



Natural Cycles

Assignment for Senior Data Scientist position

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Jeroen Buil

Contents

- **Introduction**
- **Exploratory Data Analysis**
- **Questions:**
 - 1. What is the chance of getting pregnant within 13 cycles?*
 - 2. How long does it usually take to get pregnant?*
 - 3. What factors impact the time it takes to get pregnant?*
 - 4. ML vs non-ML methods approach?*

Quick introduction



- **Jeroen Buil**

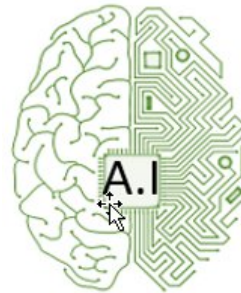
- Senior Data Scientist / Biomedical Engineer

Focus:
Practically applying AI for

Data
Collection



Models



Insights



2013



2024



EDA: Exploratory Data Analysis

=> First step before any analysis

- Q: Is the data suitable?



EDA: Quick glance

- Missing data:

- ~40% of samples miss some data points
- => Need to remove (or ideally fill) depending on analysis

```
NaN count:
bmi          0
age          0
country      113
been_pregnant_before  317
education    391
sleeping_pattern  499
n_cycles_trying  0
outcome      0
dedication   0
average_cycle_length  6
cycle_length_std  25
regular_cycle  6
intercourse_frequency  0
```

- Data range:

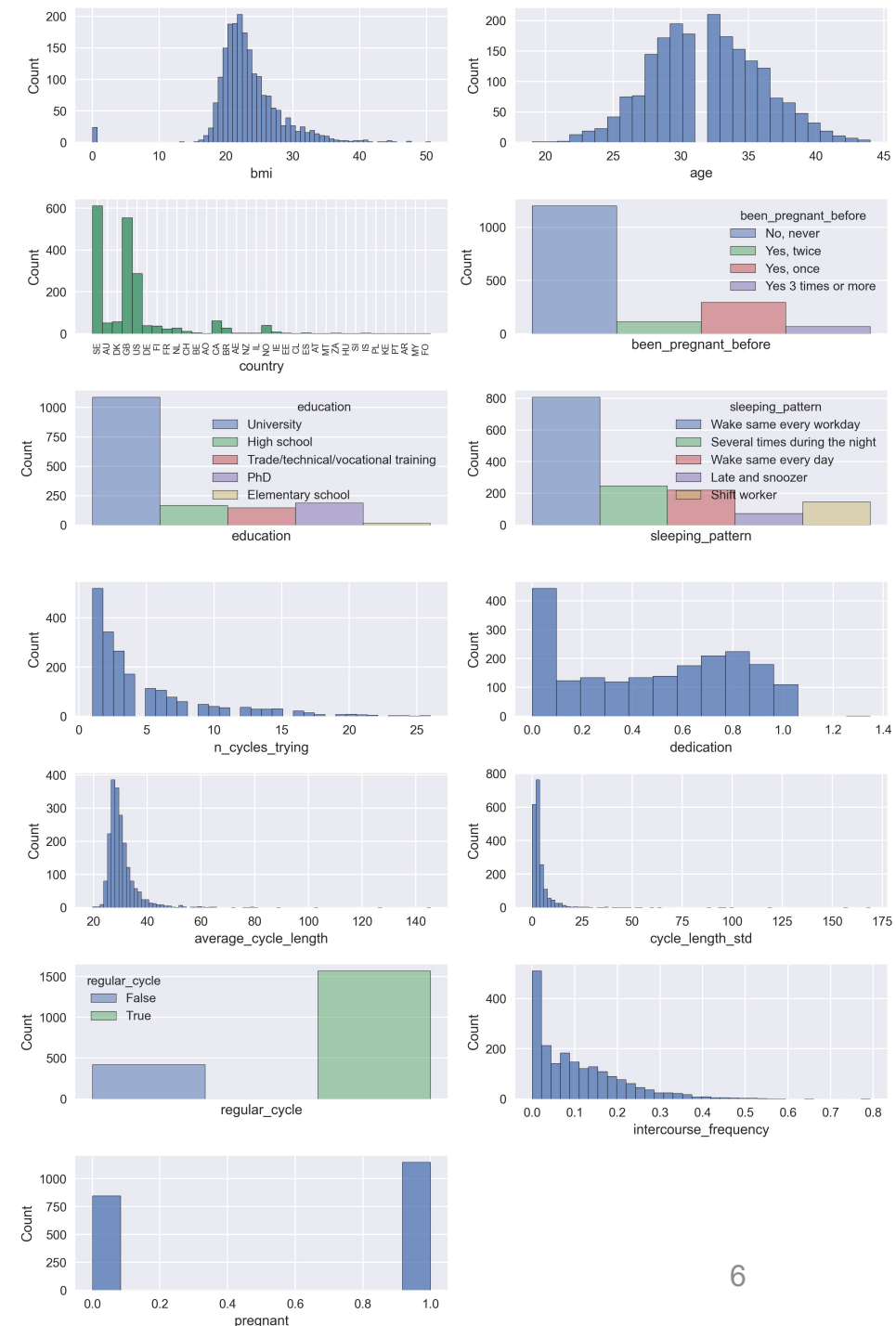
- BMI of 0 => not possible
- Big spread in (average) cycle length
- Dedication +100% => not possible

```
min      max
bmi      0.0  50.611299
age      19   44
country  None  None
been_pregnant_before  None  None
education  None  None
sleeping_pattern  None  None
n_cycles_trying  1   26
outcome    not_pregnant  pregnant
dedication  0.0  1.347826
average_cycle_length  19.5  145.5
cycle_length_std  0.0  168.998521
regular_cycle  False  True
intercourse_frequency  0.0  0.793103
```

EDA: Histograms

- BMI:
 - Contains missing data (BMI of 0) => **remove!**
 - Underweight (BMI <16) + Morbidly Obese (BMI > 40) present => **Keep or consider outliers?**
- Cycle length:
 - (Very) high cycle lengths (>35 - 145 days) => users with PCOS*? => **Keep or consider outliers?**
- Unbalanced variables:
 - Country
 - Been pregnant before
 - Education
 - Sleeping Pattern
 - Cycle regularity
 - Regular_cycle

=> **Makes them harder to use as predictors!**



*Polycystic ovary syndrome (PCOS) is characterised by irregular menstrual cycles, higher chance of diabetes type II and difficulty getting pregnant - https://en.wikipedia.org/wiki/Polycystic_ovary_syndrome

EDA: Data Inconsistencies?

- No intercourse, but still pregnant?
 - => remove samples

outcome <input type="checkbox"/>	medication <input type="checkbox"/>	average_cycle_length <input type="checkbox"/>	cycle_length_std <input type="checkbox"/>	regular_cycle <input type="checkbox"/>	intercourse_frequency <input type="checkbox"/>
pregnant	0.4166670000000000	27.266667000000000	2.3744670000000000	True	0.0
pregnant	0.8271600000000000	27.4	1.9198210000000000	True	0.0
pregnant	0.7972970000000000	24.666667000000000	1.2394480000000000	True	0.0
pregnant	0.6923080000000000	26.666667000000000	0.5773500000000000	True	0.0
pregnant	0.0967740000000000	29.75	3.0956960000000000	True	0.0
pregnant	0.5294120000000000	27.272727000000000	2.9695420000000000	True	0.0

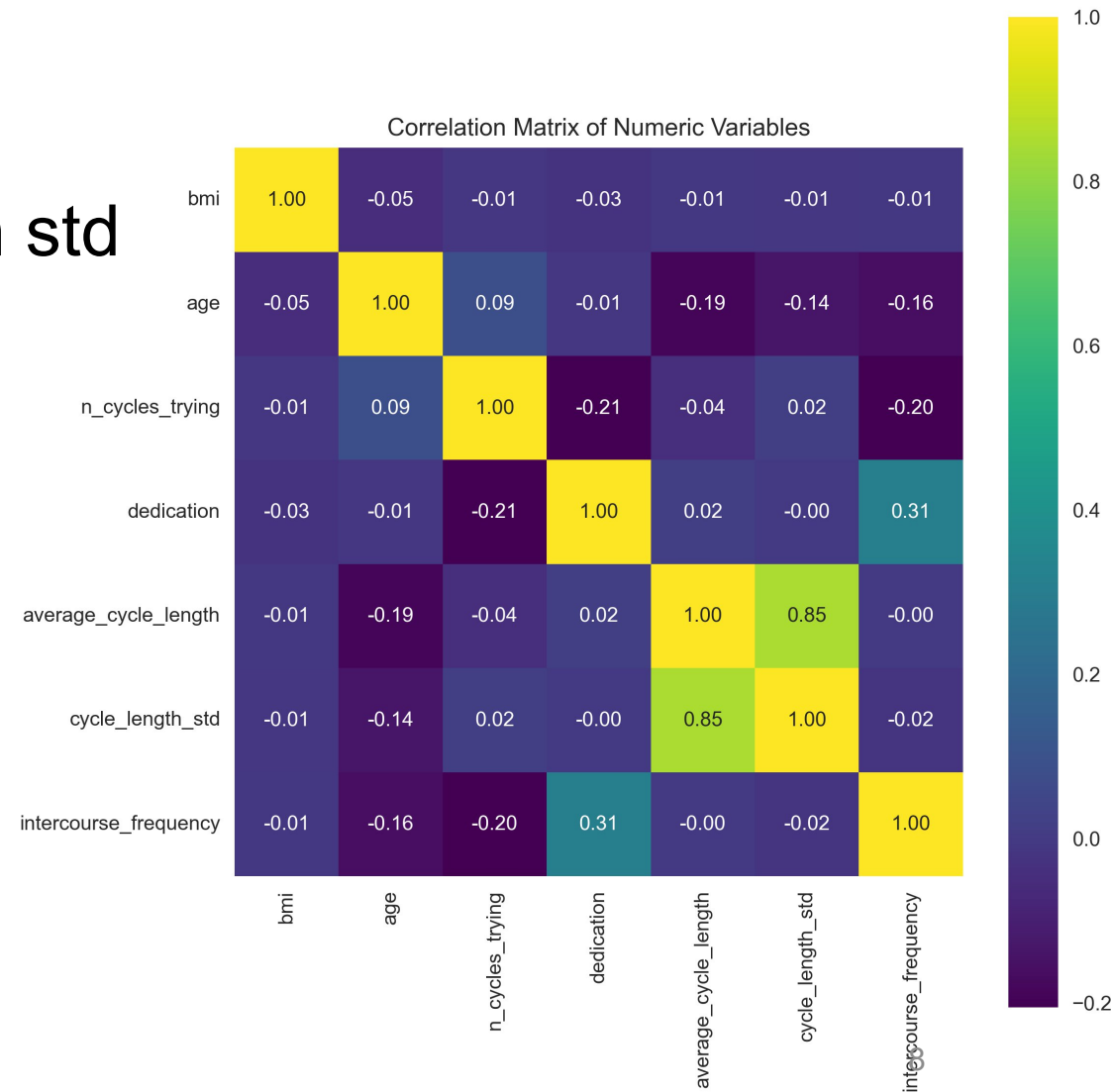
- cycles_length > 50 days are NOT regular cycles!
 - => Regularity of cycle is determined by cycle_length_std (< 5 days)

K	L	M
average_cycle_length <input type="checkbox"/>	cycle_length_std <input type="checkbox"/>	regular_cycle <input type="checkbox"/>
64.0	4.2426410000000000	True
52.0	2.8284270000000000	True
42.0	3.7416570000000000	True

EDA: Data correlation

High correlation:

- Average cycle length \Leftrightarrow Cycle length std
 - This is to be expected
 - \Rightarrow consider keeping only one of the two

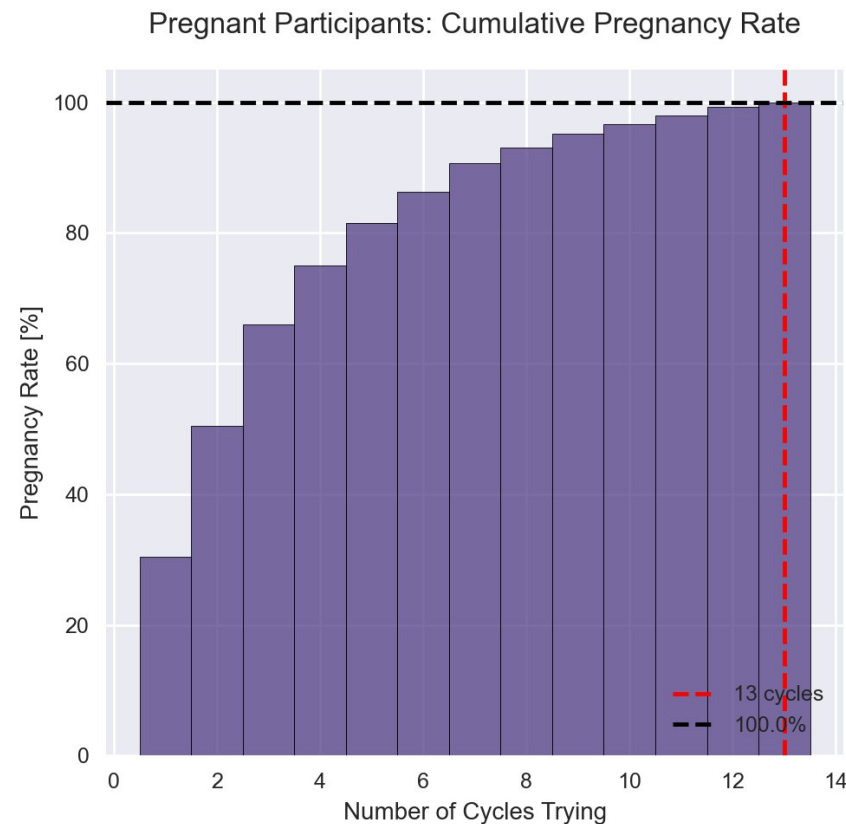
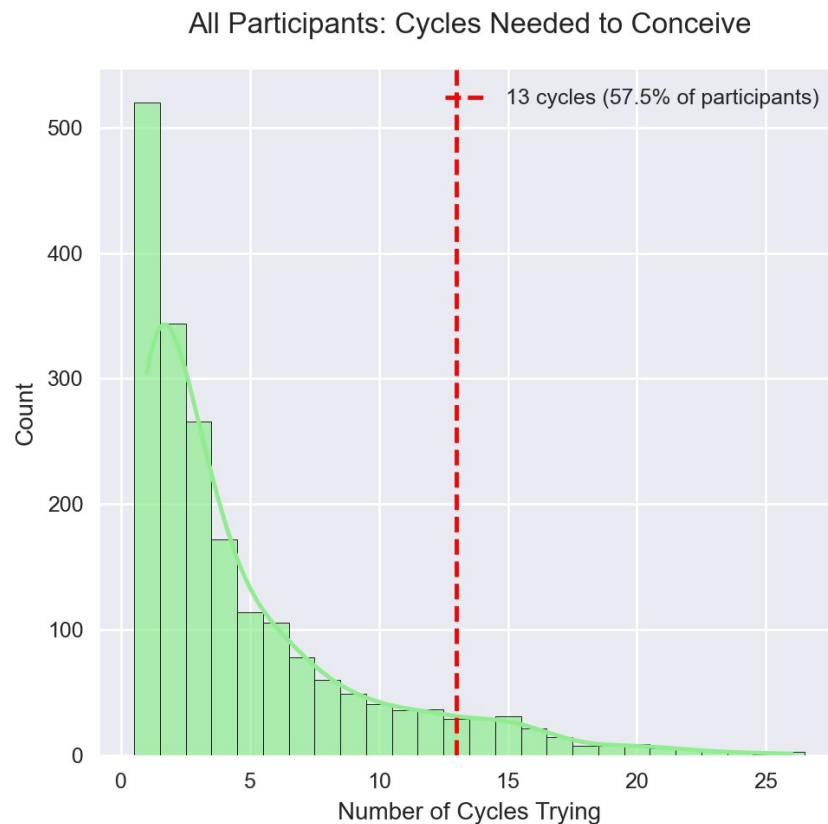


EDA: Conclusion

- Data seems usable!
- Data requires some clean-up
 - Not all samples + variable are useable for modelling

Q1: What is the chance of getting pregnant within 13 cycles?

Group:	Chance	Participants
Got pregnant within the study	57.5%	1148 / 1995 participants
Got pregnant within the study within 13 cycles	57.5%	1148 / 1995 participants
Got pregnant within 13 cycles out of all <u>pregnant</u> participants	100%	1148 / 1148 <u>pregnant</u> participants

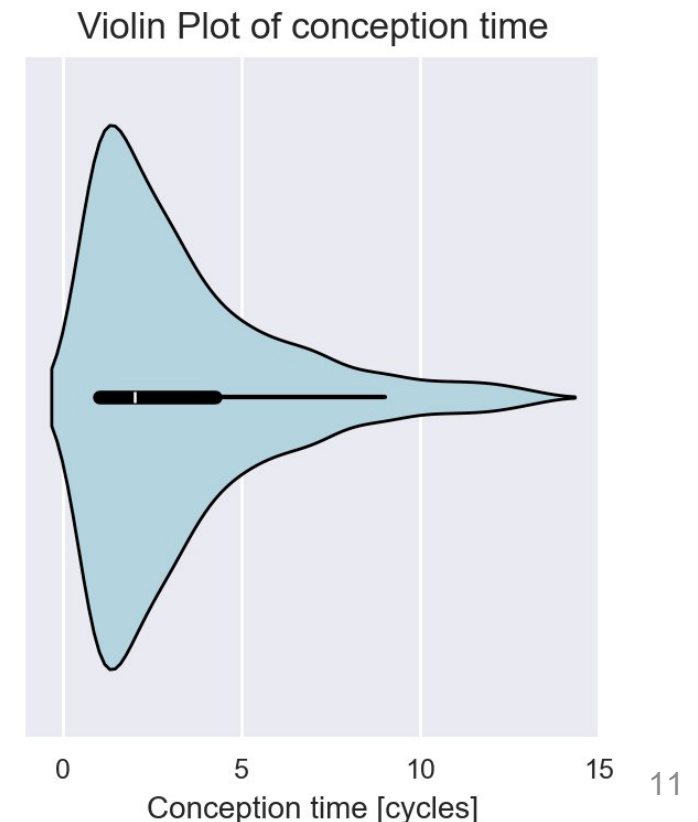
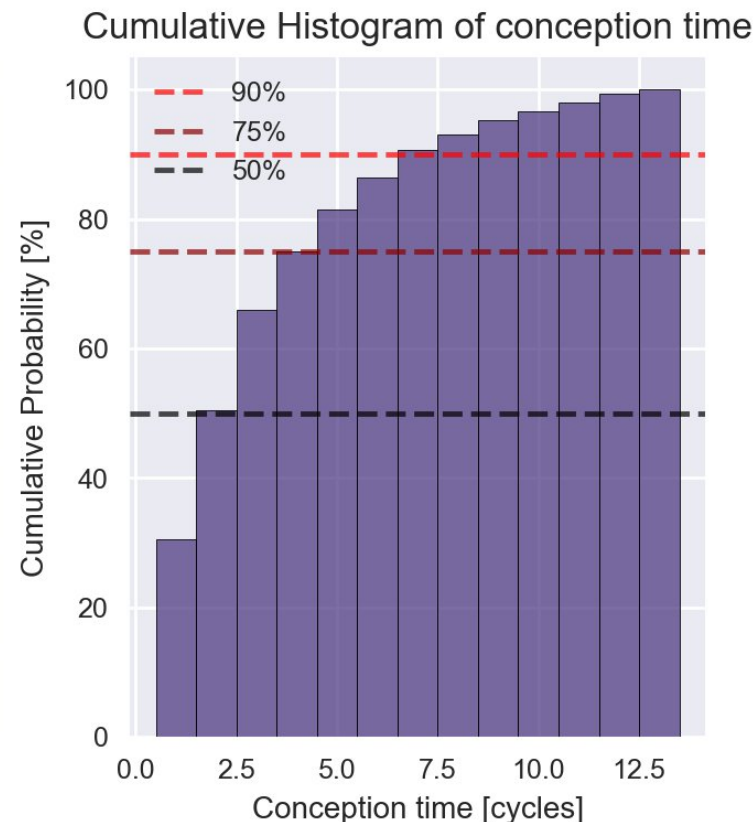
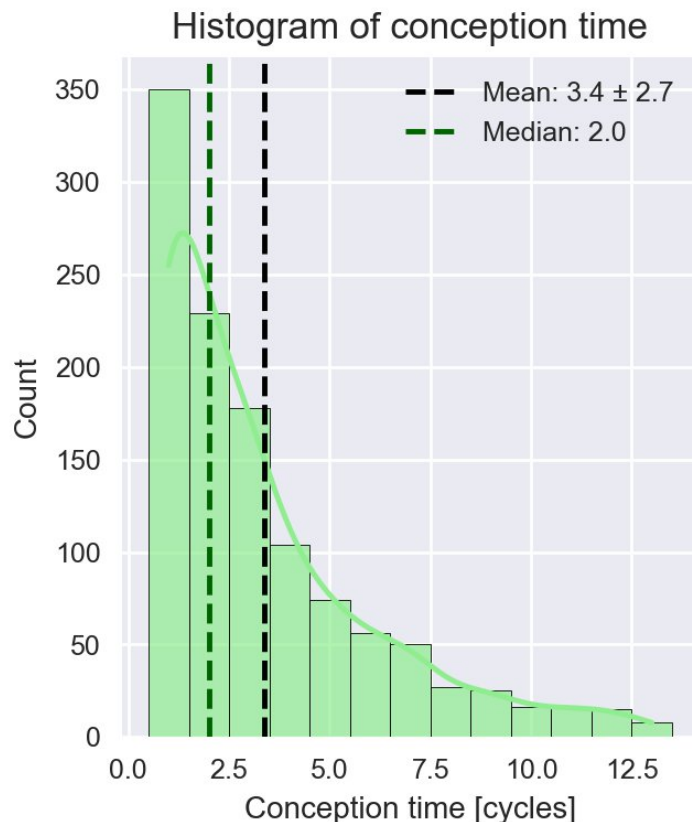


Q2: How long does it usually take to get pregnant?

- What is “usually”? => 50%? 90%?
- Answer expressed in Days/Cycles?
- Note: only pregnant participants are included

Answer:

- Majority of participants (>50%) got pregnant ≤ 2 cycles
- 90% did so ≤ 7 cycles



Q2: How long does it usually take to get pregnant?

Additional remarks:

- Not all participants got pregnant during the study
 - Only 1139/1975 participants (57,8%)
- Longer study time might show longer 'average' conceptions times

Q3: What factors impact the time it takes to get pregnant?

- Two approaches:

- Non-ML



K.I.S.S. | *Keep
It
Simple
Stupid*

- ML:

- **Why?**

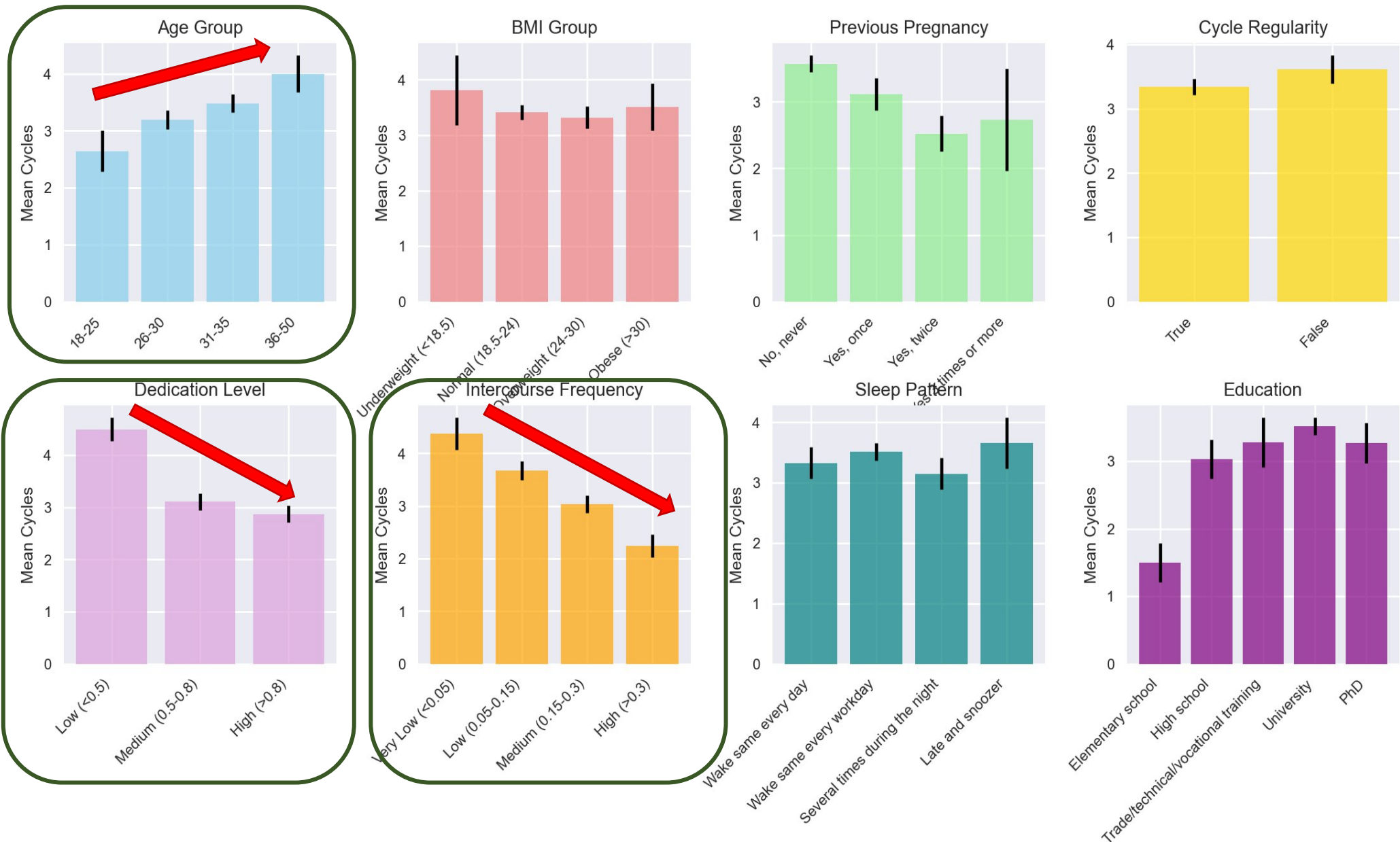
- (Relatively) Low amount of variables
 - Many categorical variables are unbalanced, can cause bias
 - Simpler method => easier interpretable results

First more data cleaning needed

Removed “outliers”	Why
Non-pregnant participants	We want to know how quickly, not IF they got pregnant.
Pregnant participants with intercourse frequency 0	Can’t get pregnant without sexual intercourse.
BMI < 12	Unrealistic value
Dedication > 1.00	You can’t log more than 100% of days
Remove samples with NaN’s / empty values	We want to investigate all parameters

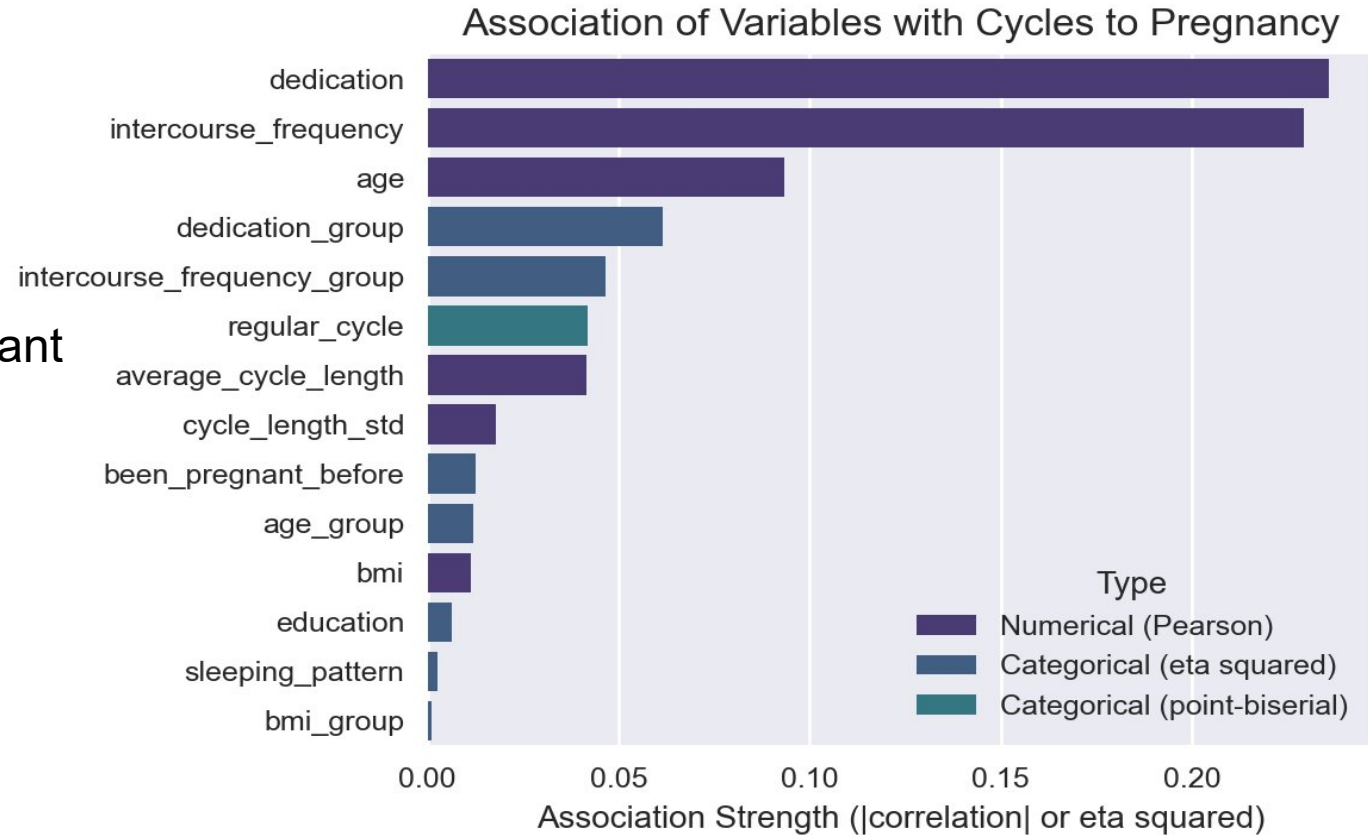
Clean samples: 1121 (out of 1995)

Simple binning in groups already shows obvious effects



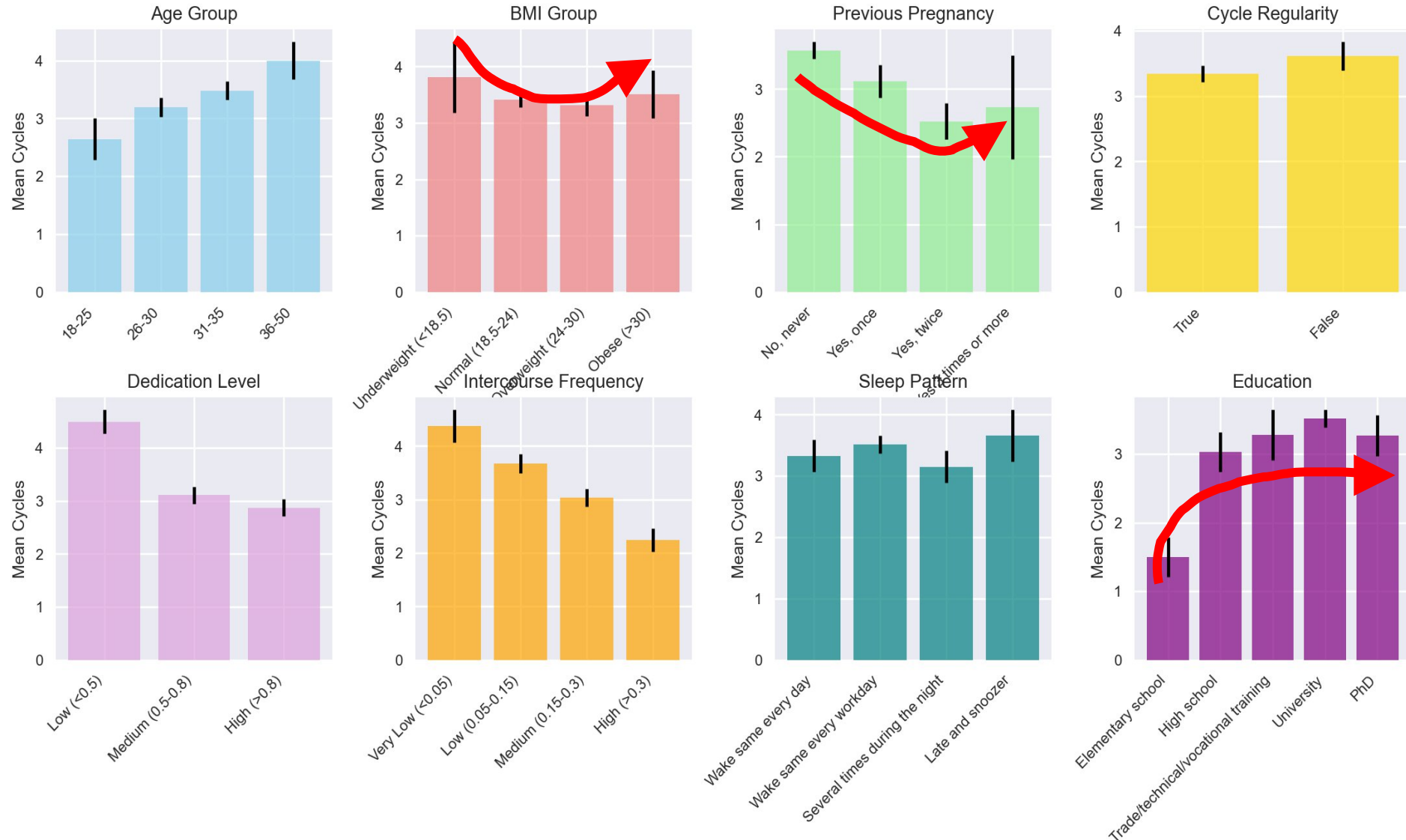
Statistical analysis confirms (linear) effect

- All 3 seem quite logical:
 - **Dedication**
 - Better tracking => quicker pregnant
 - **Intercourse frequency**
 - Regular intercourse => quicker pregnant
 - **Age**
 - Younger => more fertile => quicker pregnant



Also non-linear effects visible

Limitation of linear analysis => More advanced modelling needed?



Q4: How would your approach change if you were to use different techniques (e.g., ML or non-ML methods)?

- Two approaches:

- Non-ML

- **ML:**



- **Why?**

- Easier to investigate complex (non-linear) patterns
 - Resulting model (potentially) useable for predicting conception time

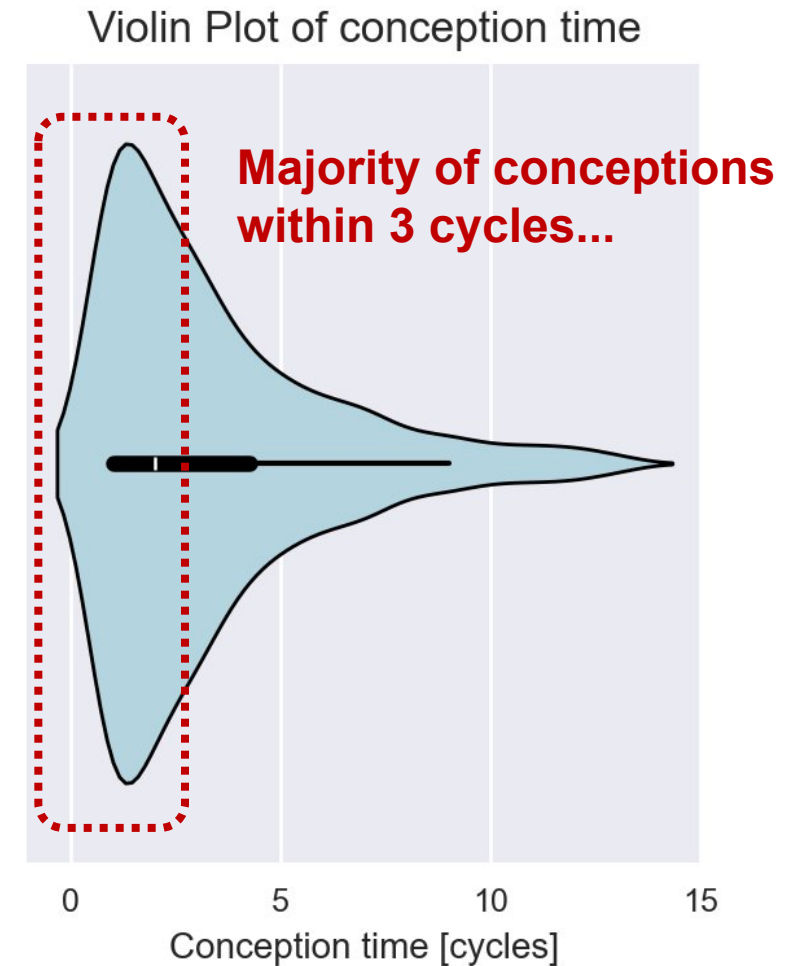
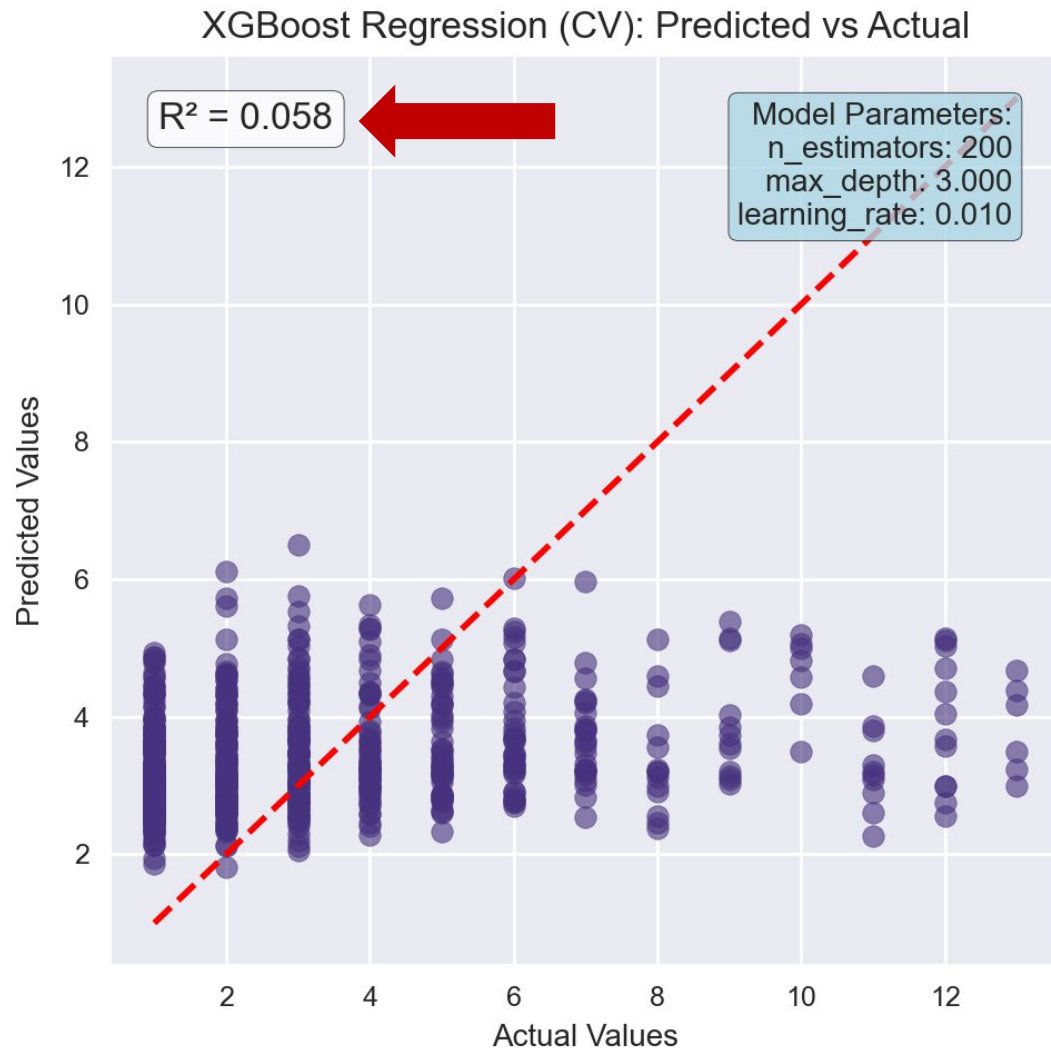
Model choice

The XGBoost logo is displayed in a bold, blue, italicized sans-serif font.

- When in doubt: **XGBoost!**
 - Easy implementation (especially compared to neural networks)
 - Accepts categorical + numerical data
 - Allows for classification + regression modeling
 - Build-in explainability tools (feature importance)
- Note: average performance over 5-fold cross validation is shown

Poor performance XGBoost regression model...

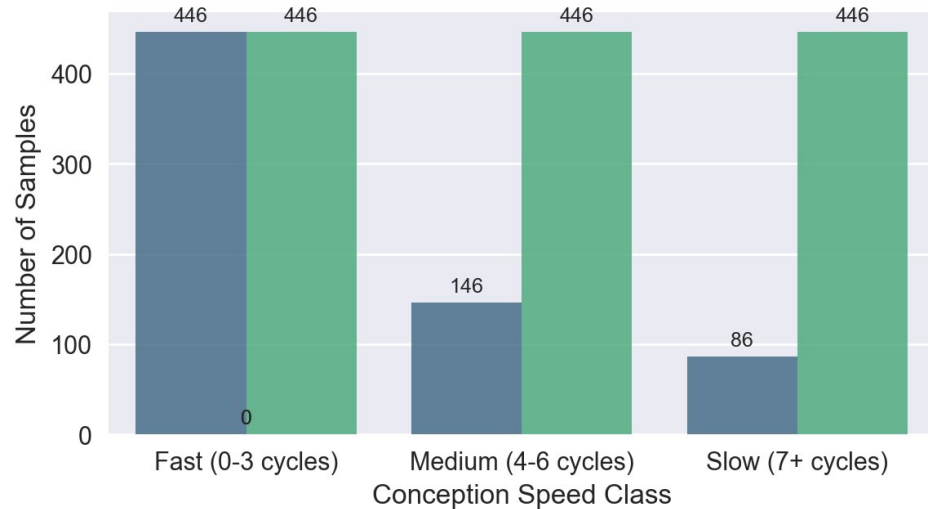
- Likely due to **data imbalance**



Solution: bin, balance and classify!

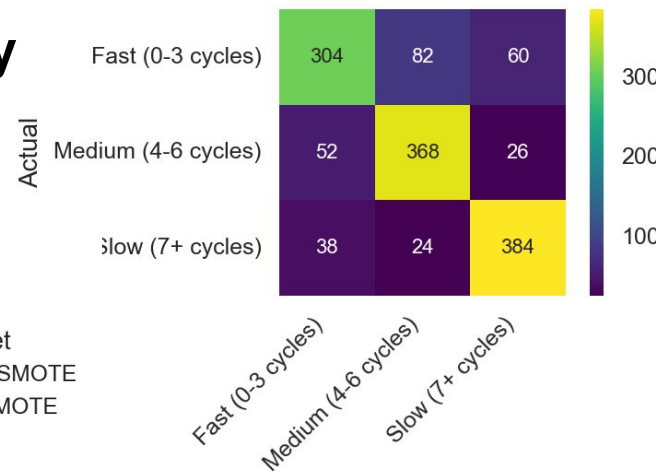
1. Group conception times into 3 bins
2. Rebalance classes
3. Classification now gives **~79% accuracy**

Class Distribution Before and After SMOTE

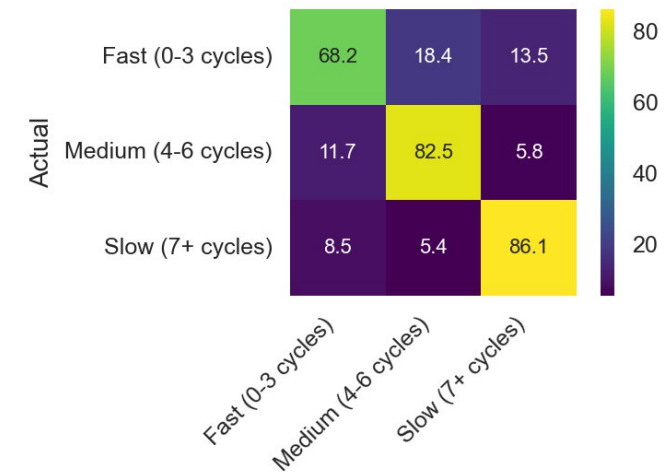


XGBoost Classification (CV): Classification Results

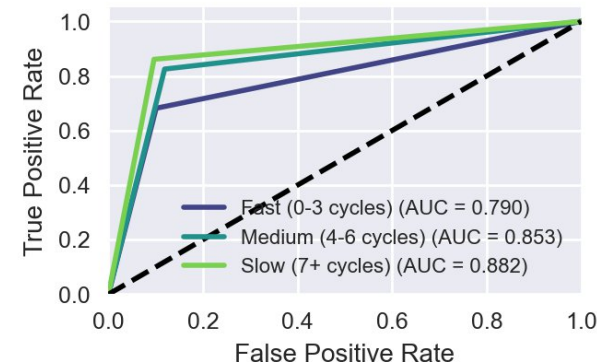
Confusion Matrix (Counts)



Confusion Matrix (%)



ROC Curves (AUC)

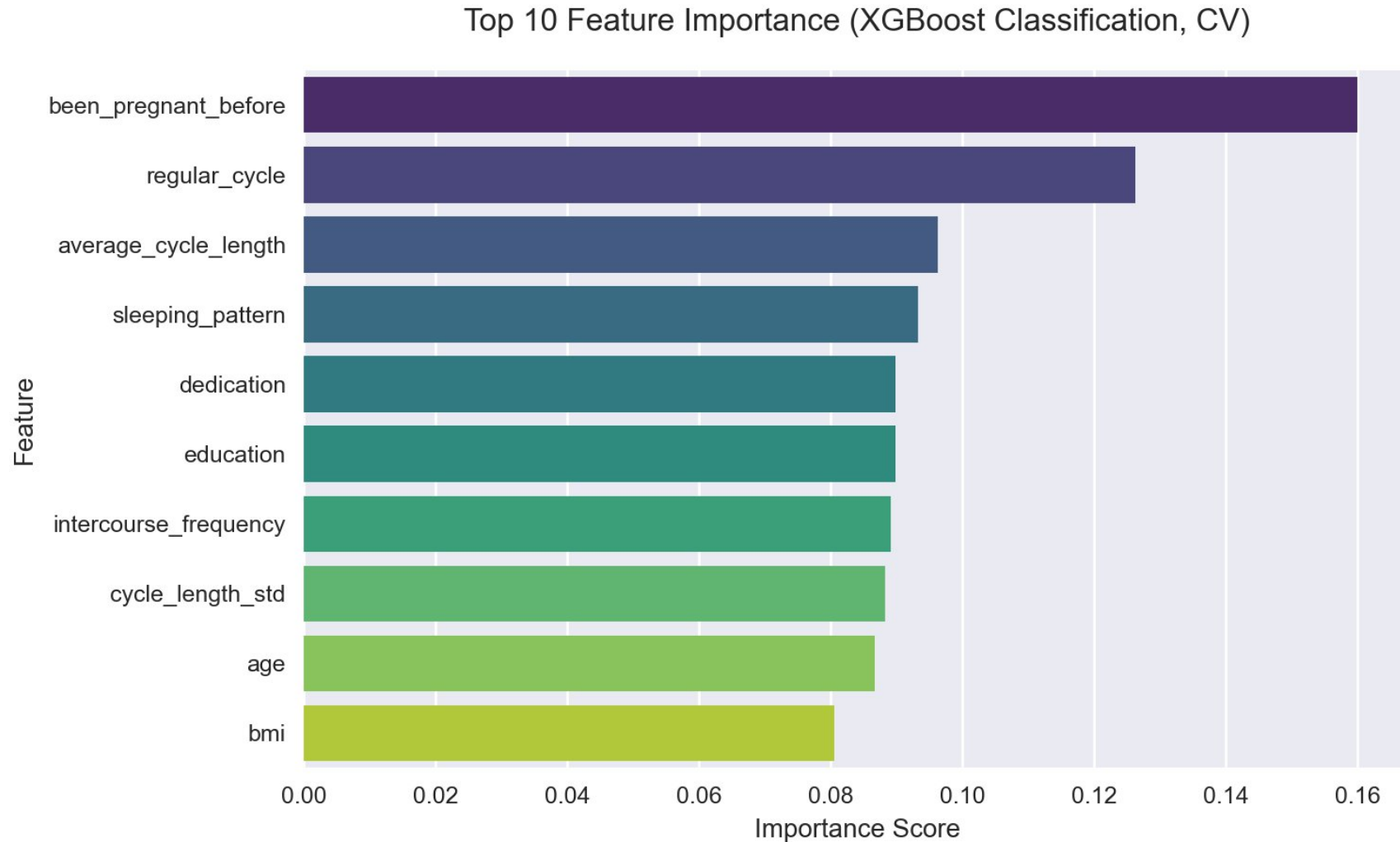


Model Configuration & Performance

Model Parameters:
 n_estimators: 180
 max_depth: 7.000
 learning_rate: 0.180

 Accuracy: 0.789
 F1 Score: 0.788

Feature importance shows new effects affecting conception time



Conclusion

- ML techniques require (more) tweaking to get working
 - But can reveal more influencing factors
- Factors affecting conception time (in order of importance):
 1. Nr of previous pregnancies
 2. Cycle regularity and average length
 3. Sleeping patterns
 4. Dedication to logging data on Natural Cycles
 5. Intercourse frequency
 6. Physical condition: Age + BMI
- (Further investigation needed **HOW** these affect conception time)