G12

### 

## **Diabetes Health Indicators**

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### 資料集介紹

資料前處理

資料清理、處理缺失值

特徵說明



#### CDC Diabetes Health Indicators (External)



The Diabetes Health Indicators Dataset contains healthcare statistics and lifestyle survey information about people in general along with their diagnosis of diabetes. The 35 features consist of some demographics, lab test results, and answers to survey questions for each patient. The target variable for classification is whether a patient has diabetes, is pre-diabetic, or healthy.

**Dataset Characteristics** 

Subject Area

Associated Tasks

Tabular, Multivariate

Health and Medicine

Classification

Feature Type

# Instances

# Features

Categorical, Integer 253680

By. UC Irvine Machine Learning Repository

整數型特徵

BMI, GenHlth, Age, Education...

二元分類特徵

HighBP, HighChol, Smoker, Stroke...





### 整數型特徵 (Integer Features)

BMI(體重指數) 0 - 100

GenHlth(自我報告健康狀)

1 = 優, 2 = 非常好, 3 = 良好,

4 = 一般, 5 = 差

MentHIth(心理健康):

有幾天心理健康狀況不佳的 1 - 30 day

PhysHlth(身體健康)

有無感冒或是受傷的情況 1~30 day

Age(年齡):

年齡區間, 分類 1 - 13 (18~)

#### Education(教育程度)

1: 未受教育 OR 幼稚園

2:小學~國中

3:國中畢業

4:高中畢業

5: 大學 or 五專

6:大學畢業

#### Income(收入水平)

1: 低於\$10,000

2: \$10,000 ~ \$14,999

3: \$15,000 ~ \$19,999

4: \$20,000 ~ \$24,999

5: \$25,000 ~ \$34,999

6: \$35,000 ~ \$50,000

7: 高於\$75,000

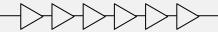




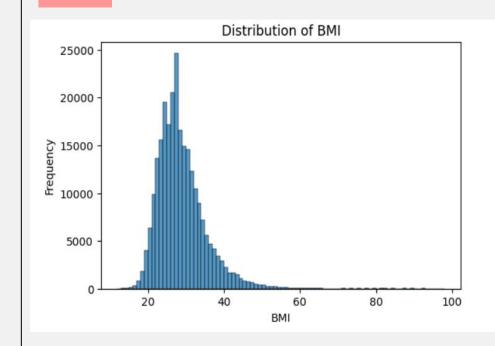
### 二元分類特徵 (Binary Features)

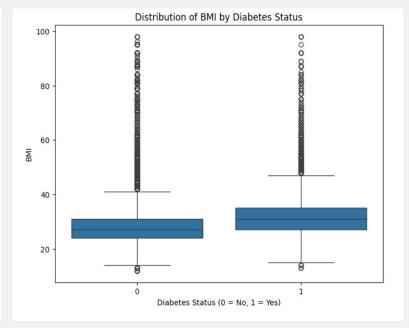
- HighBP (是否有高血壓)
- HighChol (是否有高膽固醇)
- CholCheck (近五年有無進行膽固醇檢 查)
- Smoker (一生至少吸過100支煙)
- Stroke (是否有中風)
- Heart Disease or Attack (是否有冠狀動脈 心臟病或心肌梗塞)
- PhysActivity (過去30天內的不包括工作的 體力活動)
- Fruits (每日水果攝入)

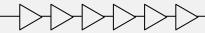
- Veggies (毎日蔬菜攝入)
- HvyAlcoholConsump (成年男性每週 飲酒超過14杯,成年女性每週飲酒超過 7杯)
- AnyHealthcare (是否有任何形式的醫療保障)
- NoDocbcCost (曾因費用原因未看醫生)
- DiffWalk (感覺行走或爬樓梯困難)
- Sex (性別)
  - 0: 女性, 1: 男性



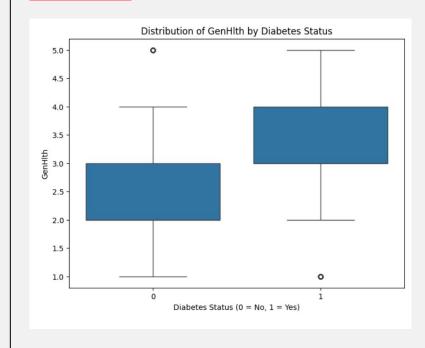
#### **BMI**



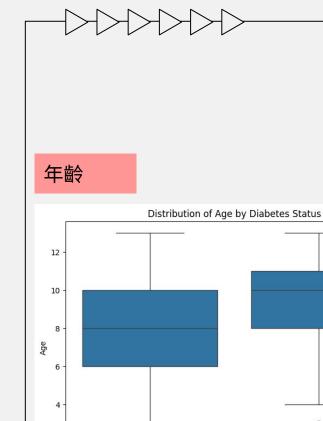




### 健康自評



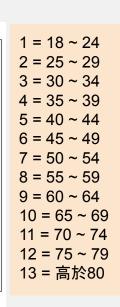
- 1 = 優 2 = 非常好 3 = 良好
- 4 = 一般
- 5 = 差



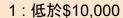
0

Diabetes Status (0 = No. 1 = Yes)

2



收入



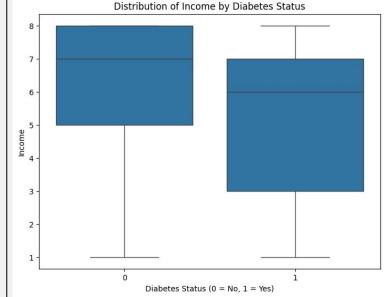
2:\$10,000 ~ \$14,999

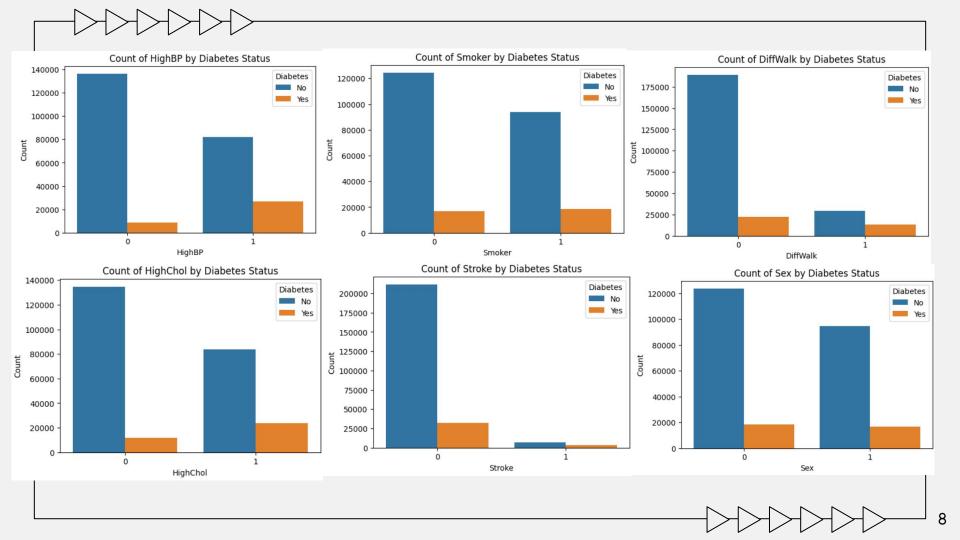
3: \$15,000 ~ \$19,999 4: \$20,000 ~ \$24,999

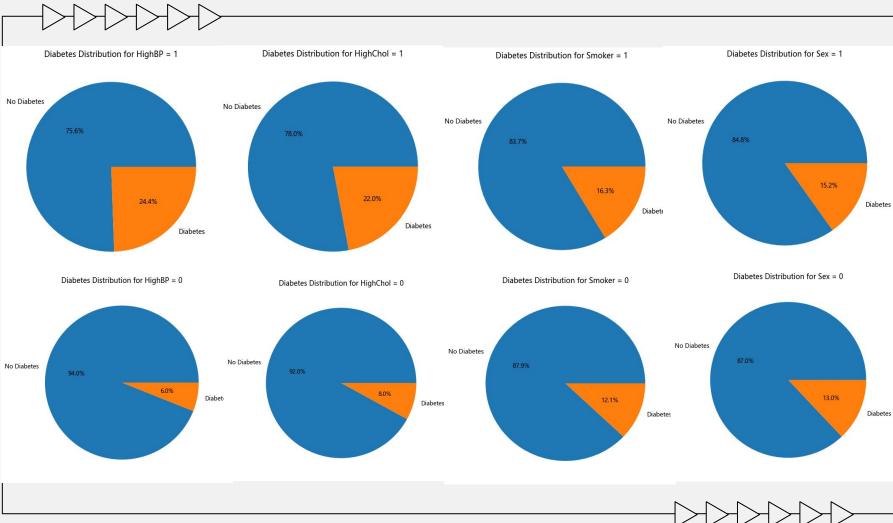
5 : \$25,000 ~ \$24,999

6: \$35,000 ~ \$49,999

7:\$50,000~\$75,000 8:高於\$75,000









95% CI

## 整數型特徵boostrap

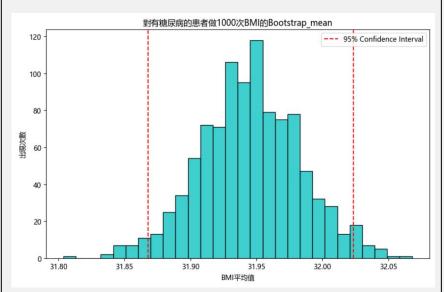
正常範圍: 18.5≦BMI<24

過重: 24≦BMI<27 輕度肥胖: 27≦BMI<30

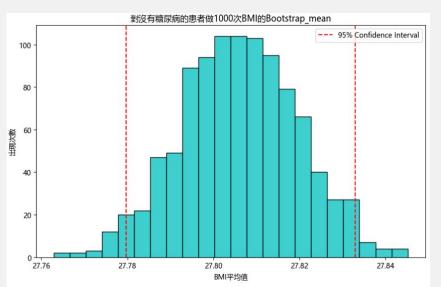
中度肥胖: 30≦BMI<35

重度肥胖: BMI≥35

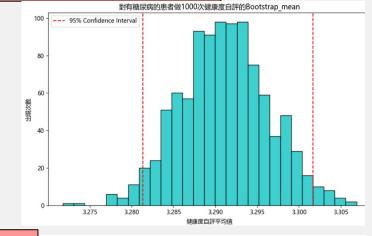
#### 有糖尿病之BMI [31.87, 32.02]

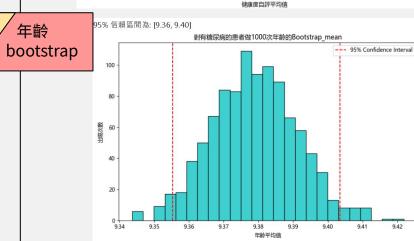


### 無糖尿病之BMI [[27.78, 27.83]

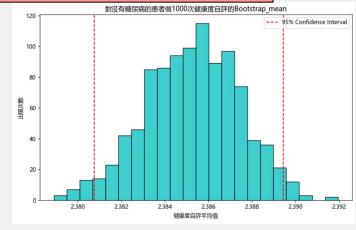


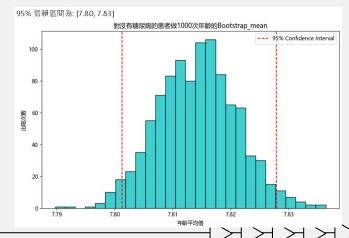
### 有糖尿病之健康度自評[3.28, 3.30] 中位數皆為3





### 無糖尿病之健康度自評[2.38, 2.39] 中位數皆為2

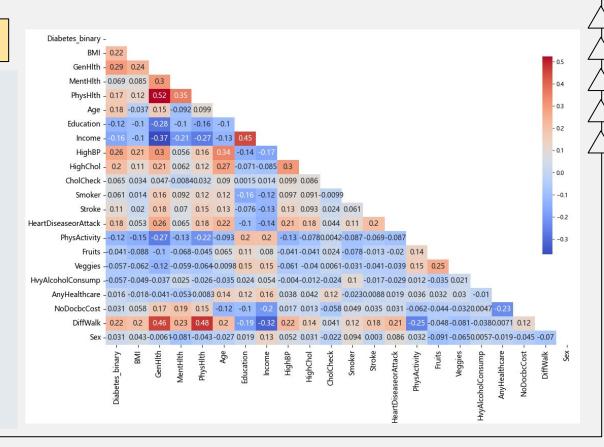




## 各特徵相關性分析 (Pearson correlation)

#### 相關性較高之組合(>0.3)

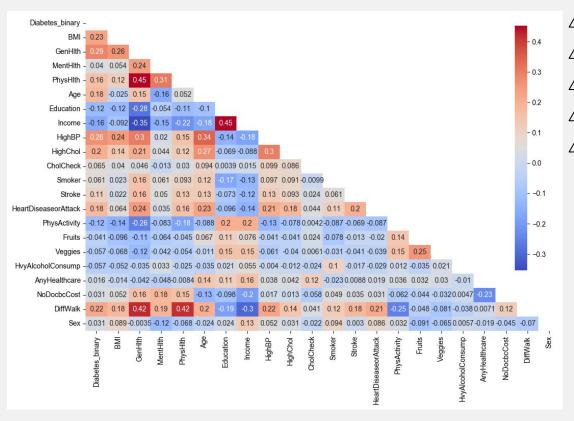
- GenHlth(自我報告健康狀):
  - MenHlth(0.3),
  - o PhysHlth(0.52)
  - Income(-0.37)
  - HighBP(0.3)
- MentHlth(心理健康)
  - PhysHlth(0.35)
- PhysHlth(身體健康)
  - DiffWalk(0.48)
  - Age: HighBP(0.34)
- Education
  - Income(0.45)
- Income:
  - $\circ$  DiffWalk(-0.32)
- HighBP(高血壓)
  - HighCol(0.3)



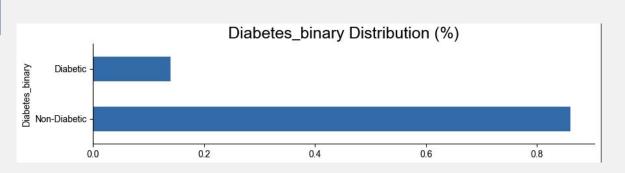
## 各特徵相關性分析 (spearman correlation)

#### 相關性較高之組合(>0.3)

- PhysHllth(身體健康)
  - GenHllth(0.45)
- Income
  - GenHllth(-0.35)
  - Education(0.45)
- DiffWalk
  - GenHllth(0.42)
  - Physhllth(0.42)
  - $\circ$  Income(-0.3)
- HighBP(高血壓)
  - GenHllth(0.3)
  - Age(0.34)
  - HighChol(0.3)

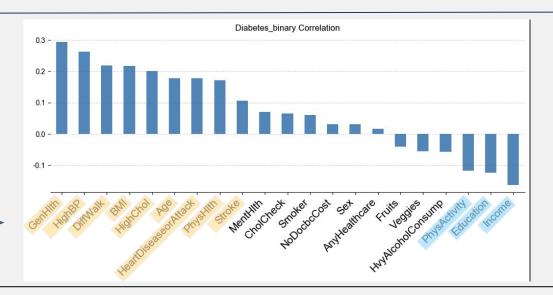


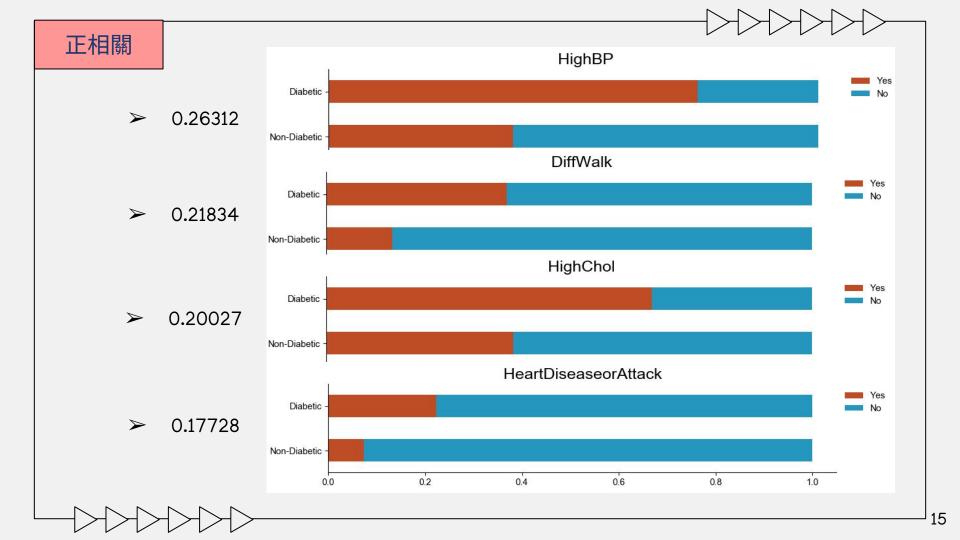
資料集糖尿病比例 14:86

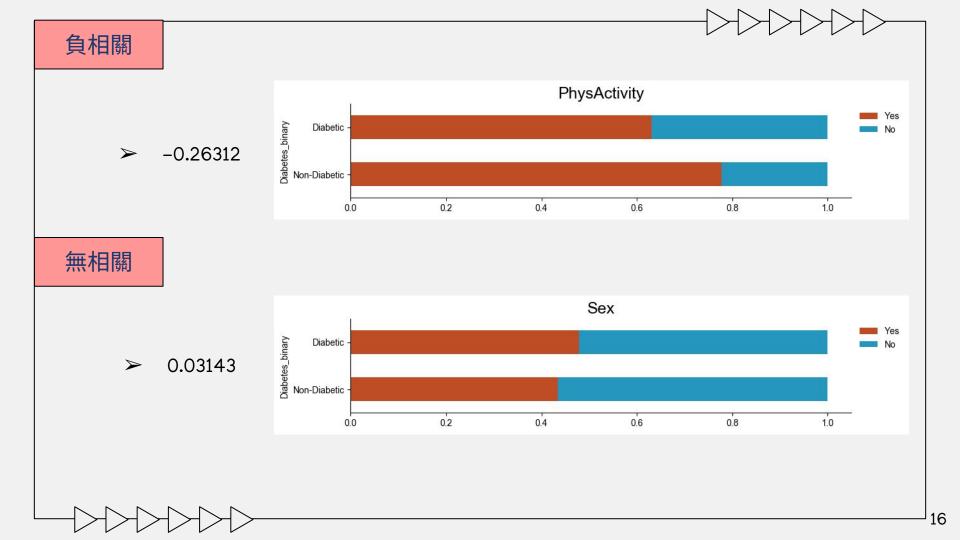




設定 alpha = 0.1 ——







### 卡方檢定

#### 顯著的特徵:

DiffWalk

HighBP

HeartDiseaseorAttack

HighChol

Stroke

PhysActivity

HvyAlcoholConsump

Smoker

NoDocbcCost

Fruits

Veggies

Sex

CholCheck

#### 非顯著的特徵:

AnyHealthcare

	ChiSqr_Score	ChiSqr_pValue
DiffWalk	10059.51	0.00e+00
HighBP	10029.01	0.00e+00
HeartDiseaseorAttack	7221.98	0.00e+00
HighChol	5859.71	0.00e+00
Stroke	2725.23	0.00e+00
PhysActivity	861.89	1.89e-189
HvyAlcoholConsump	779.42	1.61e-171
Smoker	521.98	1.57e-115
NoDocbcCost	229.54	7.50e-52
Fruits	154.29	2.00e-35
Veggies	153.17	3.52e-35
Sex	140.25	2.35e-32
CholCheck	39.72	2.94e-10
AnyHealthcare	3.28	0.07



顯著的特徵 (p<0.05):

HeartDiseaseorAttack DiffWalk HighBP HighChol

#### 非顯著的特徵:

Fruits

Smoker

HvyAlcoholConsump

NoDocbcCost

Sex

Veggies

PhysActivity

CholCheck

AnyHealthcare

Stroke

Random sample 100

	ChiSqr_Score	ChiSqr_pValue
HeartDiseaseorAttack	3.07e+00	0.08
DiffWalk	2.43e+00	0.12
HighBP	1.98e+00	0.16
HighChol	1.98e+00	0.16
Fruits	7.13e-01	0.40
Smoker	3.73e-01	0.54
HvyAlcoholConsump	3.03e-01	0.58
NoDocbcCost	3.03e-01	0.58
Sex	4.74e-02	0.83
Veggies	2.11e-02	0.88
PhysActivity	1.57e-02	0.90
CholCheck	5.57e-03	0.94
AnyHealthcare	0.00e+00	1.00
Stroke	NAN	NAN



顯著的特徵 (p<0.05):

Stroke HighChol HighBP

DiffWalk

NoDocbcCost

HeartDiseaseorAttack

#### 非顯著的特

徵:HvyAlcoholConsump

Fruits

Smoker

PhysActivity

Veggies

Sex

CholCheck

AnyHealthcare

Random sample 300

	ChiSqr_Score	ChiSqr_pValue
Stroke	1.25e+01	4.17e-04
HighChol	1.01e+01	1.46e-03
HighBP	9.76e+00	1.78e-03
DiffWalk	6.93e+00	8.46e-03
NoDocbcCost	5.45e+00	1.95e-02
HeartDiseaseorAttack	4.50e+00	3.39e-02
HvyAlcoholConsump	8.51e-01	3.56e-01
Fruits	5.66e-01	4.52e-01
Smoker	3.15e-01	5.75e-01
PhysActivity	8.25e-02	7.74e-01
Veggies	4.14e-02	8.39e-01
Sex	3.75e-02	8.46e-01
CholCheck	2.80e-02	8.67e-01
AnyHealthcare	5.05e-03	9.43e-01

**PVALUE** 

0

### 邏輯回歸

BMI, GenHlth, MentHlth, PhysHlth, Age, Education, Income

**Feature** 

Income

Feature: BMI							
		Logit Re	gressio	n Result:	5		
	=======		======	=======		=======	========
Dep. Variable:	Dia	abetes_bin	ary N	lo. Observ	vations:		253680
Model:		Lo	git D	of Residua	als:		253678
Method:			MLE D	of Model:			1
Date:	Mon.	10 Jun 2	024 P	seudo R-	sau.:		0. 04904
Time:	,	17: 35		og-Likel			-97401.
converged:		Т		L-Null:		_	1. 0242e+05
Covariance Type	:	nonrob		LR p-val	ue:		0. 000
	coef	std err	======	z l	====== P> z	Γ0. 025	0. 9751
		ci i		- '		[0.027	0. 7. 7.
const -	4. 1406	0. 025	-162. 6	88	0. 000	-4. 190	-4. 091
BMI	0. 0786	0.001	96. 5	05 (	0. 000	0. 077	0.080
	========		======				

BMI	0.0786	0
GenHlth	0.7875	0
MentHlth	0.0231	0
PhysHlth	0.0441	0
Age	0.1887	0
Education	-0.3393	0

-0.2071

COEF

- BMI, GenHlth, PhysHlth, Age, Education, Income
  - 影響非常顯著
- MentHlth
  - 有一定的預測作用,但相較於其他變數效應較小

< 0.05

BMI, GenHlth, MentHlth, PhysHlth, Age, Education, Income

Feature: BMI Logit Regression Results Dep. Variable: Diabetes binary No. Observations: Model: Logit Df Residuals: 98 Df Model: Method: MLE Tue, 11 Jun 2024 Pseudo R-squ.: 0.06337 Date: Time: 15: 34: 42 Log-Likelihood: -34. 367 LL-Null: -36, 692 converged: True Covariance Type: nonrobust LLR p-value: 0.03104

P>|z|

0.001

[0.025

-8. 378

0.011

0.9751

-2.139

0.214

•	BMI.	GenHlth.	MentHlth

coef

-5. 2585

const

BMT

std err

1. 592

0.052

-3.303

- 顯著差異
- PhysHlth, Age, Education, Income
  - 沒有顯著差異

Feature	COEF	PVALUE			
ВМІ	0.1128	0.029			
GenHlth	1.0771	0.006			
MentHlth	0.0840	0.026			
PhysHlth	0.0076	0.878			
Age	0.2357	0.062			
Education	-0.0899	0.787			
Income	-0.1211	0.423			

>0.1

Random sample 100

### 邏輯回歸

BMI, GenHlth, MentHlth, PhysHlth, Age, Education, Income

< 0.05

Logit Regression Results										
===========	======	=====	=====	=====	=====		========	=======		
Dep. Variable:		Diabe	tes bi	narv	No.	Observations:		300		
Model:			_	ogit	Df R	esiduals:		298		
Method:			_	MLE		odel:		1		
Date:		Tuo 1	4 Jun			do R-squ.:		0. 01609		
		rue, 1	1 Jun							
Time:			16: 0	7: 22	Log-	Likelihood:		-121. 31		
converged:				True	LL-N	ull:		-123. 29		
Covariance Type	e.		nonro	hust	IIR	p-value:		0. 04641		
						,				
	coef	st	d err		Z	P> z	[0. 025	0. 975]		
const	-2.9608		0. 607	- 4	. 881	0.000	-4. 150	-1. 772		
BMT	0.0403		0. 020	2	2. 054	0. 040	0. 002	0. 079		
5	0. 0.02		0. 020				0.002			

Feature	COEF	PVALUE			
BMI	0.0403	0.04			
GenHlth	0.8144	0			
MentHlth	0.0354	0.033			
PhysHlth	0.0635	0			
Age	0.2032	0.002			
Education	-0.1249	0.464			
Income	-0.2087	0.005			

- BMI, GenHlth, MentHlth, PhysHlth, Age, Income
  - 顯著差異
- Education
  - 沒有顯著差異

Random sample 300

### 邏輯回歸—預測糖尿病

```
Optimization terminated successfully.
        Current function value: 0.355211
        Iterations 7
                         Logit Regression Results
Dep. Variable:
                    Diabetes binary No. Observations:
                                                                  253680
Model:
                             Logit Df Residuals:
                                                                  253675
Method:
                               MIF Df Model:
                Sat, 08 Jun 2024 Pseudo R-squ.:
Date:
                                                                  0.1202
                         17:28:27 Log-Likelihood:
Time:
                                                                 -90110.
converged:
                              True LL-Null:
                                                             -1.0242e+05
Covariance Type: nonrobust LLR p-value:
                                                                   0.000
                 coef
                         std err
                                               P>|z|
                                                         TØ.025
                                                                    0.9751
               -4.9879
                          0.044
                                 -113.165
                                               0.000
                                                         -5.074
                                                                    -4.901
const
BMT
               0.0832
                          0.001
                                   95.691
                                               0.000
                                                       0.082
                                                                     0.085
               0.2052
                          0.002
                                   87.138
                                               0.000
                                                        0.201
                                                                     0.210
PhysActivity -0.3094
                                                         -0.335
                          0.013
                                   -23.421
                                               0.000
                                                                    -0.284
               -0.1495
                                   -52.909
                                               0.000
                                                         -0.155
                                                                    -0.144
                           0.003
```

```
Confusion Matrix:
[[216158 2176]
 [ 33622 1724]]
Classification Report:
              precision
                          recall f1-score support
                            0.99
                                      0.92
                   0.87
                                              218334
                  0.44
                            0.05
                                      0.09
                                              35346
                                              253680
    accuracy
                                      0.86
                  0.65
                            0.52
                                              253680
   macro avg
                                      0.51
weighted avg
                   0.81
                            0.86
                                      0.81
                                              253680
```

	feature	VIF
0	const	45.772267
1	BMI	1.031343
2	Age	1.025130
3	PhysActivity	1.065332
4	Income	1.060973

## Cochran-Armitage趨勢檢定

	GenHlth	MentHlth	PhysHlth	Age	Education	Income
Null_mean (H0 expected)	53421.67	112568.96	149940.58	248557.29	143166.63	178634.26
null_sd (H0 expected)	186.36	1292.92	1520.56	532.71	171.94	361.24
p-value	0.0	4.95e-267	0.0	0.0	0.0	0.0
Statistic (actual value)	80977.0	157707.0	281159.0	296166.0	132389.0	148810.0
z-score (correlation direction)	147.86	34.91	86.30	89.37	-62.68	-82.56

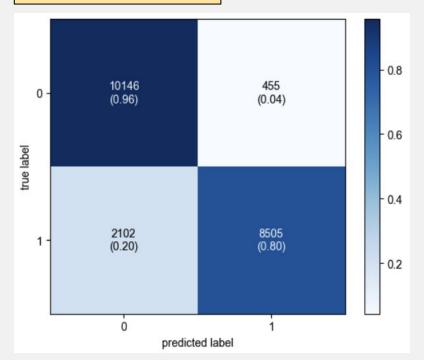
24

### **Data Mining Techniques**

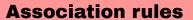
#### Classification

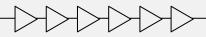
- Decision tree model
- Train-test: 0.7-0.3
- Depth: 12
- 特徵取用:BMI, GenHIth,
   MentHIth, PhysHIth, Age,
   Education, Income,
   HighBP, HighChol,
   CholCheck, Smoker,
   Stroke,
   HeartDiseaseorAttack,
   PhysActivity, Veggies,
   HvyAlcoholConsump,
   DiffWalk

#### **Confusion matrix**









### 有糖尿病

#### min\_sup = 0.7, min\_conf = 0.7

<b>►71</b> △J												
		antecedents	consequents	antecedent support	consequent support	support	confidence	lift	leverage	conviction	zhangs_metric	
	0	(CholCheck)	(AnyHealthcare)	0.993182	0.959769	0.954705	0.961259	1.001552	0.001480	1.038459	0.227327	
	1	(AnyHealthcare)	(CholCheck)	0.959769	0.993182	0.954705	0.994723	1.001552	0.001480	1.292203	0.038527	
	2	(AnyHealthcare)	(HighBP)	0.959769	0.752674	0.723844	0.754186	1.002009	0.001451	1.006152	0.049841	
		(HighBP)	(AnyHealthcare)	0.752674	0.959769	0.723844	0.961697	1.002009	0.001451	1.050345	0.008107	
	4	(AnyHealthcare)	(Veggies)	0.959769	0.756408	0.728060	0.758578	1.002869	0.002083	1.008988	0.071102	
	5	(Veggies)	(AnyHealthcare)	0.756408	0.959769	0.728060	0.962522	1.002869	0.002083	1.073465	0.011743	
	6	(CholCheck)	(HighBP)	0.993182	0.752674	0.748628	0.753767	1.001453	0.001086	1.004442	0.212806	
	7	(HighBP)	(CholCheck)	0.752674	0.993182	0.748628	0.994625	1.001453	0.001086	1.268492	0.005867	
	8	(CholCheck)	(Veggies)	0.993182	0.756408	0.751514	0.756673	1.000350	0.000263	1.001088	0.051317	
	9	(Veggies)	(CholCheck)	0.756408	0.993182	0.751514	0.993529	1.000350	0.000263	1.053725	0.001436	
	10	(CholCheck, AnyHealthcare)	(HighBP)	0.954705	0.752674	0.720647	0.754838	1.002875	0.002066	1.008828	0.063298	
	11	(CholCheck, HighBP)	(AnyHealthcare)	0.748628	0.959769	0.720647	0.962624	1.002975	0.002137	1.076389	0.011799	
	12	(AnyHealthcare, HighBP)	(CholCheck)	0.723844	0.993182	0.720647	0.995583	1.002418	0.001738	1.543774	0.008735	
	13	(CholCheck)	(AnyHealthcare, HighBP)	0.993182	0.723844	0.720647	0.725595	1.002418	0.001738	1.006379	0.353799	
	14	(AnyHealthcare)	(CholCheck, HighBP)	0.959769	0.748628	0.720647	0.750855	1.002975	0.002137	1.008939	0.073723	
	15	(HighBP)	(CholCheck, AnyHealthcare)	0.752674	0.954705	0.720647	0.957450	1.002875	0.002066	1.064514	0.011592	
	16	(CholCheck, AnyHealthcare)	(Veggies)	0.954705	0.756408	0.724382	0.758749	1.003095	0.002235	1.009705	0.068127	
	17	(CholCheck, Veggies)	(AnyHealthcare)	0.751514	0.959769	0.724382	0.963897	1.004301	0.003102	1.114340	0.017235	
	18	(AnyHealthcare, Veggies)	(CholCheck)	0.728060	0.993182	0.724382	0.994948	1.001779	0.001286	1.349711	0.006529	
	19	(CholCheck)	(AnyHealthcare, Veggies)	0.993182	0.728060	0.724382	0.729355	1.001779	0.001286	1.004785	0.260416	
	20	(AnyHealthcare)	(CholCheck, Veggies)	0.959769	0.751514	0.724382	0.754746	1.004301	0.003102	1.013179	0.106451	
	21	(Veggies)	(CholCheck, AnyHealthcare)	0.756408	0.954705	0.724382	0.957660	1.003095	0.002235	1.069796	0.012668	

#### antecedents\_len > 1

	antecedents	consequents	antecedent support	consequent support	support	confidence	lift	leverage	conviction	zhangs_metric	antecedents_len
10	(CholCheck, AnyHealthcare)	(HighBP)	0.954705	0.752674	0.720647	0.754838	1.002875	0.002066	1.008828	0.063298	2
11	(CholCheck, HighBP)	(AnyHealthcare)	0.748628	0.959769	0.720647	0.962624	1.002975	0.002137	1.076389	0.011799	2
12	(AnyHealthcare, HighBP)	(CholCheck)	0.723844	0.993182	0.720647	0.995583	1.002418	0.001738	1.543774	0.008735	2
16	(CholCheck, AnyHealthcare)	(Veggies)	0.954705	0.756408	0.724382	0.758749	1.003095	0.002235	1.009705	0.068127	2
17	(CholCheck, Veggies)	(AnyHealthcare)	0.751514	0.959769	0.724382	0.963897	1.004301	0.003102	1.114340	0.017235	2
18	(AnyHealthcare, Veggies)	(CholCheck)	0.728060	0.993182	0.724382	0.994948	1.001779	0.001286	1.349711	0.006529	2



#### min\_sup = 0.7, min\_conf = 0.7

	antecedents	consequents	antecedent support	consequent support	support	confidence	lift	leverage	conviction	zhangs_metric
0	(CholCheck)	(AnyHealthcare)	0.957730	0.949641	0.914800	0.955176	1.005828	0.005300	1.123465	0.137070
1	(AnyHealthcare)	(CholCheck)	0.949641	0.957730	0.914800	0.963311	1.005828	0.005300	1.152128	0.115054
2	(AnyHealthcare)	(PhysActivity)	0.949641	0.776943	0.741474	0.780794	1.004957	0.003657	1.017568	0.097943
3	(PhysActivity)	(AnyHealthcare)	0.776943	0.949641	0.741474	0.954349	1.004957	0.003657	1.103110	0.022112
4	(AnyHealthcare)	(Veggies)	0.949641	0.820326	0.781669	0.823121	1.003407	0.002654	1.015801	0.067425
5	(Veggies)	(AnyHealthcare)	0.820326	0.949641	0.781669	0.952877	1.003407	0.002654	1.068659	0.018898
6	(CholCheck)	(PhysActivity)	0.957730	0.776943	0.745124	0.778011	1.001375	0.001023	1.004813	0.032487
7	(PhysActivity)	(CholCheck)	0.776943	0.957730	0.745124	0.959047	1.001375	0.001023	1.032158	0.006156
8	(CholCheck)	(Veggies)	0.957730	0.820326	0.786451	0.821162	1.001019	0.000801	1.004674	0.024081
9	(Veggies)	(CholCheck)	0.820326	0.957730	0.786451	0.958706	1.001019	0.000801	1.023633	0.005665
16	(CholCheck, AnyHealthcare)	(PhysActivity)	0.914800	0.776943	0.714790	0.781362	1.005688	0.004043	1.020213	0.066385
11	(CholCheck, PhysActivity)	(AnyHealthcare)	0.745124	0.949641	0.714790	0.959290	1.010160	0.007189	1.236999	0.039461
12	(AnyHealthcare, PhysActivity)	(CholCheck)	0.741474	0.957730	0.714790	0.964012	1.006560	0.004658	1.174573	0.025208
13	(CholCheck)	(AnyHealthcare, PhysActivity)	0.957730	0.741474	0.714790	0.746338	1.006560	0.004658	1.019175	0.154175
14	(AnyHealthcare)	(CholCheck, PhysActivity)	0.949641	0.745124	0.714790	0.752695	1.010160	0.007189	1.030612	0.199723
15	(PhysActivity)	(CholCheck, AnyHealthcare)	0.776943	0.914800	0.714790	0.920004	1.005688	0.004043	1.065047	0.025357
16	(CholCheck, AnyHealthcare)	(Veggies)	0.914800	0.820326	0.753561	0.823744	1.004167	0.003127	1.019393	0.048703
17	(CholCheck, Veggies)	(AnyHealthcare)	0.786451	0.949641	0.753561	0.958179	1.008991	0.006715	1.204154	0.041726
18	3 (AnyHealthcare, Veggies)	(CholCheck)	0.781669	0.957730	0.753561	0.964041	1.006589	0.004933	1.175497	0.029983
19	(CholCheck)	(AnyHealthcare, Veggies)	0.957730	0.781669	0.753561	0.786820	1.006589	0.004933	1.024161	0.154865
26	(AnyHealthcare)	(CholCheck, Veggies)	0.949641	0.786451	0.753561	0.793522	1.008991	0.006715	1.034244	0.176941
21	(Veggies)	(CholCheck, AnyHealthcare)	0.820326	0.914800	0.753561	0.918612	1.004167	0.003127	1.046834	0.023094
	17 SEC. 18	A 187 - 187 - 187 - 187 - 187 - 187 - 187 - 187 - 187 - 187 - 187 - 187 - 187 - 187 - 187 - 187 - 187 - 187 -	·							

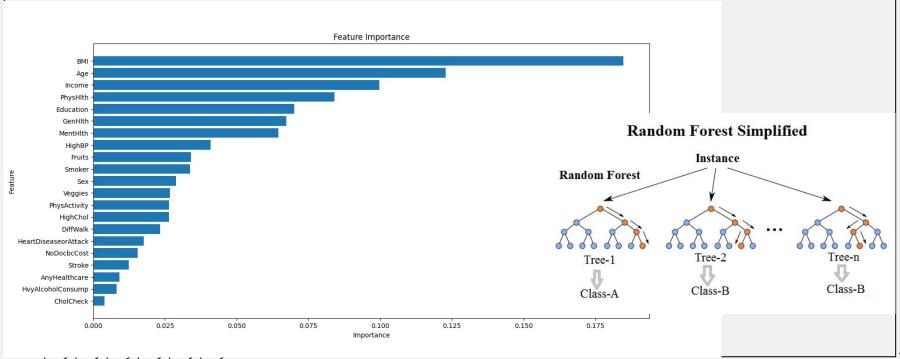
#### antecedents\_len > 1

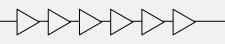
	antecedents	consequents	antecedent support	consequent support	support	confidence	lift	leverage	conviction	zhangs_metric	antecedents_len
10	(CholCheck, AnyHealthcare)	(PhysActivity)	0.914800	0.776943	0.714790	0.781362	1.005688	0.004043	1.020213	0.066385	2
11	(CholCheck, PhysActivity)	(AnyHealthcare)	0.745124	0.949641	0.714790	0.959290	1.010160	0.007189	1.236999	0.039461	2
12	(AnyHealthcare, PhysActivity)	(CholCheck)	0.741474	0.957730	0.714790	0.964012	1.006560	0.004658	1.174573	0.025208	2
16	(CholCheck, AnyHealthcare)	(Veggies)	0.914800	0.820326	0.753561	0.823744	1.004167	0.003127	1.019393	0.048703	2
17	(CholCheck, Veggies)	(AnyHealthcare)	0.786451	0.949641	0.753561	0.958179	1.008991	0.006715	1.204154	0.041726	2
18	(AnyHealthcare, Veggies)	(CholCheck)	0.781669	0.957730	0.753561	0.964041	1.006589	0.004933	1.175497	0.029983	2

### 持徵選擇

原始Random Forest模型

模型特徵重要性排序





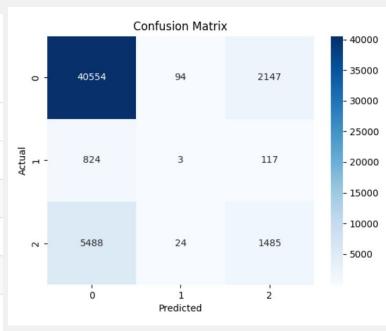
### 選擇前十個特徵並評估

#### **Confusion matrix**

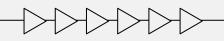
Accuracy: 0.8286423841059603

F1 Score: 0.8012423971371091

	precision	recall	f1-score	support
0	0.87	0.95	0.90	42795
1	0.02	0.00	0.01	944
2	0.40	0.21	0.28	6997
macro avg	0.43	0.39	0.40	50736
weighted avg	0.78	0.83	0.80	50736







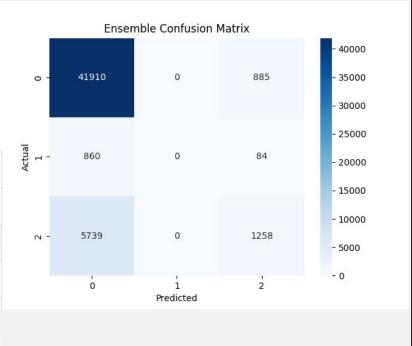
## 嘗試模型集成改良

# Random Forest / Logistic Regression / Gradient Boosting

Ensemble Accuracy: 0.8508356985178177

Ensemble F1 Score: 0.8119625305245382

	precision	recall	f1-score	support
0	0.86	0.98	0.92	42795
1	0.00	0.00	0.00	944
2	0.56	0.18	0.27	6997
macro avg	0.48	0.39	0.40	50736
weighted avg	0.81	0.85	0.81	50736





### Reference

- Dataset
  - CDC Diabetes Health Indicators
  - https://archive.ics.uci.edu/dataset/891/cdc+diabetes+health+indicators
- Tools
  - chatGPT
  - Python
    - fetch\_ucirepo
    - sklearn
    - scipy
    - statsmodels
    - mlxtend

#