NA
## amh_ng_m_14.02	NA	NA	NA	NA
## amh_ng_m_14.07	NA	NA	NA	NA
## amh_ng_m_14.1	NA	NA	NA	NA
## amh_ng_m_14.13	NA	NA	NA	NA
## amh_ng_m_14.2	NA	NA	NA	NA
## amh_ng_m_14.27	NA	NA	NA	NA
## amh_ng_m_14.3	NA	NA	NA	NA
## amh_ng_m_14.33	NA	NA	NA	NA
## amh_ng_m_14.47	NA	NA	NA	NA
## amh_ng_m_14.5	NA	NA	NA	NA
## amh_ng_m_14.6	NA	NA	NA	NA
## amh_ng_m_14.63	NA	NA	NA	NA
## amh_ng_m_14.66	NA	NA	NA	NA
## amh_ng_m_14.71	NA	NA	NA	NA
## amh_ng_m_14.8	NA	NA	NA	NA
## amh_ng_m_14.9	NA	NA	NA	NA
## amh_ng_m_15.0	NA	NA	NA	NA
## amh_ng_m_15.1	NA	NA	NA	NA
## amh_ng_m_15.2	NA	NA	NA	NA
## amh_ng_m_15.23	NA	NA	NA	NA
## amh_ng_m_15.25	NA	NA	NA	NA
## amh_ng_m_15.4	NA	NA	NA	NA
## amh_ng_m_15.42	NA	NA	NA	NA
## amh_ng_m_15.5	NA	NA	NA	NA
## amh_ng_m_15.6	NA	NA	NA	NA
## amh_ng_m_15.67	NA	NA	NA	NA
## amh_ng_m_15.69	NA	NA	NA	NA
## amh_ng_m_15.7	NA	NA	NA	NA
## amh_ng_m_15.75	NA	NA	NA	NA
## amh_ng_m_15.76	NA	NA	NA	NA
## amh_ng_m_15.78	NA	NA	NA	NA
## amh_ng_m_15.8	NA	NA	NA	NA
## amh_ng_m_15.9	NA	NA	NA	NA
## amh_ng_m_15.96	NA	NA	NA	NA
## amh_ng_m_16.0	NA	NA	NA	NA
## amh_ng_m_16.2	NA	NA	NA	NA
## amh_ng_m_16.26	NA	NA	NA	NA
## amh_ng_m_16.3	NA	NA	NA	NA
## amh_ng_m_16.33	NA	NA	NA	NA
## amh_ng_m_16.41	NA	NA	NA	NA
## amh_ng_m_16.55	NA	NA	NA	NA
## amh_ng_m_16.6	NA	NA	NA	NA
## amh_ng_m_16.8	NA	NA	NA	NA
## amh_ng_m_166.0	NA	NA	NA	NA
## amh_ng_m_17.0	NA	NA	NA	NA
## amh_ng_m_17.2	NA	NA	NA	NA
## amh_ng_m_17.25	NA	NA	NA	NA
## amh_ng_m_17.3	NA	NA	NA	NA
## amh_ng_m_17.51	NA	NA	NA	NA
## amh_ng_m_17.6	NA	NA	NA	NA
## amh_ng_m_17.7	NA	NA	NA	NA
## amh_ng_m_17.8	NA	NA	NA	NA

```
## amh_ng_m_17.9      NA      NA      NA      NA
## amh_ng_m_17.94     NA      NA      NA      NA
## amh_ng_m_18.0      NA      NA      NA      NA
## amh_ng_m_18.1      NA      NA      NA      NA
## amh_ng_m_18.46     NA      NA      NA      NA
## amh_ng_m_18.5      NA      NA      NA      NA
## amh_ng_m_18.75     NA      NA      NA      NA
## amh_ng_m_18.8      NA      NA      NA      NA
## amh_ng_m_18.9      NA      NA      NA      NA
## amh_ng_m_19.0      NA      NA      NA      NA
## amh_ng_m_19.1      NA      NA      NA      NA
## amh_ng_m_19.2      NA      NA      NA      NA
## amh_ng_m_19.7      NA      NA      NA      NA
## amh_ng_m_19.8      NA      NA      NA      NA
## amh_ng_m_19.9      NA      NA      NA      NA
## amh_ng_m_la        NA      NA      NA      NA
## prl_ng_m_l         NA      NA      NA      NA
## vit_d3_ng_m_l      NA      NA      NA      NA
## prg_ng_m_l         NA      NA      NA      NA
## rbs_mg_dl          NA      NA      NA      NA
## weight_gain_y_n1   NA      NA      NA      NA
## hair_growth_y_n1   NA      NA      NA      NA
## skin_darkening_y_n1 NA      NA      NA      NA
## hair_loss_y_n1     NA      NA      NA      NA
## pimples_y_n1       NA      NA      NA      NA
## fast_food_y_n1     NA      NA      NA      NA
## reg_exercise_y_n1  NA      NA      NA      NA
## bp_systolic_mm_hg   NA      NA      NA      NA
## bp_diastolic_mm_hg NA      NA      NA      NA
## follicle_no_l      NA      NA      NA      NA
## follicle_no_r      NA      NA      NA      NA
## avg_f_size_l_mm    NA      NA      NA      NA
## avg_f_size_r_mm    NA      NA      NA      NA
## endometrium_mm     NA      NA      NA      NA
##
## (Dispersion parameter for binomial family taken to be 1)
##
## Null deviance: 4.7918e+02 on 378 degrees of freedom
## Residual deviance: 2.1988e-09 on 0 degrees of freedom
## AIC: 758
##
## Number of Fisher Scoring iterations: 25
```

```
# AIC = 758
```

```
# removed this as not super informative for logistic models
```

```
# plot(logistic_model1)
```

```
library(broom)
```

```
# the second model we will build will only include the following variables:
```

```
# 5 of the hormone measurement levels we possess
```

```
logistic_model2 <- glm(pcos_y_n ~ i_beta_hcg_m_iu_m_l + fsh_m_iu_m_l + lh_m_iu_m_l + tsh_m_iu_l + amh_ng,
  data = train.data,
```

```
family = binomial(link="logit"))
```

```
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
```

```
summary(logistic_model2)
```

```
##
## Call:
## glm(formula = pcos_y_n ~ i_beta_hcg_m_iu_m_l + fsh_m_iu_m_l +
##      lh_m_iu_m_l + tsh_m_iu_l + amh_ng_m_l, family = binomial(link = "logit"),
##      data = train.data)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -1.76433  -0.00005  -0.00005   0.00005   1.73682
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    2.034e+01  1.773e+04   0.001   0.999
## i_beta_hcg_m_iu_m_l -8.195e-05  1.320e-04  -0.621   0.535
## fsh_m_iu_m_l    -3.354e-02  6.423e-02  -0.522   0.602
## lh_m_iu_m_l     8.632e-02  1.391e-01   0.621   0.535
## tsh_m_iu_l     3.321e-02  8.993e-02   0.369   0.712
## amh_ng_m_l0.2   -2.124e+01  1.773e+04  -0.001   0.999
## amh_ng_m_l0.28  -4.096e+01  2.507e+04  -0.002   0.999
## amh_ng_m_l0.29  -4.103e+01  2.507e+04  -0.002   0.999
## amh_ng_m_l0.3   -4.086e+01  2.507e+04  -0.002   0.999
## amh_ng_m_l0.37  -2.031e+01  1.773e+04  -0.001   0.999
## amh_ng_m_l0.45  -4.109e+01  2.507e+04  -0.002   0.999
## amh_ng_m_l0.5   -4.096e+01  2.507e+04  -0.002   0.999
## amh_ng_m_l0.56  -4.115e+01  2.507e+04  -0.002   0.999
## amh_ng_m_l0.6   -4.112e+01  2.170e+04  -0.002   0.998
## amh_ng_m_l0.69  -4.085e+01  2.507e+04  -0.002   0.999
## amh_ng_m_l0.7    2.837e-01  2.507e+04   0.000   1.000
## amh_ng_m_l0.71  -4.091e+01  2.507e+04  -0.002   0.999
## amh_ng_m_l0.72  -4.089e+01  2.507e+04  -0.002   0.999
## amh_ng_m_l0.78    6.351e-02  2.507e+04   0.000   1.000
## amh_ng_m_l0.8   -2.136e+01  1.773e+04  -0.001   0.999
## amh_ng_m_l0.85  -7.247e-02  2.507e+04   0.000   1.000
## amh_ng_m_l0.86  -4.165e+01  2.507e+04  -0.002   0.999
## amh_ng_m_l0.87    1.128e-01  2.507e+04   0.000   1.000
## amh_ng_m_l0.88  -4.103e+01  2.507e+04  -0.002   0.999
## amh_ng_m_l0.89  -2.032e+01  1.773e+04  -0.001   0.999
## amh_ng_m_l0.9   -4.094e+01  2.047e+04  -0.002   0.998
## amh_ng_m_l0.91  -6.639e-02  2.169e+04   0.000   1.000
## amh_ng_m_l0.98  -4.104e+01  2.507e+04  -0.002   0.999
## amh_ng_m_l0.99    3.182e-01  2.507e+04   0.000   1.000
## amh_ng_m_l1.0   -2.077e+01  1.773e+04  -0.001   0.999
## amh_ng_m_l1.01  -2.046e+01  1.773e+04  -0.001   0.999
## amh_ng_m_l1.03  -4.103e+01  2.507e+04  -0.002   0.999
## amh_ng_m_l1.04  -4.121e+01  2.507e+04  -0.002   0.999
## amh_ng_m_l1.06    2.690e-02  2.169e+04   0.000   1.000
```

## amh_ng_m_l1.1	-1.968e+01	1.773e+04	-0.001	0.999
## amh_ng_m_l1.15	2.286e-01	2.507e+04	0.000	1.000
## amh_ng_m_l1.2	-4.106e+01	2.046e+04	-0.002	0.998
## amh_ng_m_l1.22	-4.139e+01	2.507e+04	-0.002	0.999
## amh_ng_m_l1.25	1.490e-01	2.507e+04	0.000	1.000
## amh_ng_m_l1.28	-4.107e+01	2.507e+04	-0.002	0.999
## amh_ng_m_l1.3	-4.113e+01	2.171e+04	-0.002	0.998
## amh_ng_m_l1.4	-2.147e+01	1.773e+04	-0.001	0.999
## amh_ng_m_l1.42	-4.094e+01	2.507e+04	-0.002	0.999
## amh_ng_m_l1.5	-2.043e+01	1.773e+04	-0.001	0.999
## amh_ng_m_l1.54	-4.141e+01	2.507e+04	-0.002	0.999
## amh_ng_m_l1.56	1.499e-01	2.507e+04	0.000	1.000
## amh_ng_m_l1.58	-4.095e+01	2.507e+04	-0.002	0.999
## amh_ng_m_l1.6	-2.114e+01	1.773e+04	-0.001	0.999
## amh_ng_m_l1.61	-4.117e+01	2.507e+04	-0.002	0.999
## amh_ng_m_l1.65	-4.088e+01	2.507e+04	-0.002	0.999
## amh_ng_m_l1.67	-4.103e+01	2.507e+04	-0.002	0.999
## amh_ng_m_l1.68	-4.100e+01	2.507e+04	-0.002	0.999
## amh_ng_m_l1.69	-4.104e+01	2.507e+04	-0.002	0.999
## amh_ng_m_l1.7	-4.107e+01	2.047e+04	-0.002	0.998
## amh_ng_m_l1.8	-2.117e+01	1.773e+04	-0.001	0.999
## amh_ng_m_l1.89	-3.072e-01	2.507e+04	0.000	1.000
## amh_ng_m_l1.9	-4.105e+01	1.982e+04	-0.002	0.998
## amh_ng_m_l1.92	-4.117e+01	2.507e+04	-0.002	0.999
## amh_ng_m_l10.0	-2.035e+01	1.773e+04	-0.001	0.999
## amh_ng_m_l10.04	2.238e-01	2.507e+04	0.000	1.000
## amh_ng_m_l10.07	3.228e-01	2.507e+04	0.000	1.000
## amh_ng_m_l10.32	-4.129e+01	2.507e+04	-0.002	0.999
## amh_ng_m_l10.53	-1.605e-01	2.507e+04	0.000	1.000
## amh_ng_m_l10.7	-4.105e+01	2.507e+04	-0.002	0.999
## amh_ng_m_l10.8	-1.937e+01	1.773e+04	-0.001	0.999
## amh_ng_m_l10.9	-4.108e+01	2.507e+04	-0.002	0.999
## amh_ng_m_l11.0	-4.102e+01	2.171e+04	-0.002	0.998
## amh_ng_m_l11.1	2.305e+00	2.507e+04	0.000	1.000
## amh_ng_m_l11.2	-4.163e+01	2.507e+04	-0.002	0.999
## amh_ng_m_l11.4	-4.093e+01	2.507e+04	-0.002	0.999
## amh_ng_m_l11.48	-4.909e-01	2.507e+04	0.000	1.000
## amh_ng_m_l11.6	-1.308e-01	2.507e+04	0.000	1.000
## amh_ng_m_l11.9	2.221e-02	2.507e+04	0.000	1.000
## amh_ng_m_l12.0	-4.124e+01	2.507e+04	-0.002	0.999
## amh_ng_m_l12.7	1.589e-01	2.507e+04	0.000	1.000
## amh_ng_m_l13.6	1.886e-01	2.507e+04	0.000	1.000
## amh_ng_m_l14.6	-4.096e+01	2.507e+04	-0.002	0.999
## amh_ng_m_l14.7	5.160e-02	2.507e+04	0.000	1.000
## amh_ng_m_l15.0	-5.801e-02	2.171e+04	0.000	1.000
## amh_ng_m_l15.3	-3.978e+01	2.507e+04	-0.002	0.999
## amh_ng_m_l15.7	-8.254e-01	2.507e+04	0.000	1.000
## amh_ng_m_l15.9	-4.582e-01	2.507e+04	0.000	1.000
## amh_ng_m_l16.0	-4.115e+01	2.507e+04	-0.002	0.999
## amh_ng_m_l16.6	3.194e-01	2.507e+04	0.000	1.000
## amh_ng_m_l16.7	-4.095e+01	2.507e+04	-0.002	0.999
## amh_ng_m_l16.8	-2.155e+01	1.773e+04	-0.001	0.999
## amh_ng_m_l17.0	-4.087e+01	2.507e+04	-0.002	0.999
## amh_ng_m_l17.5	-5.258e-02	2.507e+04	0.000	1.000

## amh_ng_m_117.6	-2.584e-01	2.507e+04	0.000	1.000
## amh_ng_m_117.9	-7.344e-02	2.507e+04	0.000	1.000
## amh_ng_m_118.0	8.511e-02	2.507e+04	0.000	1.000
## amh_ng_m_118.2	-4.160e+01	2.507e+04	-0.002	0.999
## amh_ng_m_118.4	1.684e-02	2.507e+04	0.000	1.000
## amh_ng_m_118.5	1.385e-01	2.507e+04	0.000	1.000
## amh_ng_m_118.9	-4.092e+01	2.507e+04	-0.002	0.999
## amh_ng_m_119.0	1.896e-02	2.507e+04	0.000	1.000
## amh_ng_m_119.3	-4.166e+01	2.507e+04	-0.002	0.999
## amh_ng_m_119.6	-5.426e-02	2.507e+04	0.000	1.000
## amh_ng_m_119.8	-1.097e-01	2.507e+04	0.000	1.000
## amh_ng_m_12.0	-4.098e+01	2.170e+04	-0.002	0.998
## amh_ng_m_12.01	3.048e-02	2.507e+04	0.000	1.000
## amh_ng_m_12.04	-4.099e+01	2.507e+04	-0.002	0.999
## amh_ng_m_12.06	-3.942e+01	2.507e+04	-0.002	0.999
## amh_ng_m_12.07	-2.043e+01	1.773e+04	-0.001	0.999
## amh_ng_m_12.1	-4.096e+01	2.171e+04	-0.002	0.998
## amh_ng_m_12.13	-4.112e+01	2.507e+04	-0.002	0.999
## amh_ng_m_12.14	1.082e-01	2.171e+04	0.000	1.000
## amh_ng_m_12.2	-4.105e+01	2.168e+04	-0.002	0.998
## amh_ng_m_12.23	-4.093e+01	2.507e+04	-0.002	0.999
## amh_ng_m_12.25	-4.083e+01	2.507e+04	-0.002	0.999
## amh_ng_m_12.26	-4.091e+01	2.507e+04	-0.002	0.999
## amh_ng_m_12.28	-4.088e+01	2.507e+04	-0.002	0.999
## amh_ng_m_12.3	-4.116e+01	1.978e+04	-0.002	0.998
## amh_ng_m_12.31	-4.111e+01	2.171e+04	-0.002	0.998
## amh_ng_m_12.33	-4.103e+01	2.507e+04	-0.002	0.999
## amh_ng_m_12.333	-4.088e+01	2.507e+04	-0.002	0.999
## amh_ng_m_12.34	-4.091e+01	2.507e+04	-0.002	0.999
## amh_ng_m_12.35	-4.123e+01	2.507e+04	-0.002	0.999
## amh_ng_m_12.36	-4.106e+01	1.982e+04	-0.002	0.998
## amh_ng_m_12.38	-4.095e+01	2.507e+04	-0.002	0.999
## amh_ng_m_12.4	-4.103e+01	2.161e+04	-0.002	0.998
## amh_ng_m_12.5	-4.096e+01	1.880e+04	-0.002	0.998
## amh_ng_m_12.53	-1.627e-01	2.507e+04	0.000	1.000
## amh_ng_m_12.58	-4.083e+01	2.507e+04	-0.002	0.999
## amh_ng_m_12.6	-4.108e+01	2.507e+04	-0.002	0.999
## amh_ng_m_12.65	2.413e-01	2.507e+04	0.000	1.000
## amh_ng_m_12.69	-4.102e+01	2.507e+04	-0.002	0.999
## amh_ng_m_12.7	-2.039e+01	1.773e+04	-0.001	0.999
## amh_ng_m_12.78	-4.096e+01	2.507e+04	-0.002	0.999
## amh_ng_m_12.8	-2.147e+01	1.773e+04	-0.001	0.999
## amh_ng_m_12.83	-2.064e+01	1.773e+04	-0.001	0.999
## amh_ng_m_12.85	-4.099e+01	2.507e+04	-0.002	0.999
## amh_ng_m_12.9	-2.156e+01	1.773e+04	-0.001	0.999
## amh_ng_m_120.0	4.377e-02	2.170e+04	0.000	1.000
## amh_ng_m_121.0	-2.070e+01	1.773e+04	-0.001	0.999
## amh_ng_m_121.8	-5.156e-02	2.507e+04	0.000	1.000
## amh_ng_m_121.9	-3.910e+01	2.507e+04	-0.002	0.999
## amh_ng_m_122.0	7.347e-02	2.507e+04	0.000	1.000
## amh_ng_m_126.4	-4.497e-01	2.171e+04	0.000	1.000
## amh_ng_m_126.8	-4.090e+01	2.507e+04	-0.002	0.999
## amh_ng_m_128.6	-1.560e-01	2.507e+04	0.000	1.000
## amh_ng_m_13.0	-4.100e+01	2.170e+04	-0.002	0.998

## amh_ng_m_13.02	-4.039e+01	2.143e+04	-0.002	0.998
## amh_ng_m_13.03	-4.093e+01	2.507e+04	-0.002	0.999
## amh_ng_m_13.05	-4.089e+01	2.507e+04	-0.002	0.999
## amh_ng_m_13.1	-4.115e+01	2.172e+04	-0.002	0.998
## amh_ng_m_13.14	-4.121e+01	2.507e+04	-0.002	0.999
## amh_ng_m_13.17	-4.105e+01	2.507e+04	-0.002	0.999
## amh_ng_m_13.18	-4.099e+01	2.507e+04	-0.002	0.999
## amh_ng_m_13.2	-2.058e+01	1.773e+04	-0.001	0.999
## amh_ng_m_13.22	-4.098e+01	2.507e+04	-0.002	0.999
## amh_ng_m_13.29	-4.094e+01	2.507e+04	-0.002	0.999
## amh_ng_m_13.3	-4.124e+01	2.025e+04	-0.002	0.998
## amh_ng_m_13.38	-4.099e+01	2.507e+04	-0.002	0.999
## amh_ng_m_13.4	-4.102e+01	2.507e+04	-0.002	0.999
## amh_ng_m_13.49	2.652e-01	2.507e+04	0.000	1.000
## amh_ng_m_13.5	-3.968e+01	1.895e+04	-0.002	0.998
## amh_ng_m_13.55	-2.034e+01	1.773e+04	-0.001	0.999
## amh_ng_m_13.56	-4.100e+01	2.507e+04	-0.002	0.999
## amh_ng_m_13.6	-2.162e+01	1.773e+04	-0.001	0.999
## amh_ng_m_13.63	-4.080e+01	2.507e+04	-0.002	0.999
## amh_ng_m_13.64	-4.081e+01	2.507e+04	-0.002	0.999
## amh_ng_m_13.65	-6.649e-03	2.045e+04	0.000	1.000
## amh_ng_m_13.68	-4.096e+01	2.507e+04	-0.002	0.999
## amh_ng_m_13.7	-4.097e+01	1.982e+04	-0.002	0.998
## amh_ng_m_13.8	-4.102e+01	2.171e+04	-0.002	0.998
## amh_ng_m_13.81	-4.095e+01	2.507e+04	-0.002	0.999
## amh_ng_m_13.86	-4.104e+01	2.170e+04	-0.002	0.998
## amh_ng_m_13.88	-4.090e+01	2.507e+04	-0.002	0.999
## amh_ng_m_13.9	-4.089e+01	1.982e+04	-0.002	0.998
## amh_ng_m_13.91	-4.086e+01	2.507e+04	-0.002	0.999
## amh_ng_m_13.99	-4.109e+01	2.507e+04	-0.002	0.999
## amh_ng_m_132.0	2.211e-01	2.507e+04	0.000	1.000
## amh_ng_m_14.02	-4.060e+01	2.507e+04	-0.002	0.999
## amh_ng_m_14.07	-4.105e+01	2.507e+04	-0.002	0.999
## amh_ng_m_14.1	-4.088e+01	2.507e+04	-0.002	0.999
## amh_ng_m_14.13	-4.105e+01	2.171e+04	-0.002	0.998
## amh_ng_m_14.2	-2.118e+01	1.773e+04	-0.001	0.999
## amh_ng_m_14.27	-4.093e+01	2.507e+04	-0.002	0.999
## amh_ng_m_14.3	-2.117e+01	1.773e+04	-0.001	0.999
## amh_ng_m_14.33	-4.093e+01	2.507e+04	-0.002	0.999
## amh_ng_m_14.47	-4.096e+01	2.507e+04	-0.002	0.999
## amh_ng_m_14.5	-2.050e+01	1.773e+04	-0.001	0.999
## amh_ng_m_14.6	-2.117e+01	1.773e+04	-0.001	0.999
## amh_ng_m_14.63	1.450e-01	2.507e+04	0.000	1.000
## amh_ng_m_14.66	-4.095e+01	2.507e+04	-0.002	0.999
## amh_ng_m_14.71	4.215e-02	2.507e+04	0.000	1.000
## amh_ng_m_14.8	-4.098e+01	1.941e+04	-0.002	0.998
## amh_ng_m_14.9	-4.096e+01	2.170e+04	-0.002	0.998
## amh_ng_m_15.0	-4.110e+01	2.507e+04	-0.002	0.999
## amh_ng_m_15.1	-4.099e+01	2.169e+04	-0.002	0.998
## amh_ng_m_15.2	-4.120e+01	1.940e+04	-0.002	0.998
## amh_ng_m_15.23	-4.117e+01	2.507e+04	-0.002	0.999
## amh_ng_m_15.25	2.473e-01	2.507e+04	0.000	1.000
## amh_ng_m_15.4	-1.985e+01	1.773e+04	-0.001	0.999
## amh_ng_m_15.42	-4.152e+01	2.507e+04	-0.002	0.999

```

## amh_ng_m_15.5      -2.046e+01  1.773e+04 -0.001  0.999
## amh_ng_m_15.6      -4.086e+01  2.507e+04 -0.002  0.999
## amh_ng_m_15.67      1.646e-01  2.507e+04  0.000  1.000
## amh_ng_m_15.69      1.253e-01  2.507e+04  0.000  1.000
## amh_ng_m_15.7       -2.044e+01  1.773e+04 -0.001  0.999
## amh_ng_m_15.75      7.365e-02  2.507e+04  0.000  1.000
## amh_ng_m_15.76     -4.096e+01  2.507e+04 -0.002  0.999
## amh_ng_m_15.78     -4.112e+01  2.507e+04 -0.002  0.999
## amh_ng_m_15.8       -4.108e+01  2.045e+04 -0.002  0.998
## amh_ng_m_15.9       -1.273e-02  2.507e+04  0.000  1.000
## amh_ng_m_15.96     -3.858e+01  2.507e+04 -0.002  0.999
## amh_ng_m_16.0       -2.093e+01  1.773e+04 -0.001  0.999
## amh_ng_m_16.2        1.564e+00  2.137e+04  0.000  1.000
## amh_ng_m_16.26     -4.111e+01  2.507e+04 -0.002  0.999
## amh_ng_m_16.3       -4.100e+01  2.507e+04 -0.002  0.999
## amh_ng_m_16.33     -4.102e+01  2.507e+04 -0.002  0.999
## amh_ng_m_16.41      1.688e-01  2.507e+04  0.000  1.000
## amh_ng_m_16.55     -4.090e+01  2.507e+04 -0.002  0.999
## amh_ng_m_16.6       2.436e-01  2.507e+04  0.000  1.000
## amh_ng_m_16.8       -4.090e+01  2.507e+04 -0.002  0.999
## amh_ng_m_166.0      1.108e-01  2.507e+04  0.000  1.000
## amh_ng_m_17.0       -4.100e+01  2.507e+04 -0.002  0.999
## amh_ng_m_17.2       2.164e-01  2.507e+04  0.000  1.000
## amh_ng_m_17.25     -8.696e-02  2.507e+04  0.000  1.000
## amh_ng_m_17.3       -2.048e+01  1.773e+04 -0.001  0.999
## amh_ng_m_17.51      2.319e-01  2.507e+04  0.000  1.000
## amh_ng_m_17.6       -4.155e+01  2.507e+04 -0.002  0.999
## amh_ng_m_17.7       -2.105e+01  1.773e+04 -0.001  0.999
## amh_ng_m_17.8       -4.108e+01  2.507e+04 -0.002  0.999
## amh_ng_m_17.9       -4.111e+01  2.507e+04 -0.002  0.999
## amh_ng_m_17.94      4.315e-02  2.507e+04  0.000  1.000
## amh_ng_m_18.0       -2.882e-01  2.507e+04  0.000  1.000
## amh_ng_m_18.1       -4.111e+01  2.507e+04 -0.002  0.999
## amh_ng_m_18.46      1.678e-02  2.507e+04  0.000  1.000
## amh_ng_m_18.5       -4.104e+01  2.507e+04 -0.002  0.999
## amh_ng_m_18.75     -4.142e+01  2.507e+04 -0.002  0.999
## amh_ng_m_18.8       -1.580e-01  2.507e+04  0.000  1.000
## amh_ng_m_18.9       -5.404e-02  2.167e+04  0.000  1.000
## amh_ng_m_19.0       -5.904e-02  1.982e+04  0.000  1.000
## amh_ng_m_19.1       -4.109e+01  1.982e+04 -0.002  0.998
## amh_ng_m_19.2       -1.939e-01  2.507e+04  0.000  1.000
## amh_ng_m_19.7       -2.052e+01  1.773e+04 -0.001  0.999
## amh_ng_m_19.8       -4.210e-01  2.507e+04  0.000  1.000
## amh_ng_m_19.9       -4.094e+01  2.507e+04 -0.002  0.999
## amh_ng_m_la        -4.137e+01  2.507e+04 -0.002  0.999
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 479.18  on 378  degrees of freedom
## Residual deviance: 127.28  on 138  degrees of freedom
## AIC: 609.28
##
## Number of Fisher Scoring iterations: 19

```

```
# AIC = 609.28
```

```
glimpse(PCOS_data_without_infertility_formatted_final)
```

```
## Rows: 539
## Columns: 44
## $ sl_no          <fct> 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15~
## $ patient_file_no <fct> 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15~
## $ pcos_y_n        <fct> 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0~
## $ age_yrs         <dbl> 28, 36, 33, 37, 25, 36, 34, 33, 32, 36, 20, 26, 2~
## $ weight_kg        <dbl> 44.6, 65.0, 68.8, 65.0, 52.0, 74.1, 64.0, 58.5, 4~
## $ height_cm        <dbl> 152.0, 161.5, 165.0, 148.0, 161.0, 165.0, 156.0, ~
## $ bmi             <dbl> 19.30000, 24.92116, 25.27089, 29.67495, 20.06095,~
## $ blood_group      <dbl> 15, 15, 11, 13, 11, 15, 11, 13, 11, 15, 15, 13, 1~
## $ pulse_rate_bpm   <dbl> 78, 74, 72, 72, 72, 78, 72, 72, 72, 80, 80, 72, 7~
## $ rr_breaths_min   <dbl> 22, 20, 18, 20, 18, 28, 18, 20, 18, 20, 20, 20, 1~
## $ hb_g_dl          <dbl> 10.48, 11.70, 11.80, 12.00, 10.00, 11.20, 10.90, ~
## $ cycle_r_i        <dbl> 2, 2, 2, 2, 2, 2, 2, 2, 2, 4, 2, 2, 4, 2, 2, 2, 2~
## $ cycle_length_days <dbl> 5, 5, 5, 5, 5, 5, 5, 5, 5, 2, 5, 5, 2, 5, 5, 5, 5~
## $ marriage_status_yrs <dbl> 7, 11, 10, 4, 1, 8, 2, 13, 8, 4, 4, 3, 7, 15, 9, ~
## $ pregnant_y_n     <dbl> 0, 1, 1, 0, 1, 1, 0, 1, 0, 0, 1, 0, 1, 0, 0, 0, 1~
## $ no_of_aborptions <dbl> 0, 0, 0, 0, 0, 0, 0, 2, 1, 0, 2, 1, 0, 0, 0, 0, 0~
## $ i_beta_hcg_m_iu_m_l <dbl> 1.99, 60.80, 494.08, 1.99, 801.45, 237.97, 1.99, ~
## $ ii_beta_hcg_m_iu_m_l <chr> "1.99", "1.99", "494.08", "1.99", "801.45", "1.99~
## $ fsh_m_iu_m_l     <dbl> 7.95, 6.73, 5.54, 8.06, 3.98, 3.24, 2.85, 4.86, 3~
## $ lh_m_iu_m_l      <dbl> 3.68, 1.09, 0.88, 2.36, 0.90, 1.07, 0.31, 3.07, 3~
## $ fsh_lh           <dbl> 2.160326, 6.174312, 6.295455, 3.415254, 4.422222,~
## $ hip_inch         <dbl> 36, 38, 40, 42, 37, 44, 39, 44, 39, 40, 39, 39, 4~
## $ waist_inch       <dbl> 30, 32, 36, 36, 30, 38, 33, 38, 35, 38, 35, 33, 4~
## $ waist_hip_ratio   <dbl> 0.8333333, 0.8421053, 0.9000000, 0.8571429, 0.810~
## $ tsh_m_iu_l       <dbl> 0.68, 3.16, 2.54, 16.41, 3.57, 1.60, 1.51, 12.18,~
## $ amh_ng_m_l       <chr> "2.07", "1.53", "6.63", "1.22", "2.26", "6.74", "~
## $ prl_ng_m_l       <dbl> 45.16, 20.09, 10.52, 36.90, 30.09, 16.18, 26.41, ~
## $ vit_d3_ng_m_l    <dbl> 17.10, 61.30, 49.70, 33.40, 43.80, 52.40, 42.70, ~
## $ prg_ng_m_l       <dbl> 0.57, 0.97, 0.36, 0.36, 0.38, 0.30, 0.46, 0.26, 0~
## $ rbs_mg_dl        <dbl> 92, 92, 84, 76, 84, 76, 93, 91, 116, 125, 108, 10~
## $ weight_gain_y_n   <fct> 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0, 1, 0, 0, 1, 0~
## $ hair_growth_y_n   <fct> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0~
## $ skin_darkening_y_n <fct> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0~
## $ hair_loss_y_n     <fct> 0, 0, 1, 0, 1, 1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0~
## $ pimples_y_n       <fct> 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0~
## $ fast_food_y_n     <fct> 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0~
## $ reg_exercise_y_n  <fct> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0~
## $ bp_systolic_mm_hg <dbl> 110, 120, 120, 120, 120, 120, 110, 120, 120, 120, 110,~
## $ bp_diastolic_mm_hg <dbl> 80, 70, 80, 70, 80, 70, 80, 80, 80, 80, 80, 80, 8~
## $ follicle_no_l     <dbl> 3, 3, 13, 2, 3, 9, 6, 7, 5, 1, 7, 4, 15, 3, 4, 1,~
## $ follicle_no_r     <dbl> 3, 5, 15, 2, 4, 6, 6, 6, 7, 1, 15, 2, 8, 3, 1, 3,~
## $ avg_f_size_l_mm   <dbl> 18, 15, 18, 15, 16, 16, 15, 15, 17, 14, 17, 18, 2~
## $ avg_f_size_r_mm   <dbl> 18, 14, 20, 14, 14, 20, 16, 18, 17, 17, 20, 19, 2~
## $ endometrium_mm    <dbl> 8.5, 3.7, 10.0, 7.5, 7.0, 8.0, 6.8, 7.1, 4.2, 2.5~
```

```
# the third model we will build will include the following variables:
# 5 of the hormone measurement levels we possess
```

as well as the following physiological variables (as these are clinical symptoms associated with PCOS)

```
logistic_model3 <- glm(pcos_y_n ~
  i_beta_hcg_m_iu_m_l +
  fsh_m_iu_m_l +
  lh_m_iu_m_l +
  tsh_m_iu_l +
  amh_ng_m_l +
  weight_gain_y_n +
  hair_growth_y_n +
  skin_darkening_y_n +
  hair_loss_y_n +
  pimples_y_n,
  data = train.data,
  family = binomial(link="logit"))
```

```
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
```

```
summary(logistic_model3)
```

```
##
## Call:
## glm(formula = pcos_y_n ~ i_beta_hcg_m_iu_m_l + fsh_m_iu_m_l +
##   lh_m_iu_m_l + tsh_m_iu_l + amh_ng_m_l + weight_gain_y_n +
##   hair_growth_y_n + skin_darkening_y_n + hair_loss_y_n + pimples_y_n,
##   family = binomial(link = "logit"), data = train.data)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -2.22538  -0.00003  -0.00003   0.00003   2.67333
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    1.477e+01  2.923e+04   0.001  0.99960
## i_beta_hcg_m_iu_m_l -1.241e-04  1.895e-04  -0.655  0.51253
## fsh_m_iu_m_l    -5.410e-02  7.903e-02  -0.685  0.49360
## lh_m_iu_m_l     1.093e-01  1.950e-01   0.561  0.57503
## tsh_m_iu_l      8.233e-02  1.443e-01   0.571  0.56821
## amh_ng_m_l0.2   -2.149e+01  2.923e+04  -0.001  0.99941
## amh_ng_m_l0.28  -3.874e+01  4.134e+04  -0.001  0.99925
## amh_ng_m_l0.29  -3.846e+01  4.134e+04  -0.001  0.99926
## amh_ng_m_l0.3   -3.838e+01  4.134e+04  -0.001  0.99926
## amh_ng_m_l0.37  -1.599e+01  2.923e+04  -0.001  0.99956
## amh_ng_m_l0.45  -4.255e+01  4.134e+04  -0.001  0.99918
## amh_ng_m_l0.5   -3.665e+01  4.134e+04  -0.001  0.99929
## amh_ng_m_l0.56  -3.670e+01  4.134e+04  -0.001  0.99929
## amh_ng_m_l0.6   -3.892e+01  3.573e+04  -0.001  0.99913
## amh_ng_m_l0.69  -3.794e+01  4.134e+04  -0.001  0.99927
## amh_ng_m_l0.7    4.458e-01  4.134e+04   0.000  0.99999
## amh_ng_m_l0.71  -3.626e+01  4.134e+04  -0.001  0.99930
## amh_ng_m_l0.72  -3.641e+01  4.134e+04  -0.001  0.99930
## amh_ng_m_l0.78    4.881e+00  4.134e+04   0.000  0.99991
```

## amh_ng_m_10.8	-1.812e+01	2.923e+04	-0.001	0.99951
## amh_ng_m_10.85	6.194e+00	4.134e+04	0.000	0.99988
## amh_ng_m_10.86	-3.739e+01	4.134e+04	-0.001	0.99928
## amh_ng_m_10.87	6.528e+00	4.134e+04	0.000	0.99987
## amh_ng_m_10.88	-3.904e+01	4.134e+04	-0.001	0.99925
## amh_ng_m_10.89	-1.696e+01	2.923e+04	-0.001	0.99954
## amh_ng_m_10.9	-3.988e+01	3.280e+04	-0.001	0.99903
## amh_ng_m_10.91	1.112e+00	3.511e+04	0.000	0.99997
## amh_ng_m_10.98	-3.654e+01	4.134e+04	-0.001	0.99929
## amh_ng_m_10.99	6.721e+00	4.134e+04	0.000	0.99987
## amh_ng_m_11.0	-1.881e+01	2.923e+04	-0.001	0.99949
## amh_ng_m_11.01	-1.811e+01	2.923e+04	-0.001	0.99951
## amh_ng_m_11.03	-3.650e+01	4.134e+04	-0.001	0.99930
## amh_ng_m_11.04	-3.682e+01	4.134e+04	-0.001	0.99929
## amh_ng_m_11.06	3.852e-01	3.553e+04	0.000	0.99999
## amh_ng_m_11.1	-2.089e+01	2.923e+04	-0.001	0.99943
## amh_ng_m_11.15	-1.738e+00	4.134e+04	0.000	0.99997
## amh_ng_m_11.2	-3.850e+01	3.366e+04	-0.001	0.99909
## amh_ng_m_11.22	-3.751e+01	4.134e+04	-0.001	0.99928
## amh_ng_m_11.25	2.424e+00	4.134e+04	0.000	0.99995
## amh_ng_m_11.28	-3.850e+01	4.134e+04	-0.001	0.99926
## amh_ng_m_11.3	-3.846e+01	3.514e+04	-0.001	0.99913
## amh_ng_m_11.4	-2.027e+01	2.923e+04	-0.001	0.99945
## amh_ng_m_11.42	-3.642e+01	4.134e+04	-0.001	0.99930
## amh_ng_m_11.5	-1.832e+01	2.923e+04	-0.001	0.99950
## amh_ng_m_11.54	-3.914e+01	4.134e+04	-0.001	0.99924
## amh_ng_m_11.56	3.701e-01	4.134e+04	0.000	0.99999
## amh_ng_m_11.58	-3.642e+01	4.134e+04	-0.001	0.99930
## amh_ng_m_11.6	-1.868e+01	2.923e+04	-0.001	0.99949
## amh_ng_m_11.61	-3.689e+01	4.134e+04	-0.001	0.99929
## amh_ng_m_11.65	-3.625e+01	4.134e+04	-0.001	0.99930
## amh_ng_m_11.67	-3.660e+01	4.134e+04	-0.001	0.99929
## amh_ng_m_11.68	-3.901e+01	4.134e+04	-0.001	0.99925
## amh_ng_m_11.69	-3.648e+01	4.134e+04	-0.001	0.99930
## amh_ng_m_11.7	-3.936e+01	3.299e+04	-0.001	0.99905
## amh_ng_m_11.8	-1.915e+01	2.923e+04	-0.001	0.99948
## amh_ng_m_11.89	-2.515e+00	4.134e+04	0.000	0.99995
## amh_ng_m_11.9	-3.836e+01	3.224e+04	-0.001	0.99905
## amh_ng_m_11.92	-3.681e+01	4.134e+04	-0.001	0.99929
## amh_ng_m_110.0	-2.140e+01	2.923e+04	-0.001	0.99942
## amh_ng_m_110.04	-1.741e+00	4.134e+04	0.000	0.99997
## amh_ng_m_110.07	-1.535e+00	4.134e+04	0.000	0.99997
## amh_ng_m_110.32	-4.068e+01	4.134e+04	-0.001	0.99921
## amh_ng_m_110.53	2.063e+00	4.134e+04	0.000	0.99996
## amh_ng_m_110.7	-3.910e+01	4.134e+04	-0.001	0.99925
## amh_ng_m_110.8	-1.884e+01	2.923e+04	-0.001	0.99949
## amh_ng_m_110.9	-3.853e+01	4.134e+04	-0.001	0.99926
## amh_ng_m_111.0	-3.841e+01	3.504e+04	-0.001	0.99913
## amh_ng_m_111.1	1.547e+00	4.134e+04	0.000	0.99997
## amh_ng_m_111.2	-3.724e+01	4.134e+04	-0.001	0.99928
## amh_ng_m_111.4	-4.048e+01	4.134e+04	-0.001	0.99922
## amh_ng_m_111.48	-5.125e-01	4.134e+04	0.000	0.99999
## amh_ng_m_111.6	-5.912e-02	4.134e+04	0.000	1.00000
## amh_ng_m_111.9	1.867e+00	4.134e+04	0.000	0.99996

## amh_ng_m_112.0	-4.085e+01	4.134e+04	-0.001	0.99921
## amh_ng_m_112.7	5.433e-01	4.134e+04	0.000	0.99999
## amh_ng_m_113.6	1.318e-01	4.134e+04	0.000	1.00000
## amh_ng_m_114.6	-3.656e+01	4.134e+04	-0.001	0.99929
## amh_ng_m_114.7	-2.111e+00	4.134e+04	0.000	0.99996
## amh_ng_m_115.0	9.712e-01	3.507e+04	0.000	0.99998
## amh_ng_m_115.3	-3.661e+01	4.134e+04	-0.001	0.99929
## amh_ng_m_115.7	-2.860e+00	4.134e+04	0.000	0.99994
## amh_ng_m_115.9	1.928e+00	4.134e+04	0.000	0.99996
## amh_ng_m_116.0	-3.889e+01	4.134e+04	-0.001	0.99925
## amh_ng_m_116.6	5.181e+00	4.134e+04	0.000	0.99990
## amh_ng_m_116.7	-4.049e+01	4.134e+04	-0.001	0.99922
## amh_ng_m_116.8	-2.158e+01	2.923e+04	-0.001	0.99941
## amh_ng_m_117.0	-4.047e+01	4.134e+04	-0.001	0.99922
## amh_ng_m_117.5	3.664e+00	4.134e+04	0.000	0.99993
## amh_ng_m_117.6	1.767e+00	4.134e+04	0.000	0.99997
## amh_ng_m_117.9	1.163e-01	4.134e+04	0.000	1.00000
## amh_ng_m_118.0	-2.000e+00	4.134e+04	0.000	0.99996
## amh_ng_m_118.2	-3.722e+01	4.134e+04	-0.001	0.99928
## amh_ng_m_118.4	3.766e-02	4.134e+04	0.000	1.00000
## amh_ng_m_118.5	-1.605e+00	4.134e+04	0.000	0.99997
## amh_ng_m_118.9	-3.897e+01	4.134e+04	-0.001	0.99925
## amh_ng_m_119.0	5.881e-01	4.134e+04	0.000	0.99999
## amh_ng_m_119.3	-3.963e+01	4.134e+04	-0.001	0.99924
## amh_ng_m_119.6	4.075e+00	4.134e+04	0.000	0.99992
## amh_ng_m_119.8	2.069e-01	4.134e+04	0.000	1.00000
## amh_ng_m_12.0	-4.013e+01	3.541e+04	-0.001	0.99910
## amh_ng_m_12.01	-1.711e-01	4.134e+04	0.000	1.00000
## amh_ng_m_12.04	-3.897e+01	4.134e+04	-0.001	0.99925
## amh_ng_m_12.06	-3.666e+01	4.134e+04	-0.001	0.99929
## amh_ng_m_12.07	-1.828e+01	2.923e+04	-0.001	0.99950
## amh_ng_m_12.1	-3.923e+01	3.412e+04	-0.001	0.99908
## amh_ng_m_12.13	-3.678e+01	4.134e+04	-0.001	0.99929
## amh_ng_m_12.14	1.198e+00	3.429e+04	0.000	0.99997
## amh_ng_m_12.2	-3.847e+01	3.509e+04	-0.001	0.99913
## amh_ng_m_12.23	-4.070e+01	4.134e+04	-0.001	0.99921
## amh_ng_m_12.25	-3.882e+01	4.134e+04	-0.001	0.99925
## amh_ng_m_12.26	-3.639e+01	4.134e+04	-0.001	0.99930
## amh_ng_m_12.28	-4.261e+01	4.134e+04	-0.001	0.99918
## amh_ng_m_12.3	-4.318e+01	3.110e+04	-0.001	0.99889
## amh_ng_m_12.31	-3.796e+01	3.528e+04	-0.001	0.99914
## amh_ng_m_12.33	-3.665e+01	4.134e+04	-0.001	0.99929
## amh_ng_m_12.333	-3.632e+01	4.134e+04	-0.001	0.99930
## amh_ng_m_12.34	-3.638e+01	4.134e+04	-0.001	0.99930
## amh_ng_m_12.35	-3.685e+01	4.134e+04	-0.001	0.99929
## amh_ng_m_12.36	-3.804e+01	3.212e+04	-0.001	0.99906
## amh_ng_m_12.38	-4.283e+01	4.134e+04	-0.001	0.99917
## amh_ng_m_12.4	-3.764e+01	3.573e+04	-0.001	0.99916
## amh_ng_m_12.5	-3.966e+01	3.075e+04	-0.001	0.99897
## amh_ng_m_12.53	2.419e+00	4.134e+04	0.000	0.99995
## amh_ng_m_12.58	-3.623e+01	4.134e+04	-0.001	0.99930
## amh_ng_m_12.6	-4.101e+01	4.134e+04	-0.001	0.99921
## amh_ng_m_12.65	5.496e-02	4.134e+04	0.000	1.00000
## amh_ng_m_12.69	-4.057e+01	4.134e+04	-0.001	0.99922

## amh_ng_m_12.7	-1.568e+01	2.923e+04	-0.001	0.99957
## amh_ng_m_12.78	-3.650e+01	4.134e+04	-0.001	0.99930
## amh_ng_m_12.8	-2.228e+01	2.923e+04	-0.001	0.99939
## amh_ng_m_12.83	-2.105e+01	2.923e+04	-0.001	0.99943
## amh_ng_m_12.85	-3.648e+01	4.134e+04	-0.001	0.99930
## amh_ng_m_12.9	-2.044e+01	2.923e+04	-0.001	0.99944
## amh_ng_m_120.0	-4.925e-01	3.529e+04	0.000	0.99999
## amh_ng_m_121.0	-2.153e+01	2.923e+04	-0.001	0.99941
## amh_ng_m_121.8	-1.884e+00	4.134e+04	0.000	0.99996
## amh_ng_m_121.9	-3.584e+01	4.134e+04	-0.001	0.99931
## amh_ng_m_122.0	1.660e-01	4.134e+04	0.000	1.00000
## amh_ng_m_126.4	-3.648e-01	3.561e+04	0.000	0.99999
## amh_ng_m_126.8	-3.660e+01	4.134e+04	-0.001	0.99929
## amh_ng_m_128.6	-1.280e-01	4.134e+04	0.000	1.00000
## amh_ng_m_13.0	-4.010e+01	3.515e+04	-0.001	0.99909
## amh_ng_m_13.02	-3.944e+01	3.343e+04	-0.001	0.99906
## amh_ng_m_13.03	-4.051e+01	4.134e+04	-0.001	0.99922
## amh_ng_m_13.05	-3.634e+01	4.134e+04	-0.001	0.99930
## amh_ng_m_13.1	-4.043e+01	3.494e+04	-0.001	0.99908
## amh_ng_m_13.14	-3.950e+01	4.134e+04	-0.001	0.99924
## amh_ng_m_13.17	-3.907e+01	4.134e+04	-0.001	0.99925
## amh_ng_m_13.18	-3.647e+01	4.134e+04	-0.001	0.99930
## amh_ng_m_13.2	-1.944e+01	2.923e+04	-0.001	0.99947
## amh_ng_m_13.22	-3.651e+01	4.134e+04	-0.001	0.99930
## amh_ng_m_13.29	-3.893e+01	4.134e+04	-0.001	0.99925
## amh_ng_m_13.3	-3.741e+01	3.327e+04	-0.001	0.99910
## amh_ng_m_13.38	-4.069e+01	4.134e+04	-0.001	0.99921
## amh_ng_m_13.4	-3.664e+01	4.134e+04	-0.001	0.99929
## amh_ng_m_13.49	6.975e-01	4.134e+04	0.000	0.99999
## amh_ng_m_13.5	-4.101e+01	3.098e+04	-0.001	0.99894
## amh_ng_m_13.55	-1.912e+01	2.923e+04	-0.001	0.99948
## amh_ng_m_13.56	-3.878e+01	4.134e+04	-0.001	0.99925
## amh_ng_m_13.6	-1.935e+01	2.923e+04	-0.001	0.99947
## amh_ng_m_13.63	-4.074e+01	4.134e+04	-0.001	0.99921
## amh_ng_m_13.64	-3.623e+01	4.134e+04	-0.001	0.99930
## amh_ng_m_13.65	5.286e+00	3.214e+04	0.000	0.99987
## amh_ng_m_13.68	-3.640e+01	4.134e+04	-0.001	0.99930
## amh_ng_m_13.7	-3.761e+01	3.224e+04	-0.001	0.99907
## amh_ng_m_13.8	-3.801e+01	3.507e+04	-0.001	0.99914
## amh_ng_m_13.81	-3.647e+01	4.134e+04	-0.001	0.99930
## amh_ng_m_13.86	-3.828e+01	3.504e+04	-0.001	0.99913
## amh_ng_m_13.88	-3.628e+01	4.134e+04	-0.001	0.99930
## amh_ng_m_13.9	-4.297e+01	3.131e+04	-0.001	0.99890
## amh_ng_m_13.91	-3.629e+01	4.134e+04	-0.001	0.99930
## amh_ng_m_13.99	-3.889e+01	4.134e+04	-0.001	0.99925
## amh_ng_m_132.0	2.704e+00	4.134e+04	0.000	0.99995
## amh_ng_m_14.02	-3.819e+01	4.134e+04	-0.001	0.99926
## amh_ng_m_14.07	-3.653e+01	4.134e+04	-0.001	0.99929
## amh_ng_m_14.1	-4.479e+01	4.134e+04	-0.001	0.99914
## amh_ng_m_14.13	-3.658e+01	3.577e+04	-0.001	0.99918
## amh_ng_m_14.2	-2.262e+01	2.923e+04	-0.001	0.99938
## amh_ng_m_14.27	-3.646e+01	4.134e+04	-0.001	0.99930
## amh_ng_m_14.3	-1.855e+01	2.923e+04	-0.001	0.99949
## amh_ng_m_14.33	-3.642e+01	4.134e+04	-0.001	0.99930

## amh_ng_m_14.47	-3.640e+01	4.134e+04	-0.001	0.99930
## amh_ng_m_14.5	-2.030e+01	2.923e+04	-0.001	0.99945
## amh_ng_m_14.6	-1.968e+01	2.923e+04	-0.001	0.99946
## amh_ng_m_14.63	3.730e-01	4.134e+04	0.000	0.99999
## amh_ng_m_14.66	-3.873e+01	4.134e+04	-0.001	0.99925
## amh_ng_m_14.71	-1.879e+00	4.134e+04	0.000	0.99996
## amh_ng_m_14.8	-4.004e+01	3.140e+04	-0.001	0.99898
## amh_ng_m_14.9	-3.860e+01	3.552e+04	-0.001	0.99913
## amh_ng_m_15.0	-4.087e+01	4.134e+04	-0.001	0.99921
## amh_ng_m_15.1	-4.201e+01	3.534e+04	-0.001	0.99905
## amh_ng_m_15.2	-4.090e+01	3.129e+04	-0.001	0.99896
## amh_ng_m_15.23	-3.675e+01	4.134e+04	-0.001	0.99929
## amh_ng_m_15.25	-1.485e+00	4.134e+04	0.000	0.99997
## amh_ng_m_15.4	-1.796e+01	2.923e+04	-0.001	0.99951
## amh_ng_m_15.42	-3.720e+01	4.134e+04	-0.001	0.99928
## amh_ng_m_15.5	-1.805e+01	2.923e+04	-0.001	0.99951
## amh_ng_m_15.6	-4.037e+01	4.134e+04	-0.001	0.99922
## amh_ng_m_15.67	2.635e-01	4.134e+04	0.000	0.99999
## amh_ng_m_15.69	2.847e-01	4.134e+04	0.000	0.99999
## amh_ng_m_15.7	-1.921e+01	2.923e+04	-0.001	0.99948
## amh_ng_m_15.75	2.264e+00	4.134e+04	0.000	0.99996
## amh_ng_m_15.76	-4.088e+01	4.134e+04	-0.001	0.99921
## amh_ng_m_15.78	-3.920e+01	4.134e+04	-0.001	0.99924
## amh_ng_m_15.8	-3.967e+01	3.223e+04	-0.001	0.99902
## amh_ng_m_15.9	2.878e-01	4.134e+04	0.000	0.99999
## amh_ng_m_15.96	-3.703e+01	4.134e+04	-0.001	0.99929
## amh_ng_m_16.0	-1.972e+01	2.923e+04	-0.001	0.99946
## amh_ng_m_16.2	3.056e+00	3.544e+04	0.000	0.99993
## amh_ng_m_16.26	-3.892e+01	4.134e+04	-0.001	0.99925
## amh_ng_m_16.3	-3.637e+01	4.134e+04	-0.001	0.99930
## amh_ng_m_16.33	-3.916e+01	4.134e+04	-0.001	0.99924
## amh_ng_m_16.41	-1.866e+00	4.134e+04	0.000	0.99996
## amh_ng_m_16.55	-3.632e+01	4.134e+04	-0.001	0.99930
## amh_ng_m_16.6	-1.752e+00	4.134e+04	0.000	0.99997
## amh_ng_m_16.8	-3.655e+01	4.134e+04	-0.001	0.99929
## amh_ng_m_166.0	2.260e+00	4.134e+04	0.000	0.99996
## amh_ng_m_17.0	-3.644e+01	4.134e+04	-0.001	0.99930
## amh_ng_m_17.2	2.001e+00	4.134e+04	0.000	0.99996
## amh_ng_m_17.25	1.328e-01	4.134e+04	0.000	1.00000
## amh_ng_m_17.3	-1.835e+01	2.923e+04	-0.001	0.99950
## amh_ng_m_17.51	2.742e+00	4.134e+04	0.000	0.99995
## amh_ng_m_17.6	-3.775e+01	4.134e+04	-0.001	0.99927
## amh_ng_m_17.7	-1.909e+01	2.923e+04	-0.001	0.99948
## amh_ng_m_17.8	-3.673e+01	4.134e+04	-0.001	0.99929
## amh_ng_m_17.9	-3.857e+01	4.134e+04	-0.001	0.99926
## amh_ng_m_17.94	-2.175e+00	4.134e+04	0.000	0.99996
## amh_ng_m_18.0	-3.295e-01	4.134e+04	0.000	0.99999
## amh_ng_m_18.1	-3.682e+01	4.134e+04	-0.001	0.99929
## amh_ng_m_18.46	6.405e+00	4.134e+04	0.000	0.99988
## amh_ng_m_18.5	-3.665e+01	4.134e+04	-0.001	0.99929
## amh_ng_m_18.75	-4.148e+01	4.134e+04	-0.001	0.99920
## amh_ng_m_18.8	3.186e-01	4.134e+04	0.000	0.99999
## amh_ng_m_18.9	2.054e+00	3.510e+04	0.000	0.99995
## amh_ng_m_19.0	2.569e+00	3.191e+04	0.000	0.99994

```
## amh_ng_m_l9.1      -4.127e+01  3.218e+04 -0.001  0.99898
## amh_ng_m_l9.2      -2.138e-01  4.134e+04  0.000  1.00000
## amh_ng_m_l9.7      -1.953e+01  2.923e+04 -0.001  0.99947
## amh_ng_m_l9.8       1.344e+00  4.134e+04  0.000  0.99997
## amh_ng_m_l9.9      -3.659e+01  4.134e+04 -0.001  0.99929
## amh_ng_m_la        -3.990e+01  4.134e+04 -0.001  0.99923
## weight_gain_y_n1    1.745e+00  8.464e-01  2.061  0.03928 *
## hair_growth_y_n1    2.083e+00  9.792e-01  2.127  0.03341 *
## skin_darkening_y_n1 2.125e+00  7.923e-01  2.682  0.00731 **
## hair_loss_y_n1      2.072e-01  8.679e-01  0.239  0.81126
## pimples_y_n1        2.371e+00  9.350e-01  2.536  0.01123 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 479.178  on 378  degrees of freedom
## Residual deviance:  74.445  on 133  degrees of freedom
## AIC: 566.45
##
## Number of Fisher Scoring iterations: 20
```

```
# AIC = 566.45
```

3.3 Model cross-validation (without RIDGE or LASSO) (using the caret package)

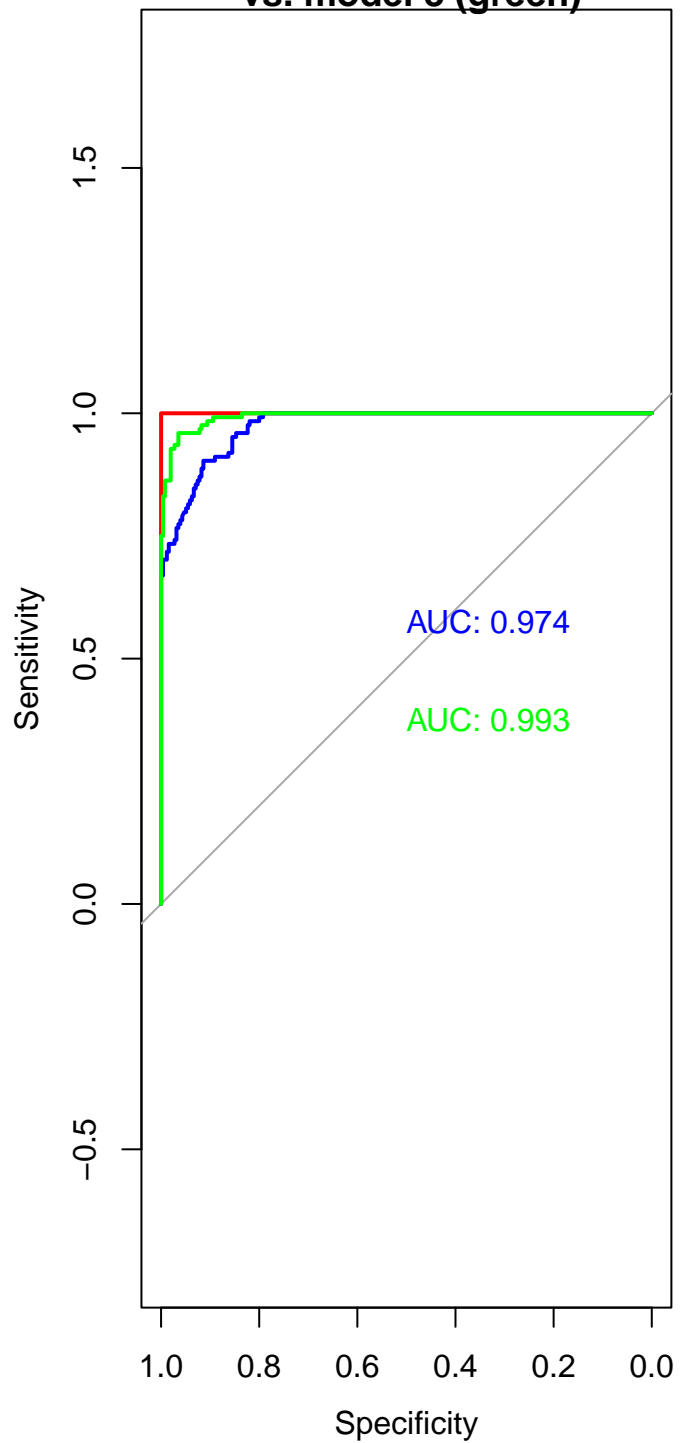
- We performed model cross-validation on our three logistic regression models and generated a summary of the cross-validation results.

```
##           Min.    1st Qu.    Median    Mean    3rd Qu.    Max. NA's
## model1 0.3289474 0.6710526 0.6710526 0.6044211 0.6710526 0.6800000    0
## model2 0.4605263 0.5000000 0.5263158 0.5383626 0.5827193 0.6266667    0
## model3 0.5000000 0.5197368 0.5657895 0.5886784 0.6093860 0.7763158    0
```

3.4 Assessing model classification performance

- Assessing model performance using ROC curves

**ROC curves: model 1 (red) vs. model 2 (blue)
vs. model 3 (green)**



Penalized logistic regression with cross-validation (RIDGE and LASSO)

Exhaustive searches (may not include?)

```
# library(bestglm)

#train.data.covariates.1 <- train.data %>%
# select(
#   pcos_y_n,
#   i_beta_hcg_m_iu_m_l,
#   fsh_m_iu_m_l,
#   lh_m_iu_m_l,
#   tsh_m_iu_l,
#   amh_ng_m_l ,
#   weight_gain_y_n ,
#   hair_growth_y_n,
#   skin_darkening_y_n,
#   hair_loss_y_n ,
#   pimples_y_n
# ) %>%
# mutate_at(c("pcos_y_n",
#             "weight_gain_y_n",
#             "hair_growth_y_n",
#             "skin_darkening_y_n",
#             "hair_loss_y_n",
#             "pimples_y_n"),as.factor) %>%
# mutate_at(c(
#   "i_beta_hcg_m_iu_m_l",
#   "fsh_m_iu_m_l" ,
#   "lh_m_iu_m_l" ,
#   "tsh_m_iu_l" ,
#   "amh_ng_m_l"), as.numeric)
#
# x_tr <- train.data.covariates.1
# # x_tr$pcos_labelled <- ifelse(x_tr$pcos_labelled=="yes",1,0)
# colnames(x_tr)[1] <- "y"
# res.bestglm <-bestglm::bestglm(Xy = x_tr,
#   family = binomial,
#   IC = "AIC",           # Information criteria for
#   method = "exhaustive")
#
# ## Show top 5 models
# res.bestglm$BestModels
#
# summary(res.bestglm$BestModel)
```

1.4 Modelling our data using Trees and Forests

Let's select only the predictor columns we decided we wanted to include earlier. From logistic model 3, here is the list of covariates: * pcos_y_n * i_beta_hcg_m_iu_m_l * fsh_m_iu_m_l * lh_m_iu_m_l * tsh_m_iu_l * amh_ng_m_l * weight_gain_y_n * hair_growth_y_n * skin_darkening_y_n * hair_loss_y_n * pimples_y_n

```

train.data.covariates <- train.data %>%
  select(
    pcos_y_n,
    i_beta_hcg_m_iu_m_l,
    fsh_m_iu_m_l ,
    lh_m_iu_m_l ,
    tsh_m_iu_l ,
    amh_ng_m_l ,
    weight_gain_y_n ,
    hair_growth_y_n,
    skin_darkening_y_n,
    hair_loss_y_n ,
    pimples_y_n
  ) %>% mutate(
    amh_ng_m_l = as.numeric(amh_ng_m_l),
    pcos_labelled = as.factor(ifelse(pcos_y_n==1, "yes", "no")))

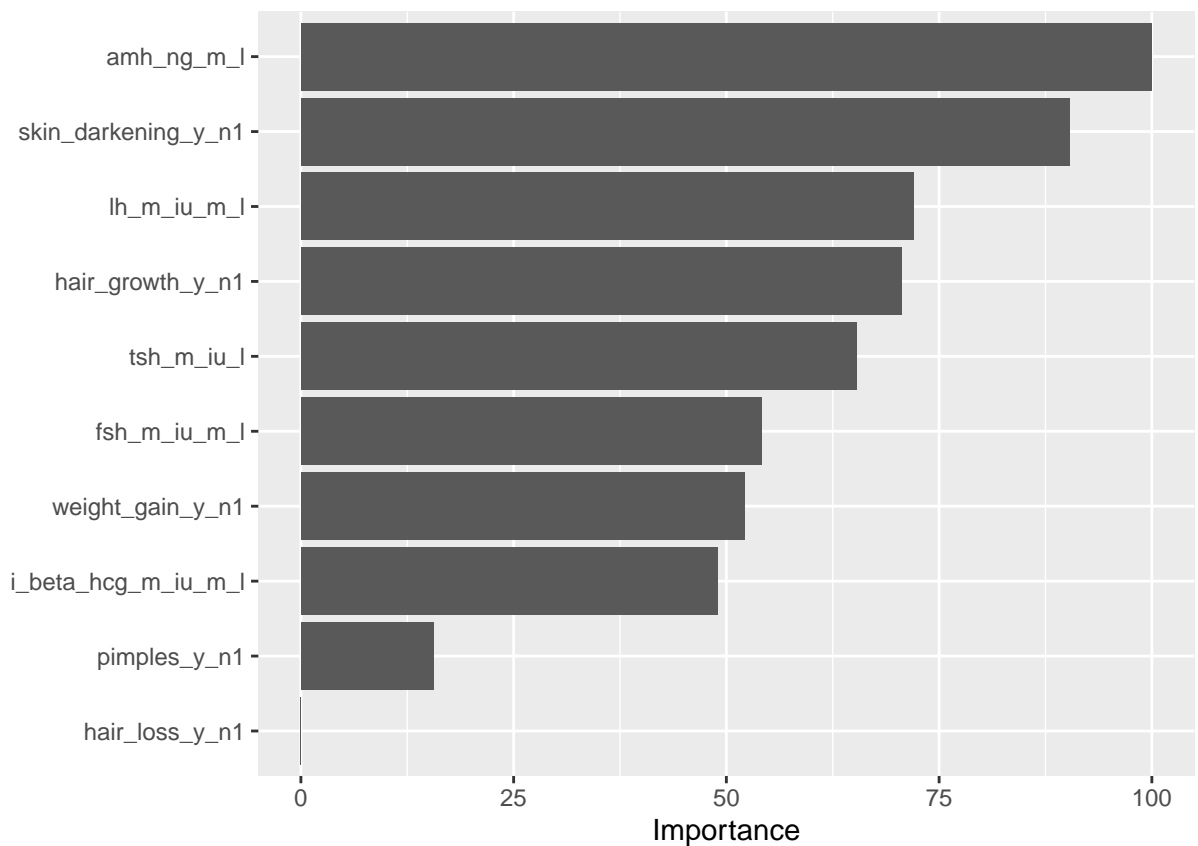
```

```
## Warning in mask$eval_all_mutate(quo): NAs introduced by coercion
```

Random Forest – note that I've had to remove one missing ob from amh_ng_m_l for now.

RF Variable Importance

```
vip::vip(caret_rf)
```



XGboost

```
set.seed(123)
xgb_grid_1 <- expand.grid(
  nrounds = 50,
  eta = c(0.03),
  max_depth = 1,
  gamma = 0,
  colsample_bytree = 0.6,
  min_child_weight = 1,
  subsample = 0.5
)

caret_xgb <- caret::train(pcos_labelled ~., data = select(train.data.covariates, -(pcos_y_n)),
  method = "xgbTree",
  metric = "ROC",
  tuneGrid=xgb_grid_1,
  na.action = na.pass,
  trControl = trainControl(method = "cv", number = 5, classProbs = T, summaryFunc=summary2)

caret_xgb
```

```
## eXtreme Gradient Boosting
##
## 379 samples
## 10 predictor
## 2 classes: 'no', 'yes'
##
## No pre-processing
## Resampling: Cross-Validated (5 fold)
## Summary of sample sizes: 303, 304, 303, 303, 303
## Resampling results:
##
##   ROC      Sens      Spec
## 0.8800556 0.9372549 0.525
##
## Tuning parameter 'nrounds' was held constant at a value of 50
## Tuning
## held constant at a value of 1
## Tuning parameter 'subsample' was held
## constant at a value of 0.5
```

Comparison code for later

5. Discussion and Conclusions

5.1 Discussion

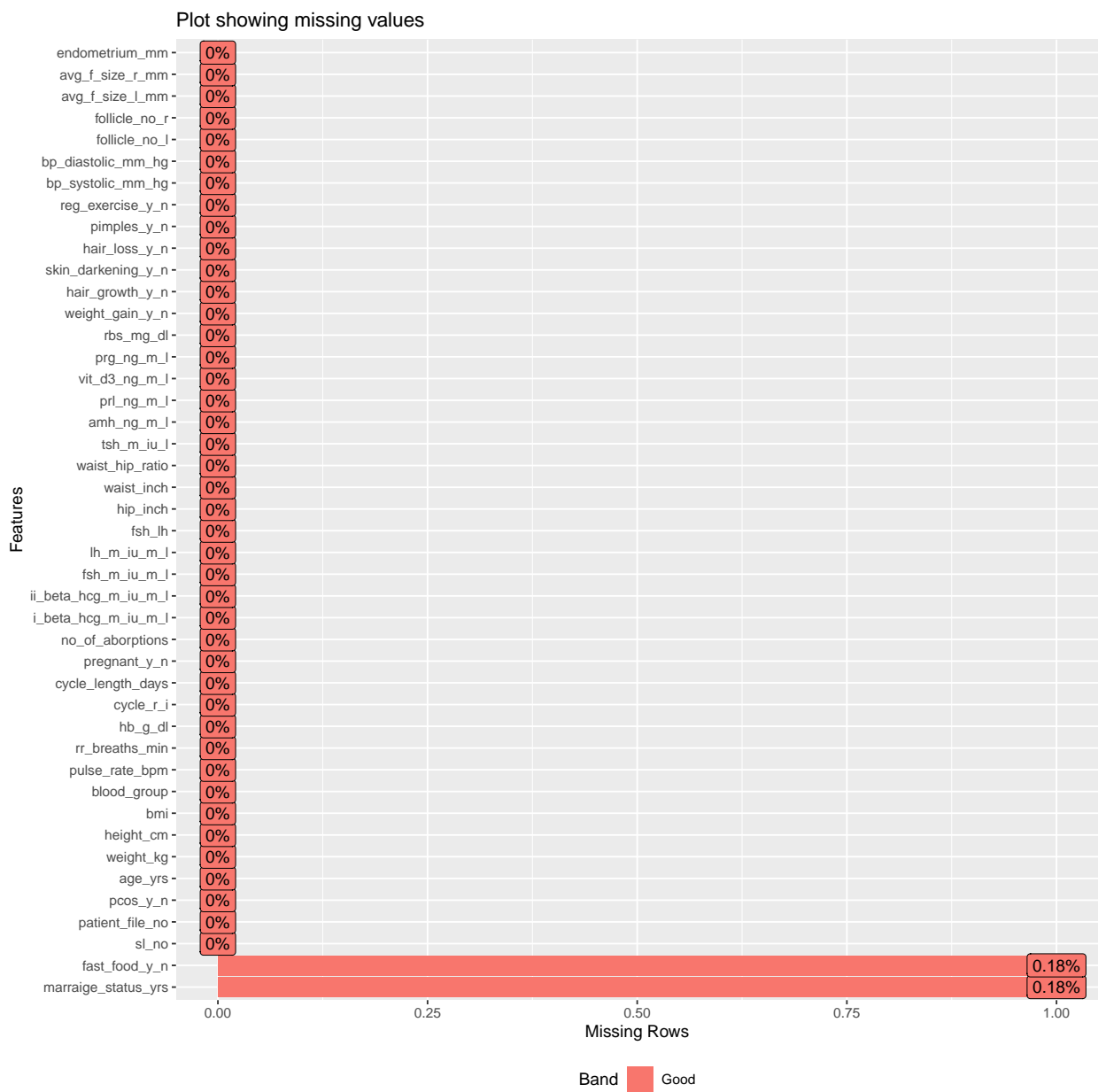
- Ethics statement

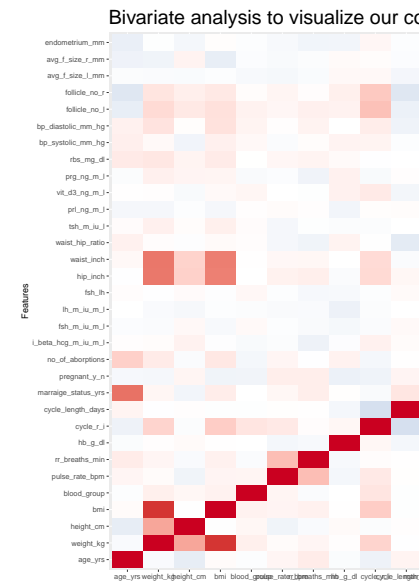
References

1. R for Data Science by Hadley Wickham (<https://r4ds.had.co.nz>)
2. Ajmal N, Khan SZ, Shaikh R. Polycystic ovary syndrome (PCOS) and genetic predisposition: A review article. Eur J Obstet Gynecol Reprod Biol X. 2019 Jun 8;3:100060. doi: 10.1016/j.eurox.2019.100060. PMID: 31403134; PMCID: PMC6687436.

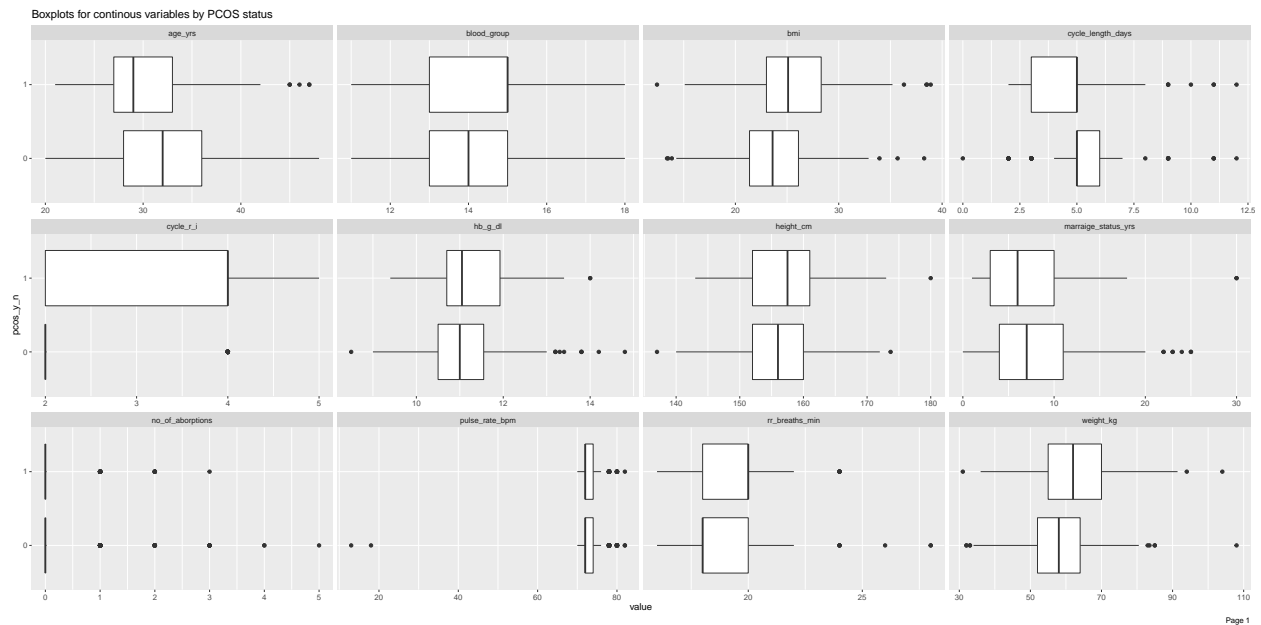
Appendix

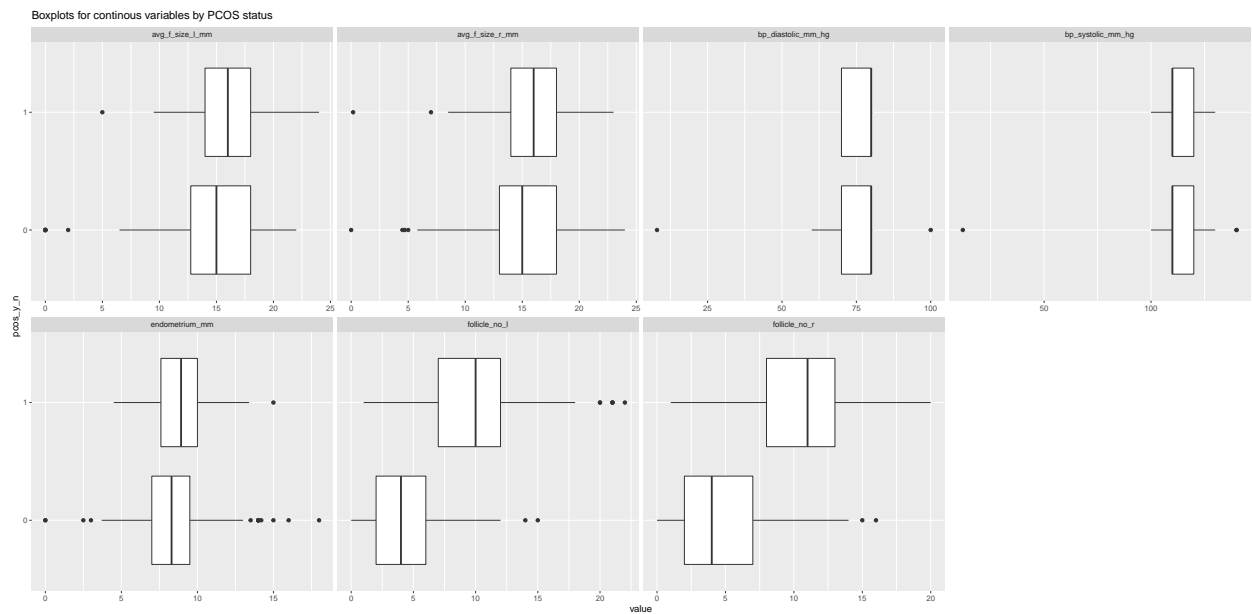
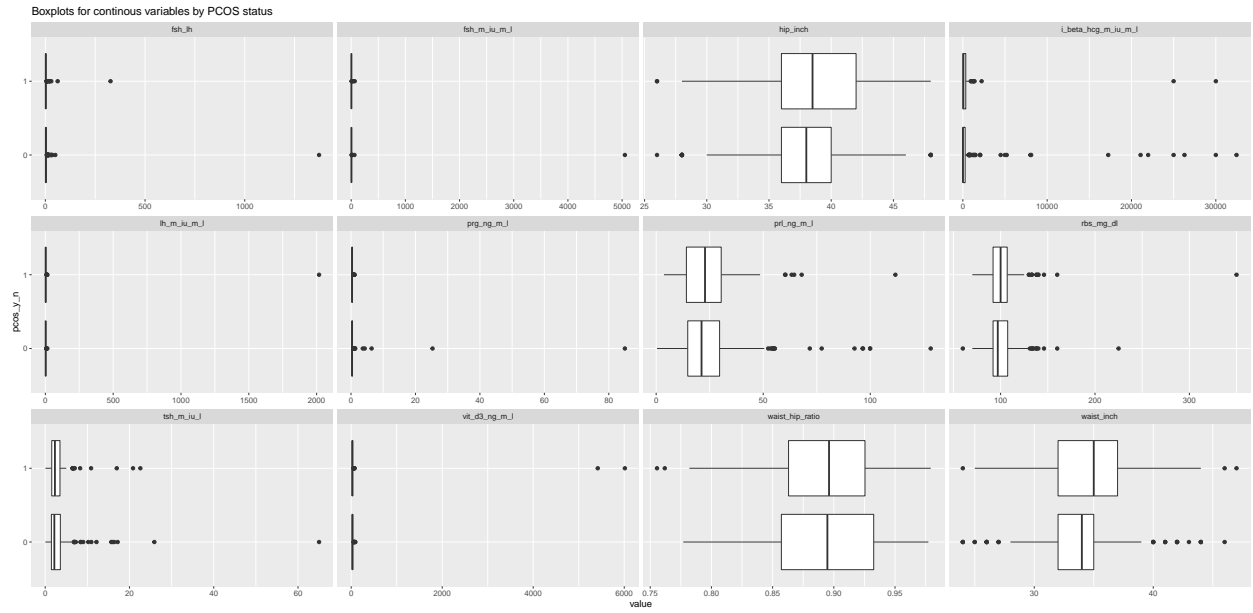
- Appendix Plot 1, as part of EDA, showing missing values in our dataset





- Appendix plot 2, as part of EDA, examining correlations between continuous features
- From this correlation plot, we find that several continuous variables do co-vary with one another.
- Specifically, as we would expect, we find a positive correlation between the variables waist and hip (in inches) with weight. We find the same positive correlation for BMI.
- Another obvious correlation we observe is that between age (in years) and marriage (in years)
-
- Appendix Plot 3 showing boxplots generated to investigate potential associations between continuous variables and PCOS status.





4a.1 Missing Values

Next in our EDA we sought to identify missing values in our dataset. Here, we see a plot showing our variables and the percentage of missing rows per variable. The plot shows the missing values in our dataset. Our EDA identified some missing values for two variables: fast food and marriage status. The analysis indicates that only 0.18% of the rows for these variables are missing. Therefore, as this is below the generally used threshold of 5 %, we will simply ignore these missing values for our subsequent analyses.

4a.2 Univariable distributions for continuous variables

To get a sense of the variation in our dataset, we plotted histograms for each continuous variable using the `plot_histogram()` from the DataExplorer Package.

We see that the age distribution in our dataset reveals most individuals are in the range of 20 to 40 years. No individuals in the dataset are younger than 20 or older than 50. As we would expect, the BMI values follow an approximately normal distribution. The most common blood type we observe is O+, which is consistent with the fact that O+ is the most frequent bloodtype globally. The most common cycle length is 5 days. The endometrium thickness data suggests there are two most common thickness values (two clear peaks in the distribution).