

Research on the Prediction Model for Early Biliary Stasis after Sleeve Gastrectomy Based on Machine Learning Algorithms

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01 ▶ **Research Background**

02 ▶ **Research Objective**

03 ▶ **Research Methods**

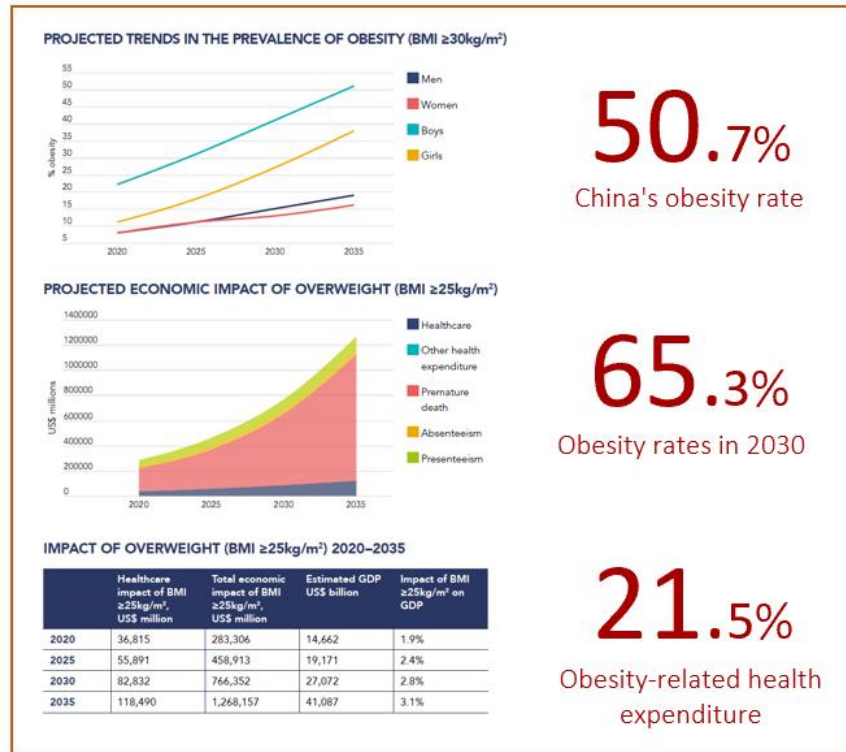
04 ▶ **Research Results**

05 ▶ **Conclusion**

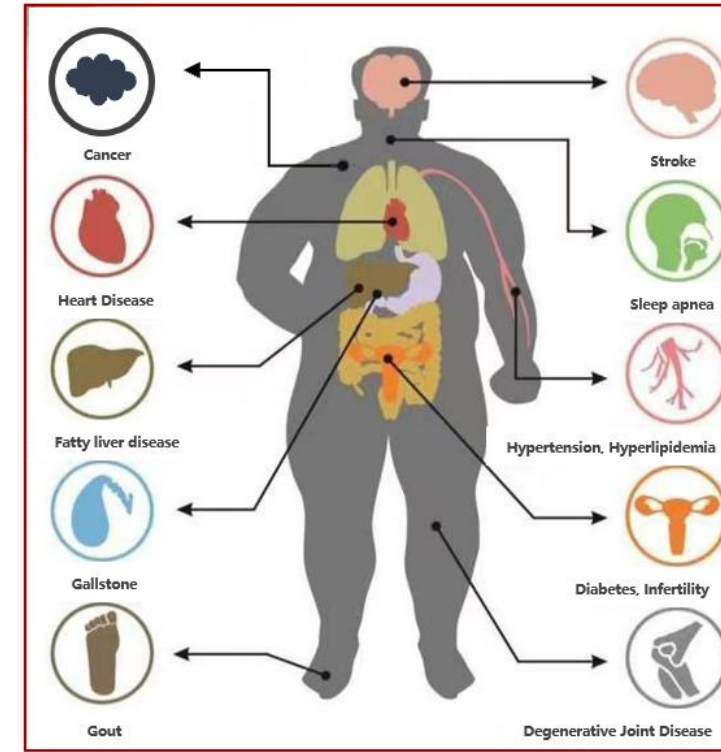
PART 01[▽]

Research Background

Obesity and related chronic diseases are a major public health problem that seriously endangers the health of the nation

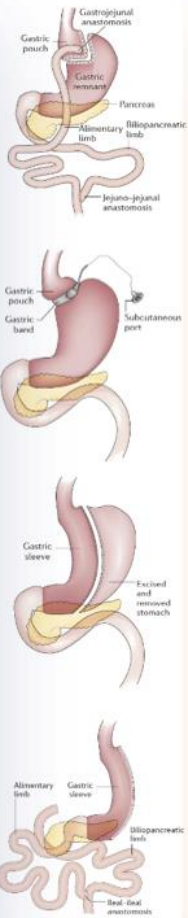


Serious epidemic



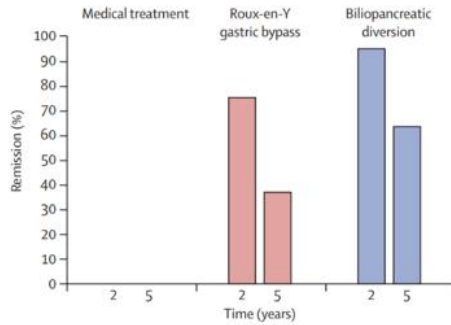
Independent risk factors for chronic diseases

Lancet Diabetes Endocrinol, 2021, 9(6): 373-392.
Lancet Diabetes Endocrinol, 2021, 9(7): 446-461.

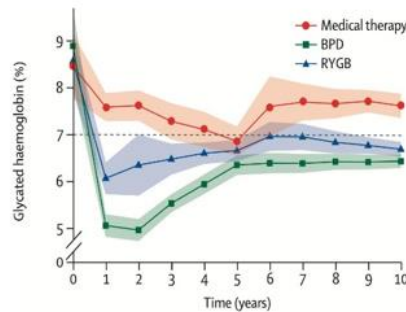


Diabetes remission

- Diabetes remission rate 2 years after surgery, **75%~95%**



- 10 years after surgery, **87.5%** of T2DM achieve HbA1c <7.0%

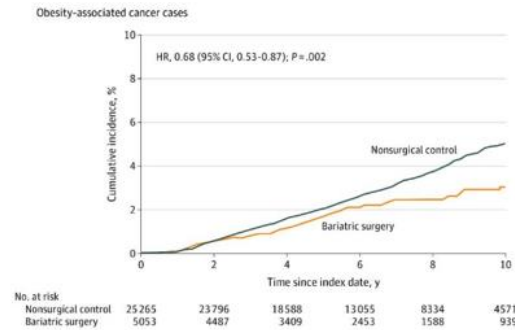


Lancet. 2015, 386(9997):964-973.

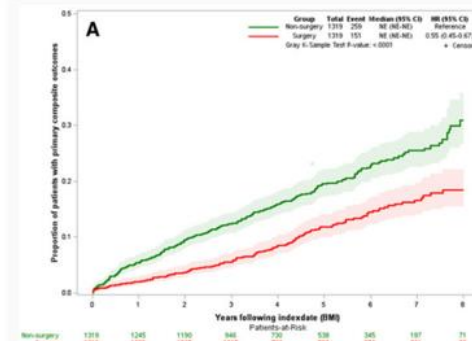
Lancet, 2021, 397(10271): 293-304.

Reducing chronic diseases

- 32%** reduction in cancer risk



- 42%** reduction in risk of serious adverse cardiovascular events

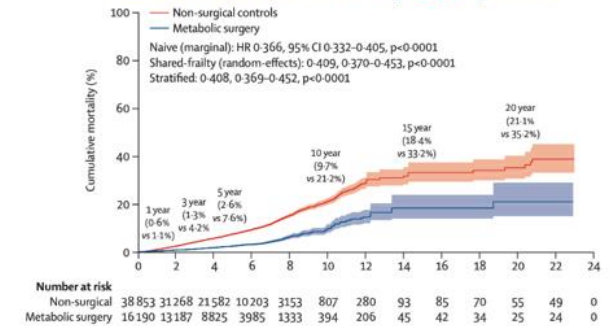


JAMA. 2022;327(24):2423-2433.

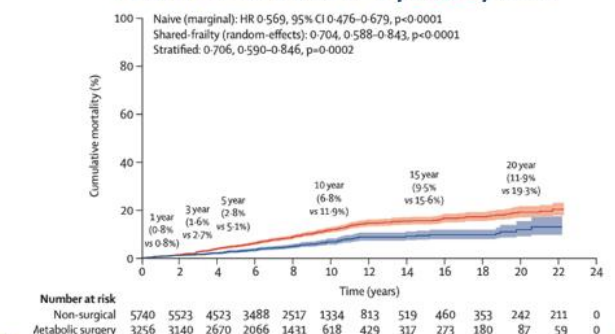
Circulation. 2021;143(15):1468-1480.

Extension of life expectancy

- Median life expectancy for T2DM increased by **9.3** years

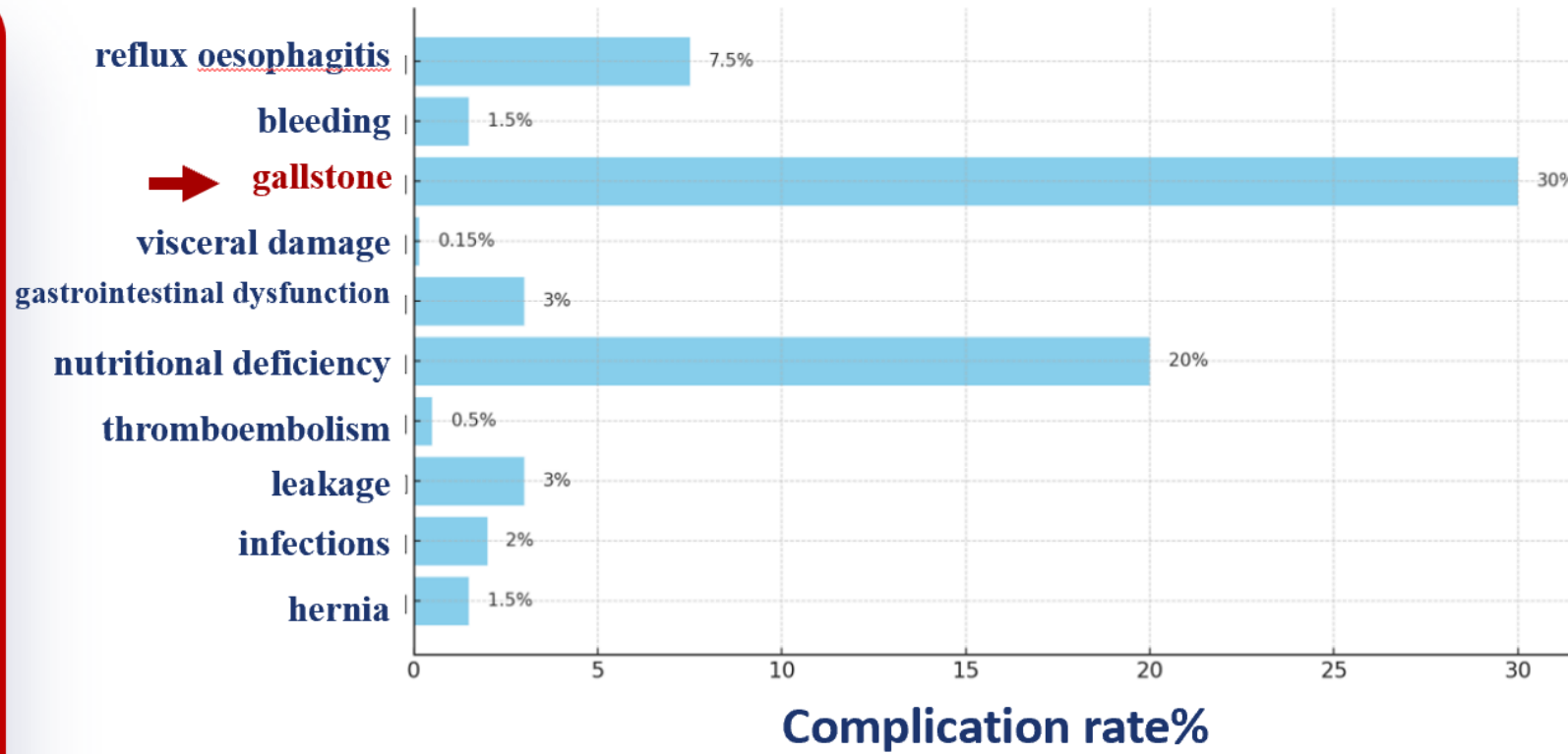


- Median life expectancy for non-T2DM increased by **5.1** years



Lancet, 2021, 397(10287): 1830-1841.

Postoperative-related complications



postoperative cholelithiasis

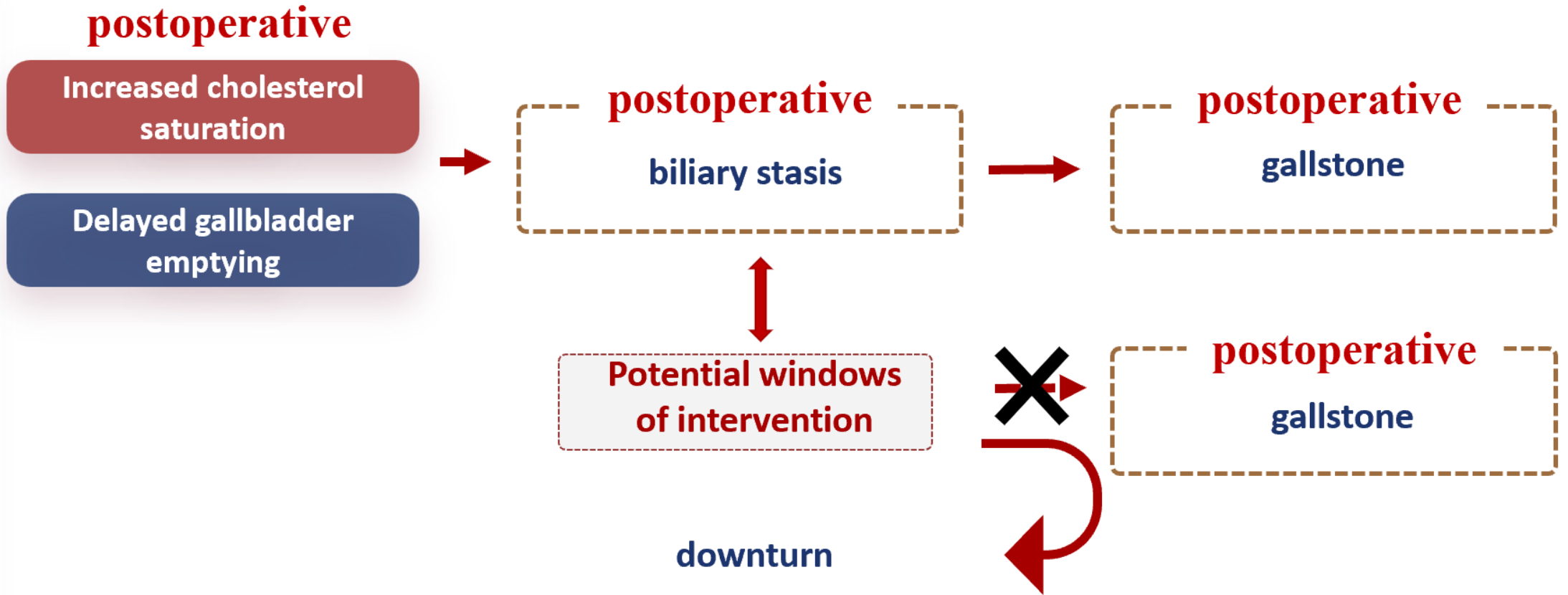
- Rapid weight loss
- Changes in bile composition
- Abnormal gallbladder function
- Hormonal changes in the gut
- Dietary and nutritional changes

5.5 times
symptomatic
gallbladder stones

5.4 times
Cholecystectomy

3.5 times
Complication of
cholecystectomy

Pathophysiological changes after bariatric surgery



Existing problems

Uncertainty about the onset of cholelithiasis after metabolic surgery

Unclear programme of postoperative pharmacological interventions

01



02

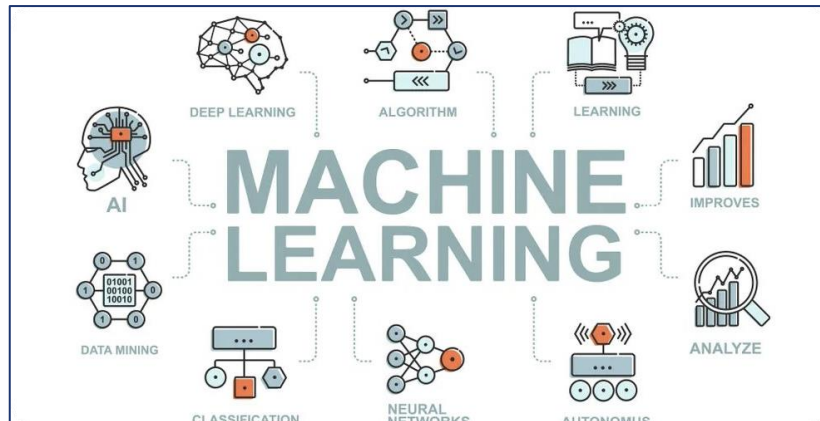
The risk of patient morbidity cannot be predicted **preoperatively**

04

Postoperative management cannot be **personalised**

PART 02[▽]

Research Objective



- ▶ machine learning-based model to predict the risk of early biliary stasis

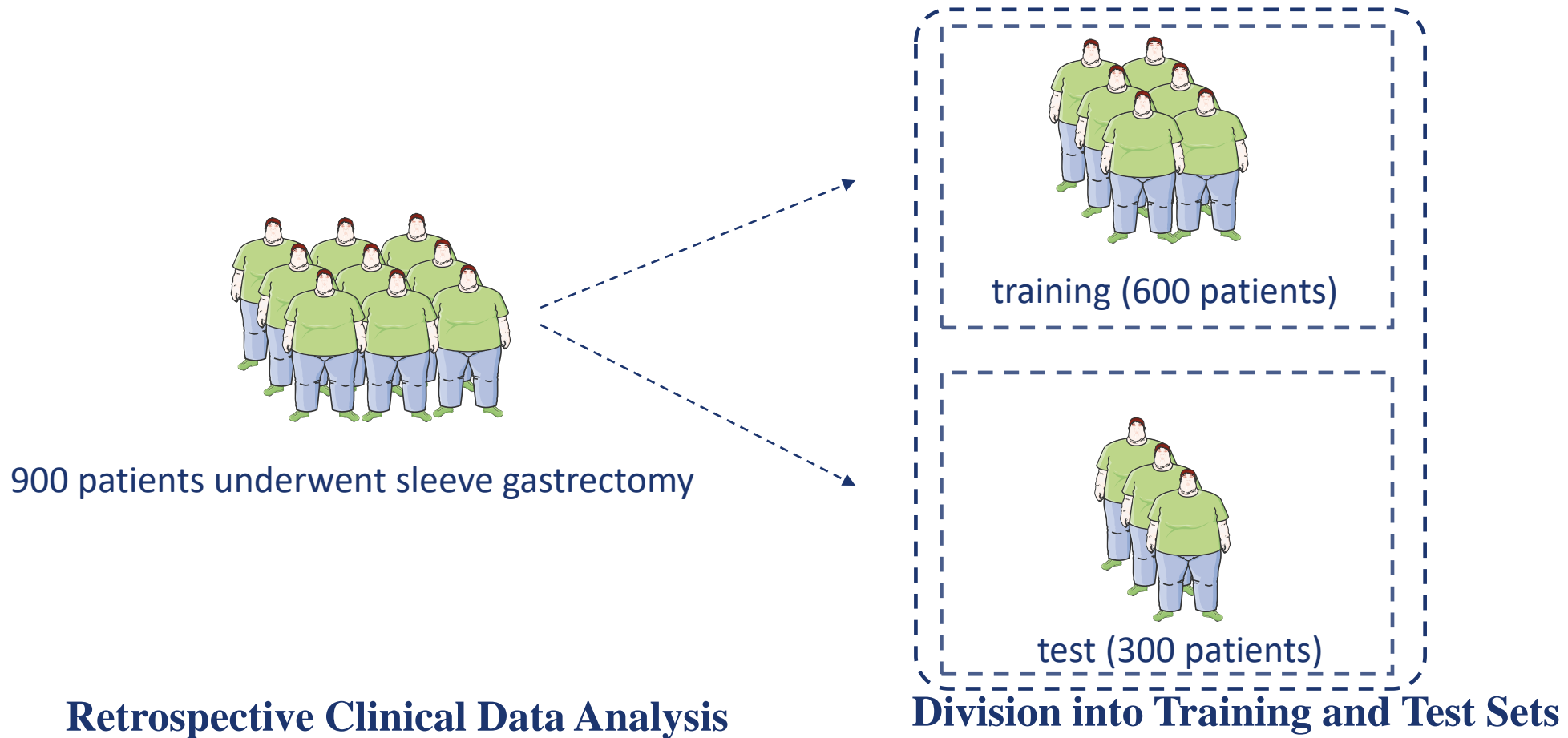


- ▶ guidance on clinical treatment

PART 03[▽]

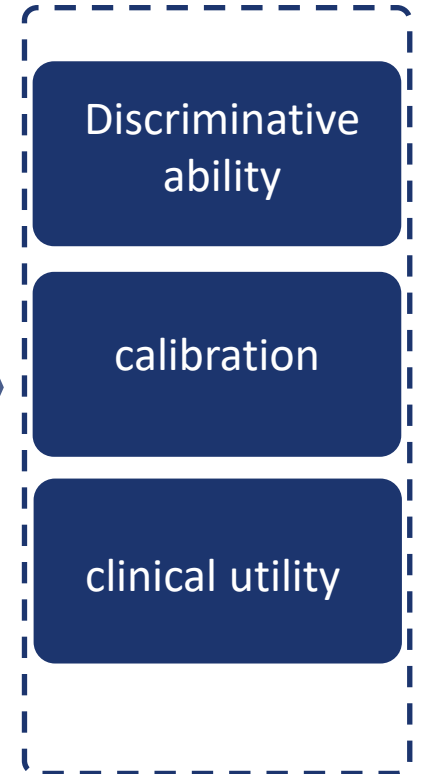
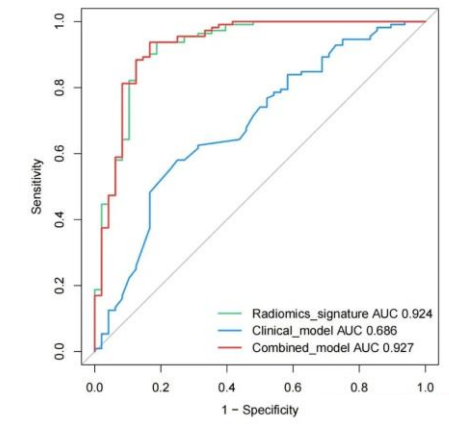
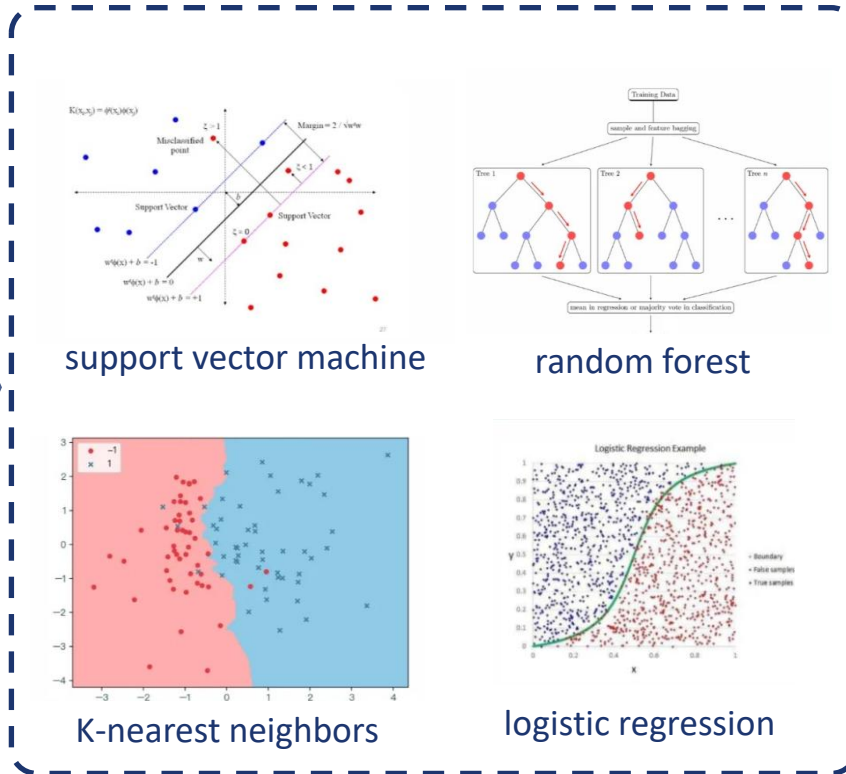
Research Methods

Data Collection and Patient Selection



Feature Selection and Model Construction

Univariate analysis and identifying variables most predictive of early biliary stasis



Univariate Analysis for Feature Selection

Application of Machine Learning Algorithms

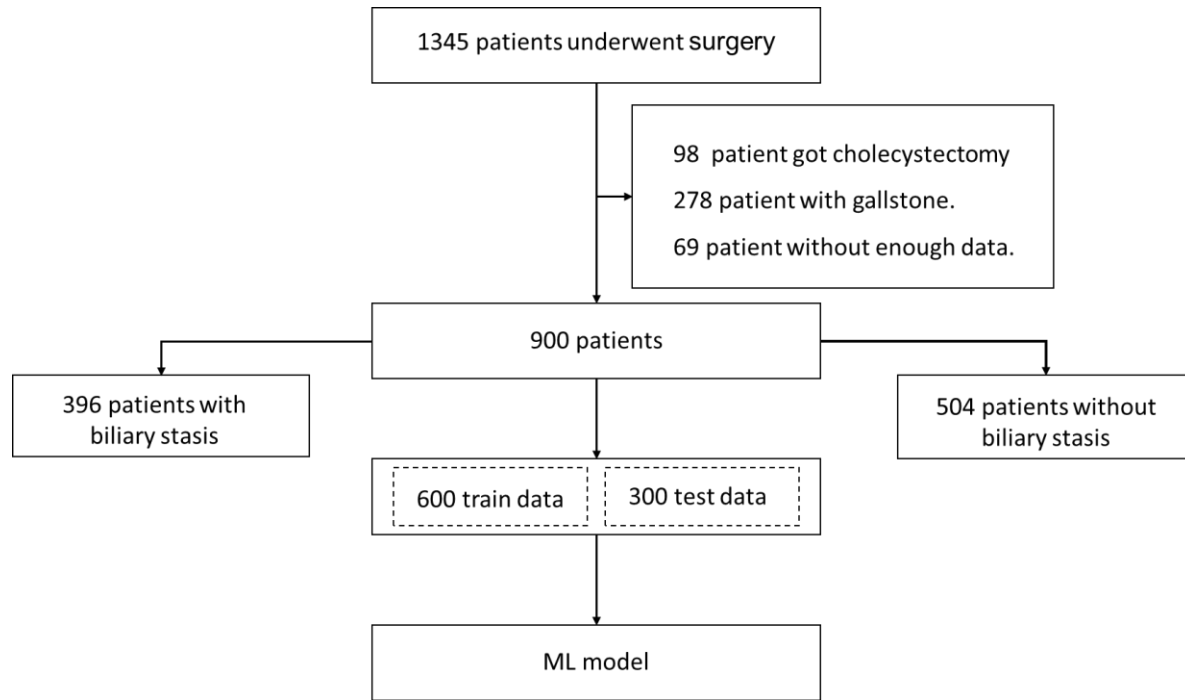
Receiver Operating Characteristic Curve Analysis

Evaluation

PART 04[▽]

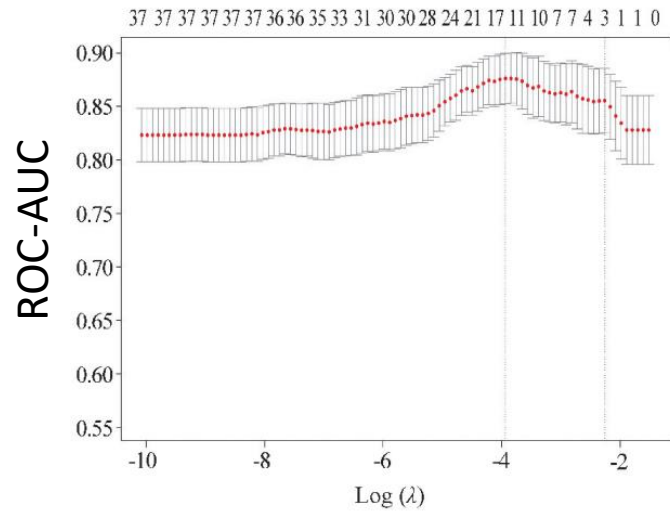
Research Results

1. Flowchart and patients features

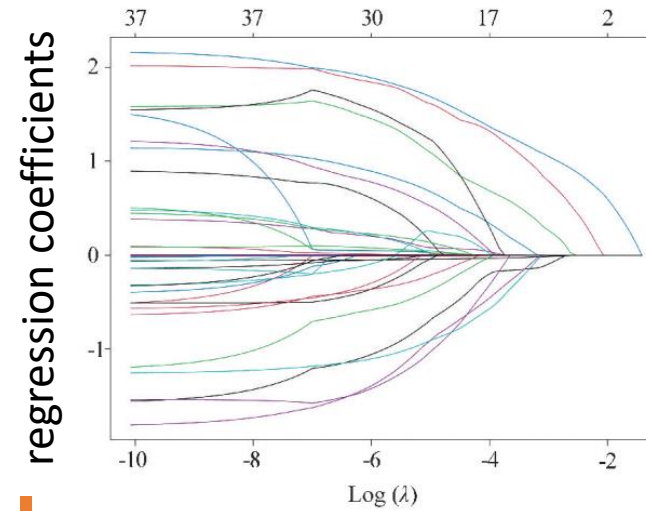


	Train data	Test data	X2	p
Age(year)	36.51±18.32	38.22±17.56	-1.313	0.192
Sex(male)	378 (63.0%)	181 (60.3%)	1.082	0.312
Diabetes (yes)	45.21	43.67	0.882	0.232
FLD (yes)	501 (83.5%)	233 (77.7%)	2.011	0.712
TC (high)	458 (76.3%)	214 (71.3%)	0.151	0.281
HDL-C (low)	398 (66.3%)	184 (61.3%)	0.523	0.322
LDL-C (high)	411 (68.5%)	221 (73.7%)	0.432	0.115
UA (high)	256 (42.7%)	138 (46.0%)	0.895	0.354
AST (high)	429 (71.5%)	189 (63.0%)	0.832	0.655
biliary stasis	275 (45.8%)	121 (40.3%)	0.422	0.562

2. Screening of clinical variables



The tuning interval for the Lasso regression parameter λ ;



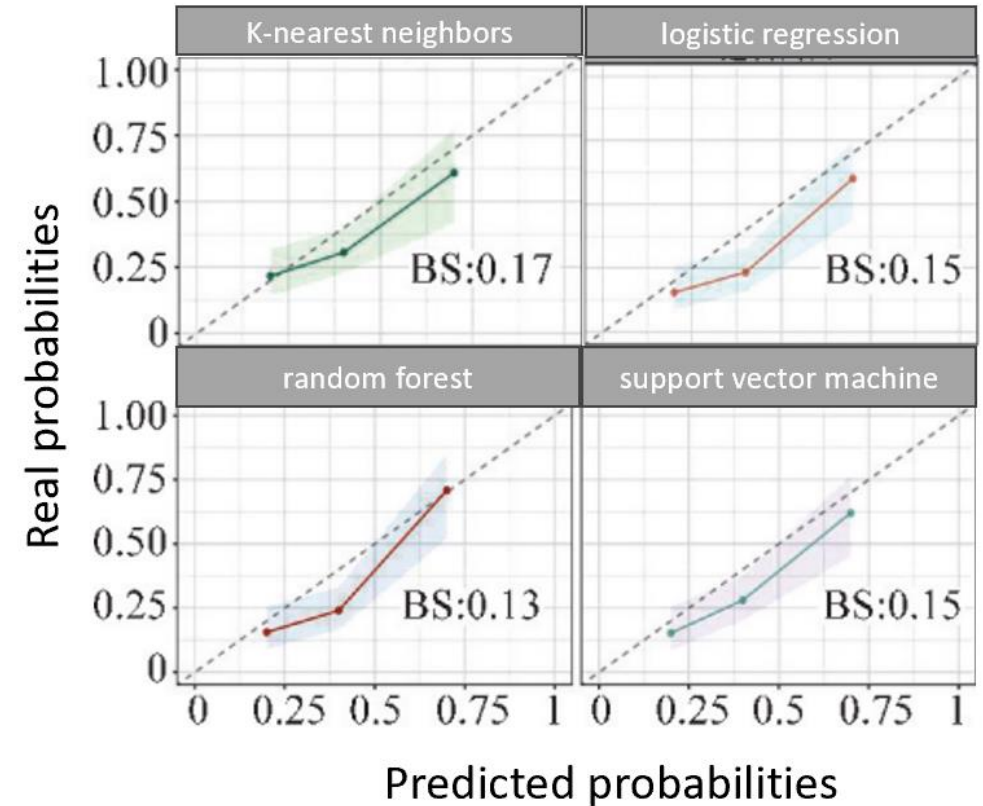
The regression coefficients for each clinical variable;

3. Superiority of the Random Forest Model

Machine Learning	Train				Test			
	AUC	95%CI	sensitivity	specificity	AUC	95%CI	sensitivity	specificity
logistic regression	0.822	(0.731, 0.905)	69.1%	61.0%	0.744	(0.687, 0.930)	64.9%	63.4%
support vector machine	0.715	(0.662, 0.801)	65.4%	60.2%	0.693	(0.600, 0.711)	55.3%	69.2%
K-nearest neighbors	0.818	(0.733, 0.872)	70.2%	66.9%	0.713	(0.609, 0.842)	61.2%	66.2%
random forest	0.873	(0.832, 0.912)	71.9%	76.0%	0.771	(0.602, 0.892)	68.3%	66.5%

The diagnostic performance of each model in the training and test sets;

Calibration curves for various machine learning models;



PART 05[▽]
Conclusion

Summary of Key Findings



Effectiveness of the Random Forest Algorithm in predicting early biliary stasis;

improving patient care and management through the use of the predictive nomogram;

Acknowledgment



中日友好医院
CHINA-JAPAN FRIENDSHIP HOSPITAL



减重糖尿病手术
健康管理中心

*We look forward to
more co-operation!*

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