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Enhancing the detection of postpartum depression from electronic health records using machine learning algorithms

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Background and goals

Motivation

- Postpartum depression is one of the most common complications of pregnancy and childbirth, with estimated prevalence of 10-15%

- PPD risk is associated with biological, phycological and sociodemographic factors

- There are no quantitative tools for risk estimation, and screening is typically based on symptom questionnaires (such as the Edinburgh postnatal depression scale)

- Early identification of PPD risk during or before pregnancy may enable effective early intervention

Suggested solution

- A predictive model that uses electronic health records (EHR) for learning to identify patients at risk

- May enable early identification of patients at risk

- May be used to augment current screening tools
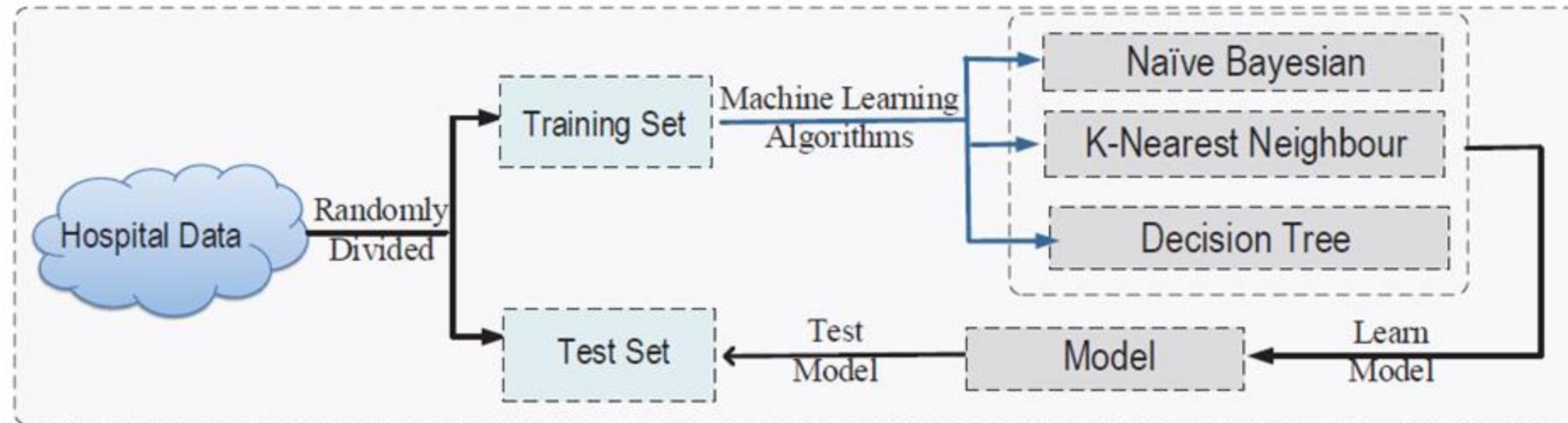
Expected impact

- Improved outcome for mother and child by early intervention
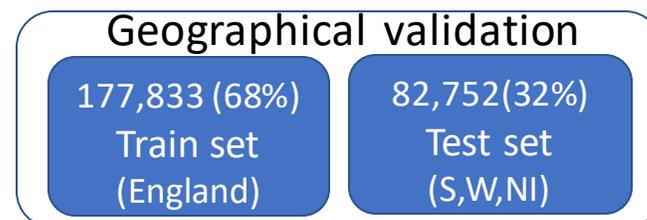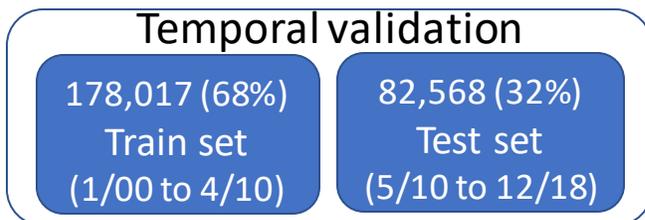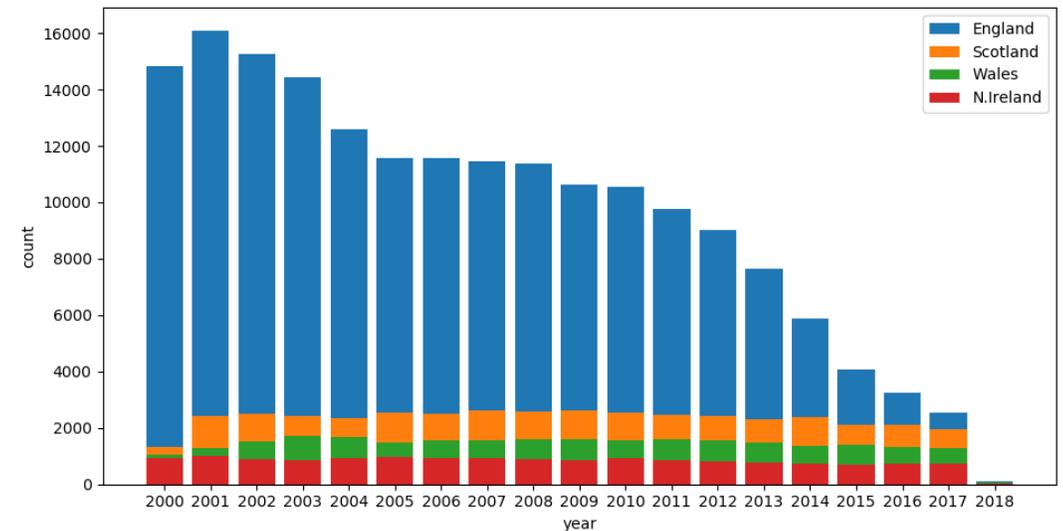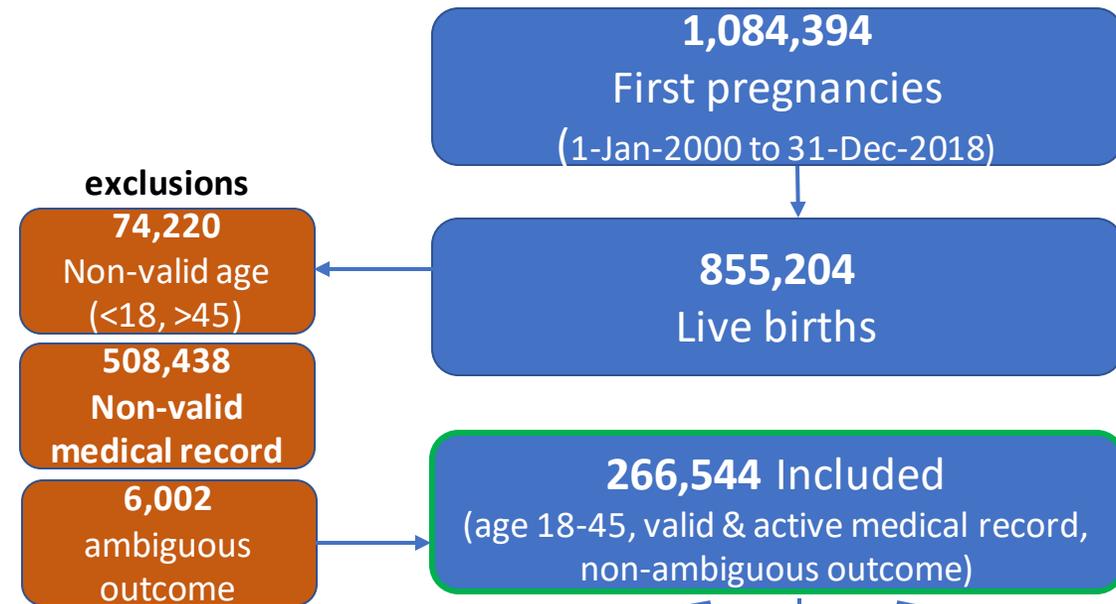
Machine learning in healthcare

Research approach: Use Machine Learning to identify signals within readily available EHR data to indicate individuals and sub-populations with higher risk for health-related outcomes

Data and patient cohort

- Dataset: UK primary care electronic health records (IQVIA-IMRD, ~18M patients)

1,084,394
First pregnancies
(1-Jan-2000 to 31-Dec-2018)

855,204
Live births

266,544 Included
(age 18-45, valid & active medical record, non-ambiguous outcome)

exclusions

74,220
Non-valid age
(<18, >45)

508,438
Non-valid
medical record

6,002
ambiguous
outcome

Temporal validation
178,017 (68%)
Train set
(1/00 to 4/10)

82,568 (32%)
Test set
(5/10 to 12/18)

Geographical validation
177,833 (68%)
Train set
(England)

82,752(32%)
Test set
(S,W,NI)

Holdout set
5959
EPDS

PPD Outcome definition

At least one of the following indications <u>during the first year after birth</u>:

1. Women with depression related diagnoses
   (excluding depression symptoms and anxiety)

2. Women with new antidepressant prescriptions
   (excluding drugs with dual indications)

3. Women with depression related non-pharmacological treatment (NPT)

	Recorded Depression Tx	No recorded Depression Tx	Total (% pts)
Recorded Depression Dx	**16,284**	**5,869**	22,153 (8.3%)
No recorded Depression Dx	**13,555**	**230,836**	
Total (% pts)	29,839 (11.2%)		35,708 (13.4%)

PPD prevalence	Train set	Test set	Holdout set
Geographical validation	12% (E)	15% (S,W,NI)	20% (18%,26%) E S,W,NI
Temporal validation	14% (00-10)	12% (10-18)	20% (18%,29%) 00-10 10-18

Tx: treatment (antidepressants or NPT); Dx: diagnosis code
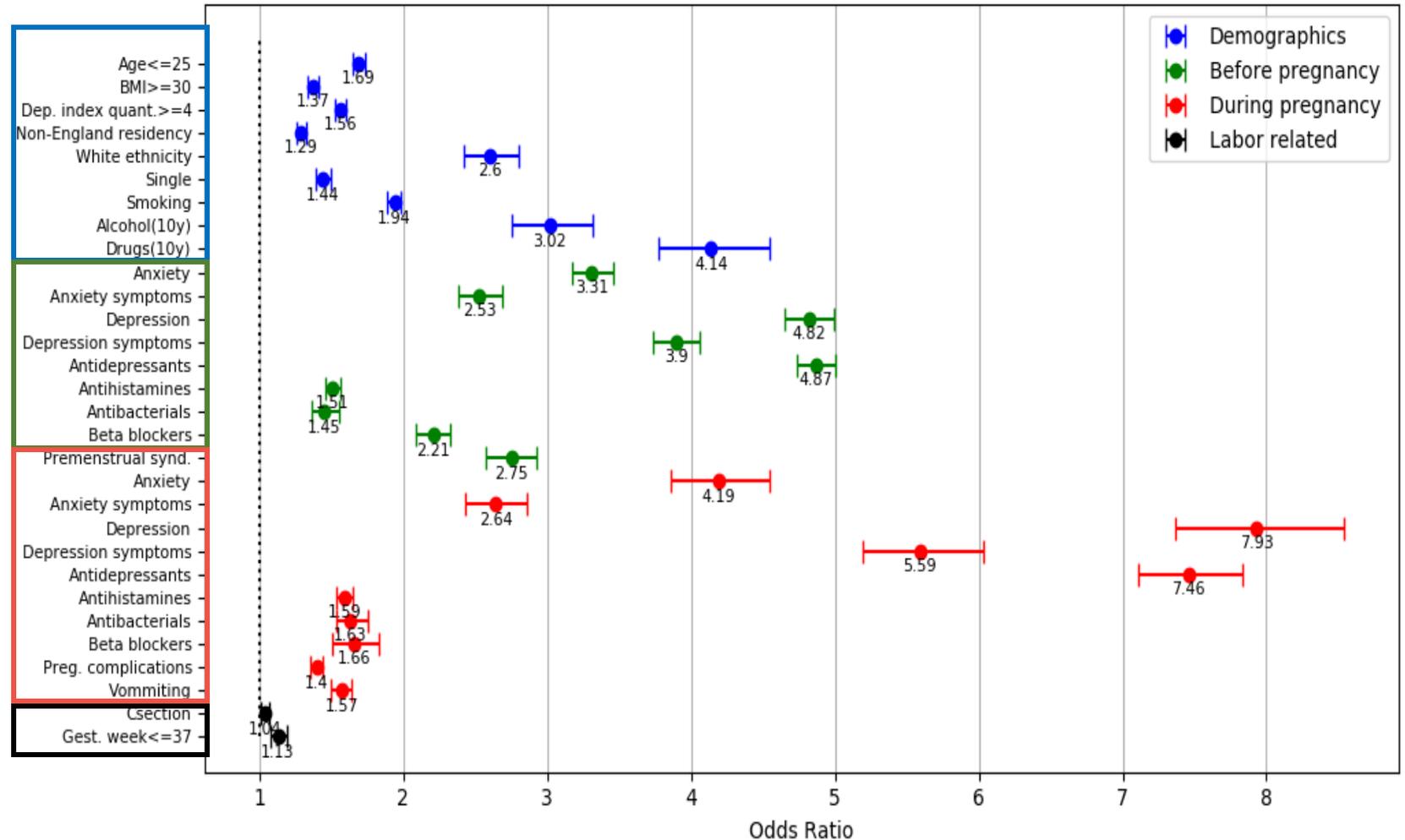
Predictor variables

- Demographic and socioeconomic
  - Age, BMI, marital status, ethnicity
  - Deprivation index, smoking, alcohol-use, drug-use

- Medical diagnoses during pregnancy
  - Mental disorders and symptoms: depression, anxiety, psychosi
  - Pregnancy complications: GDM, preeclampsia, vomiting
  - Miscellaneous health conditions: migraine, diarrhea

- Labor and infant-related
  - Labor complications: cesarean section, episiotomy
  - Infant-related: gestational week, birth weight, APGAR, feeding type

- Medical diagnoses before pregnancy (2y)
  - Mental disorders and symptoms, PMS
  - Visit count

- Drug prescriptions during / before pregnancy (2y)
  - Antidepressants, antibacterial, antihistamines, beta-blocking
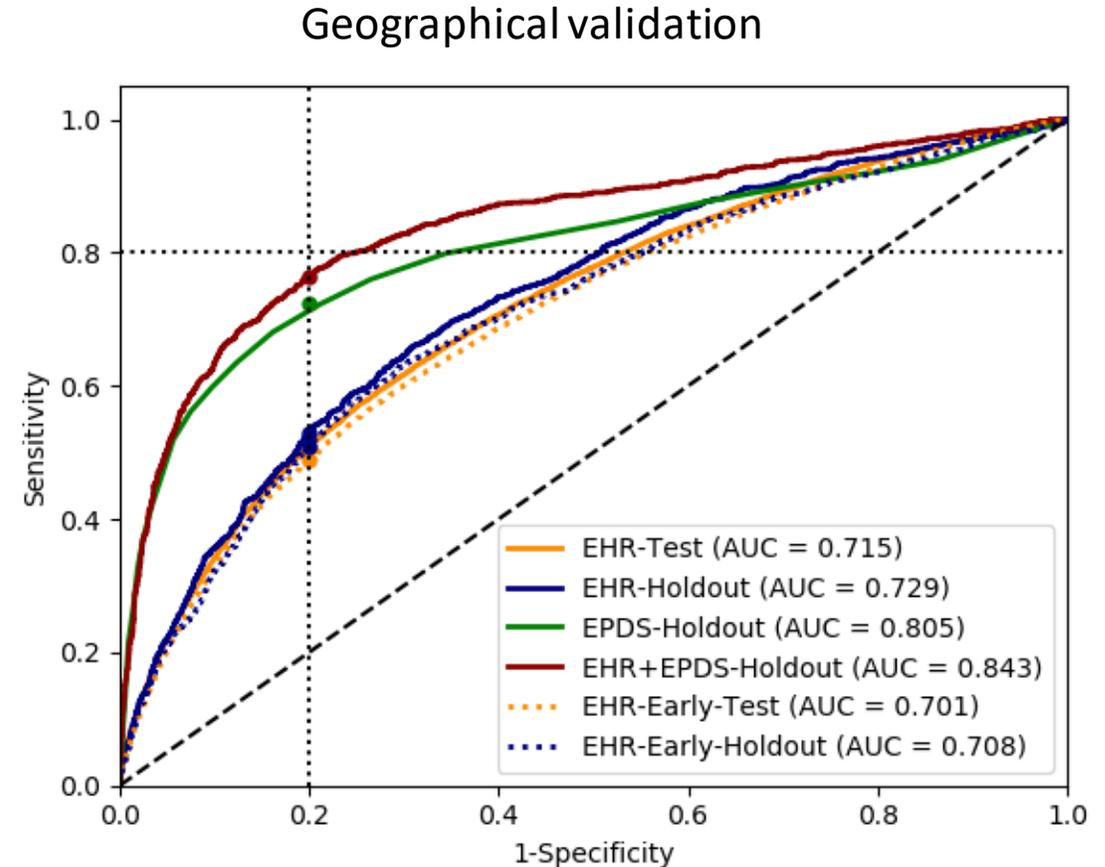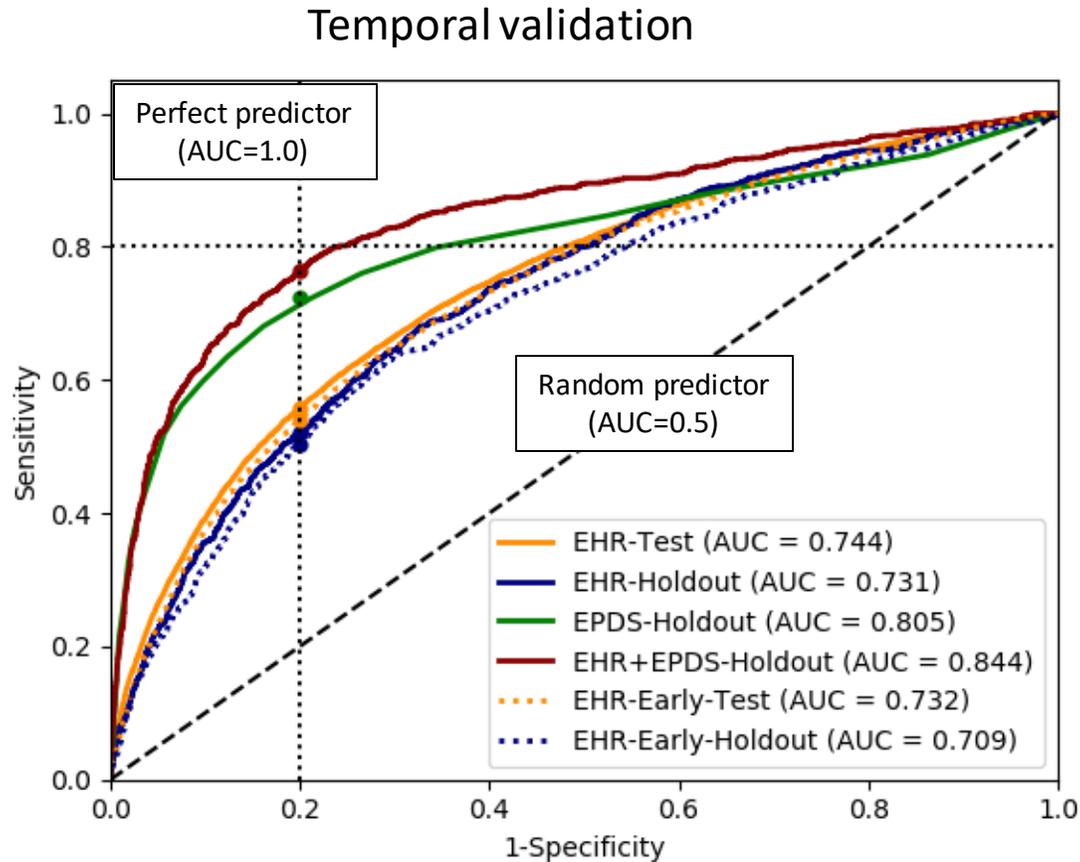 - Drug prescription count

Patient characteristics and univariate analysis

Characteristic	Value
N	266544
Age (yrs)	30.0±5.8
Ethnicity	
White	99971 (37.5%)
Asian	7367 (2.8%)
Black	3167 (1.2%)
Other	2412 (0.9%)
Unknown	152573 (57.2%)
Marital status	
Single	34145 (12.8%)
Married	62929 (23.6%)
Unknown	169470 (63.6%)
Country	
England	182506 (68.5%)
Scotland	42113 (15.8%)
Wales	26565 (10.0%)
N. Ireland	15360 (5.8%)
Deprivation index (quantile)	3.03±1.3
Pre-pregnancy BMI	25.4±5.0
Cesarean section	51151 (19.2%)
Smoking	64778 (24.3%)
History of depression	17384 (6.5%)

Prediction performance (gradient boosting trees models)

Temporal validation

Perfect predictor (AUC=1.0)

Random predictor (AUC=0.5)

EHR-Test (AUC = 0.744)
EHR-Holdout (AUC = 0.731)
EPDS-Holdout (AUC = 0.805)
EHR+EPDS-Holdout (AUC = 0.844)
EHR-Early-Test (AUC = 0.732)
EHR-Early-Holdout (AUC = 0.709)

Geographical validation

EHR-Test (AUC = 0.715)
EHR-Holdout (AUC = 0.729)
EPDS-Holdout (AUC = 0.805)
EHR+EPDS-Holdout (AUC = 0.843)
EHR-Early-Test (AUC = 0.701)
EHR-Early-Holdout (AUC = 0.708)

- Combining EHR-based prediction with EPDS score improved of EPDS-alone (sensitivity 0.76 vs. 0.72, specificity 0.8)
- Early EHR-based prediction (before pregnancy) is nearly as accurate as late prediction (after labor)

Variable importance (SHAP analysis[†])

Younger age (<30 yrs) has strong impact on the predictor
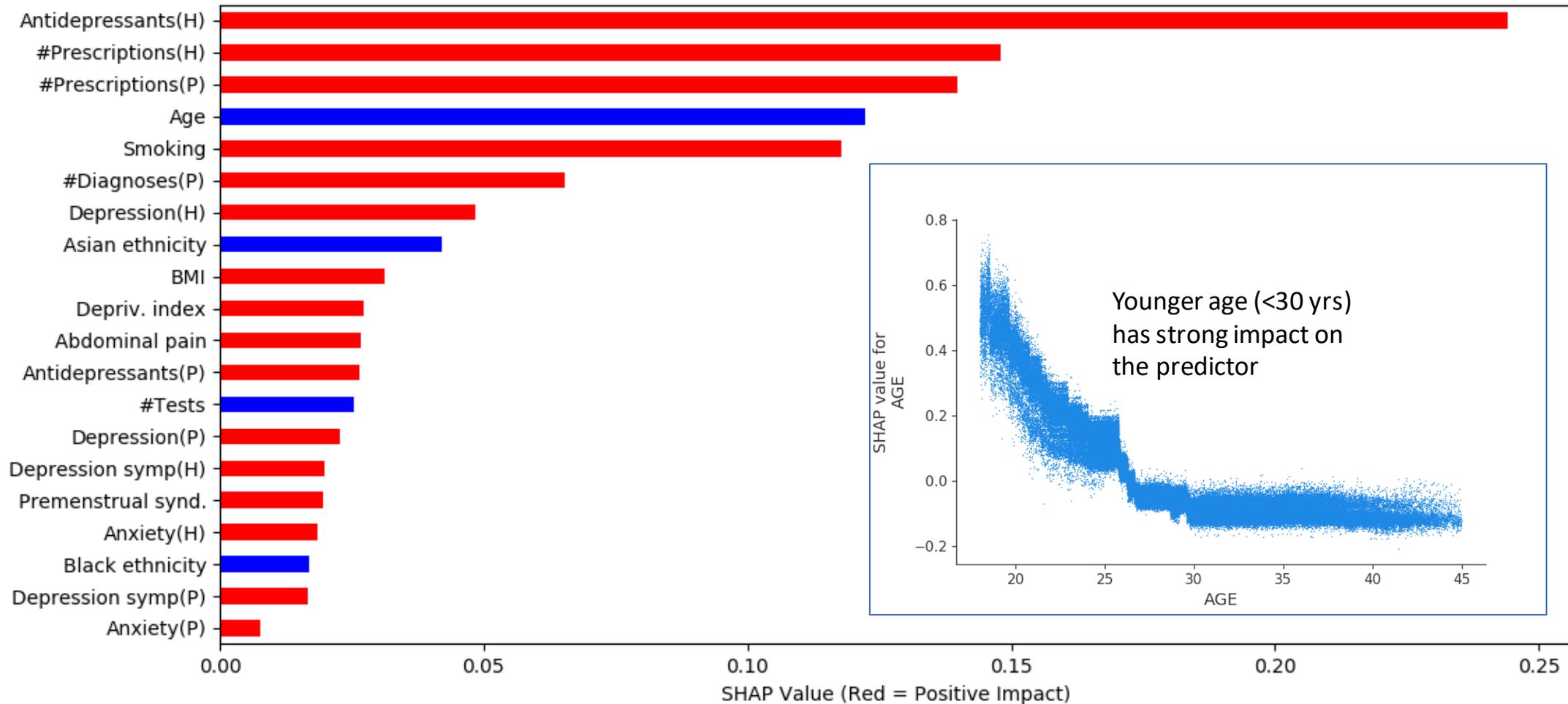
RED = increase risk ; BLUE = decreased risk
P=during pregnancy, H=History (2y before pregnancy)
Higher SHAP value = stronger variable contribution

† Lundberg SM, Lee SI, A Unified Approach to Interpreting Model Predictions, NeurIPS 2017

Summary and conclusions

- PPD can be predicted from EHR data with fair accuracy, even before pregnancy

- EHR-based prediction can improve the accuracy of EPDS for PPD screening

- Although some of the risk factors are well known, integrating and quantifying them into an accessible risk score may have clinical value

- Future work:
    - External validation
    - Deployment in a clinical environment
    - A prospective study

Collaborators

- **Weil Cornell Medicine**
  - Jyotishman Pathak
  - Yiye Zhang
  - Alison Herman
  - Rochelle Joly
  - Meghan R. Turchioe
- **Sheba Medical Center**
  - Vered Bar
- **Our team at KI**
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