Data Science Project: Epigenetic Classifier for Twin Zygosity

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Overview

What is zygosity of Twins and why are they important?

- Monozygotic twin (MZ) or identical twin
- Dizygotic twin (DZ) or fraternal twin

Importance:

- Cheaper classification method compared to traditional questionnaires from hospitals
- It helps better understand disorders that's correlated with twins, e.g. spine disorders.

Overview

Aim 1: Test the provided classifier by van Dongen et al. [2021] on a new data set to check its performance.

Aim 2: Train a new classifier based on machine learning methods to obtain a classifier with better performance compared to classifier provided by van Dongen et al. [2021].

We **finished Aim 1** within the time frame of the 1st semester and will present results in the following section

Group Members

Zexi Liu - Review Review Literature

Atefeh Zamani - Review Literature

Haoze Xia - Coding

Ni Zhang - Coding

Tianyu Zhou - Coding

Agile development cycle

Host Organisation

Twins Research Centre of Epidemiology and Biostatistics UoM

Dr. Shuai Li and Prof. John Hopper

What is the Data Science Problem?

- Data linkage of medical data gives more opportunity for research in recent years
- The organisation wants to explore machine learning techniques to this sudden available data, and this project is to explore options.

DataSets

- In van Dongen et al. [2021],
 - One dataset to train (1957 Samples), Five datasets for testing (4485 Samples).

	N Total	N MZ Twins	N DZ Twins	N Family Members
NTR	1957	924	1033	237
E-Risk	1164	470	694	-
FTC	1708	559	1149	-
TwinksUK	492	395	97	-
BSGS	356	134	222	257
NTR	765	564	201	

Table 1: Datasets analysed in van Dongen et al. (2021)

Difficulties and Classifiers

Difficulties:

- Dimension of the datasets:
 - Big datasets (450K columns)
 - Reducing the dimension of the data
 - Meta-Analysis
 - Monozygotic twinning is associated with a persistent DNA methylation profile in adult somatic tissues.
 - This MZ-signature comprises 834 CpG sites.

Classifiers:

- Classifier 13: Classify based on their zygosity (MZ or DZ)
- Classifier 14: Classify MZ twins from the rest (DZ twins and family members).



Datasets

- The datasets we have access to
 - E-Risk (1658 Samples), 4 other datasets (1921 Samples)

	N Total	N MZ Twins	N DZ Twins	N XZ Twins	N Family Members
E-Risk (NCBI-GSE105018)	1658	860	608	190	-
AMDTSS (NCBI-GSE100227)	479	132	132	-	215
BSGS (NCBI-GSE56105)	614	135	223	-	256
Denmark (NCBI-GSE73115)	180	94	86	-	-
E-MTAB (E-MTAB-1866648)	648	240	408	-	-

Table 2: The datasets that will be used in this research

Findings - Summarized Results

- Classifier 13 & 14 have produced similar results apart from the proportion of non-twins correctly predicted.
- Since classifier 14 was trained from the dataset containing family members while classifier 13 wasn't, classifier 14 would therefore have a better performance when predicting non-twins.
- The average AUC is around 0.65 and the proportion of DZ twins correctly predicted is low.

Average values of the five testing datasets									
	ALIC	proportion MZ twins correctly	proportion DZ twins correctly	non-twins correctly	proportion MZ twins incorrectly	proportion DZ twins incorrectly	non-twins incorrectly		
Classifi and 2	AUC	predicted	predicted	predicted	predicted	predicted	predicted		
Classifier 13	0.645	0.762	0.481	0.440	0.238	0.519	0.560		
Classifier 14	0.647	0.665	0.571	0.724	0.335	0.429	0.276		

Findings - Detailed Results

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606	30010-300	339	239	0.751	0.843	0.469	0.444	0.157	0.531	0.556
000	131	220	255	0.796	0.916	0.450	0.302	0.084	0.550	0.698
1658	852	612	0	0.728	0.684	0.683	NaN	0.316	0.317	NaN
614	135	223	256	0.784	0.904	0.404	0.313	0.096	0.596	0.688
180	94	86	0	0.544	0.596	0.570	NaN	0.404	0.430	NaN
479	132	132	215	0.648	0.742	0.545	0.567	0.258	0.455	0.433
648	240	408	0	0.522	0.882	0.201	NaN	0.118	0.799	NaN
					proportion	proportion DZ	proportion	proportion	proportion DZ	proportion
					MZ twins	twins	non-twins	MZ twins	twins	non-twins
					correctly	correctly	correctly	incorrectly	incorrectly	incorrectly
otal N	N MZ twins	N DZ twins	N non-twin	AUC	predicted	predicted	predicted	predicted	predicted	predicted
934	520	334	80	0.766	0.808	0.572	0.625	0.192	0.428	0.375
606	131	220	255	0.799	0.901	0.468	0.451	0.099	0.532	0.549
1658	852	612	0	0.739	0.621	0.740	NaN	0.379	0.260	NaN
614	135	223	256	0.774	0.637	0.798	0.848	0.363	0.202	0.152
180	94	86	0	0.563	0.755	0.395	NaN	0.245	0.605	NaN
479	132	132	215	0.667	0.750	0.530	0.600	0.250	0.470	0.400
648	240	408	0	0.492	0.564	0.390	NaN	0.436	0.610	NaN
ot:	614 180 479 648 al 934 606 1658 614 180 479	al N MZ twins 934 520 606 131 1658 852 614 135 180 94 479 132	al NMZ twins NDZ twins 934 520 1658 852 614 135 223 180 94 86 479 132 132 132 132 132 1334 134 135 223 136 94 86 137 132	614 135 223 256 180 94 86 0 479 132 132 215 648 240 408 0 934 520 334 80 606 131 220 255 1658 852 612 0 614 135 223 256 180 94 86 0 479 132 132 215	614 135 223 256 0.784 180 94 86 0 0.544 479 132 132 215 0.648 648 240 408 0 0.522 Aux Visions N non-twin AUC 934 520 334 80 0.766 606 131 220 255 0.799 1658 852 612 0 0.739 614 135 223 256 0.774 180 94 86 0 0.563 479 132 132 215 0.667	614 135 223 256 0.784 0.904 180 94 86 0 0.544 0.596 479 132 132 215 0.648 0.742 648 240 408 0 0.522 0.882 al N MZ twins N non-twin AUC proportion MZ twins correctly predicted 934 520 334 80 0.766 0.808 606 131 220 255 0.799 0.901 1658 852 612 0 0.739 0.621 614 135 223 256 0.774 0.637 180 94 86 0 0.563 0.755 479 132 132 215 0.667 0.750	614 135 223 256 0.784 0.904 0.404 180 94 86 0 0.544 0.596 0.570 479 132 132 215 0.648 0.742 0.545 648 240 408 0 0.522 0.882 0.201 al N MZ twins N DZ twins N non-twin AUC proportion MZ twins correctly predicted predicted 934 520 334 80 0.766 0.808 0.572 606 131 220 255 0.799 0.901 0.468 1658 852 612 0 0.739 0.621 0.740 614 135 223 256 0.774 0.637 0.798 180 94 86 0 0.563 0.755 0.395 479 132 132 215 0.667 0.750 0.530	614 135 223 256 0.784 0.904 0.404 0.313 180 94 86 0 0.544 0.596 0.570 NaN 479 132 132 215 0.648 0.742 0.545 0.567 648 240 408 0 0.522 0.882 0.201 NaN al N MZ twins N DZ twins N non-twin AUC proportion MZ twins correctly twins correctly predicted 0.625 0.625 0.625 0.625 0.625 0.625 0.625 0.625 0.621 0.740 NaN 0.848 0.848 0.848 0.848 0.667 0.755 0.3	614 135 223 256 0.784 0.904 0.404 0.313 0.096 180 94 86 0 0.544 0.596 0.570 NaN 0.404 479 132 132 215 0.648 0.742 0.545 0.567 0.258 648 240 408 0 0.522 0.882 0.201 NaN 0.118 al N MZ twins N DZ twins N non-twin AUC proportion MZ twins correctly wins correctly predicted predicted predicted predicted predicted predicted predicted predicted predicted predicted predicted predicted predicted predicted predicted predicted predicted predicted predicted predicted predicted predicted predicted 0.625 0.192 0.625 0.192 0.625 0.192 0.625 0.799 0.621 0.740 NaN 0.379 0.621 0.740 NaN 0.379 0.621 0.740 NaN	614 135 223 256 0.784 0.904 0.404 0.313 0.096 0.596 180 94 86 0 0.544 0.596 0.570 NaN 0.404 0.430 479 132 132 215 0.648 0.742 0.545 0.567 0.258 0.455 648 240 408 0 0.522 0.882 0.201 NaN 0.118 0.799

proportion

MZ twins

correctly

proportion DZ proportion

non-twins

correctly

twins

correctly

proportion

MZ twins

incorrectly

proportion DZ proportion

non-twins

incorrectly

twins

incorrectly

Data Science Pipelines

- Data engineering
 - Dimension reduction from 450k variables
- Machine learning
 - SVM, Adaboost, Random forests, etc.
- Information output
 - Result tables & Figures

Plan for Coming Semester

- Dimension reduction methods
 - PCA, Auto encoder, etc.
- Training classifiers using different machine learning methods
 - Support vector machine (SVM)
 - Logistic regression
 - Bayesian algorithm
 - Random forests
 - Adaptive boosting (Adaboost)
 - Stochastic gradient boosting
 - Deep learning
 - Stacking and Embedding Models

Key Challenges

- The size of the dataset
 - Find a new method to load the dataset in local machine
 - Cloud services
 - Spartan
- 2. High dimensional dataset
 - Do some research and testing different machine learning algorithms
- 3. Unbalanced training data
 - Re-structure the training data
 - Over-sampling for minority class
 - Random Forest

Key tasks, milestones and roles



Thanks for watching!