LipiDisease: associate lipids to diseases using literature mining

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Supplementary Information

Computational methods

The data retrieved from PubChem and PubMed was filtered to limit PubMed records corresponding to at least one disease and one lipid. Among these, the significant associations of diseases and lipids were identified by performing one-tailed Fisher's exact test as described before (Fontaine and Andrade-Navarro, 2016). In short, articles associated with diseases as well as articles associated with lipids were represented in a 2-by-2 contingency matrix to identify the over-representation of the articles associated with both diseases and lipids. All computations were limited to the resulting 709,038 PubMed records associated with 4,488 diseases and 2,771 lipids.

To evaluate the enrichment of a set of lipids to a particular disease, one-tailed Fisher's exact test was performed by creating the contingency matrix depicted in Table S1.

Table S1. The contingency matrix used for performing a one-tailed Fisher's exact test. *S* represents a lipid set under investigation and *D* represents a disease. The number of lipids in the groups a, b, c, and d were used to identify the over-representation.

Lipids	From the <i>S</i>	Not from the <i>S</i>		
Associated with the D	а	b		
Not associated with the D	С	С		

The resulting p-value was corrected for multiple testing by calculating the false discovery rate (FDR) by Benjamini and Hochberg method (Benjamini and Hochberg, 1995) using the R statistical environment (R Core Team, 2020).

Lipid-level statistics were also considered for disease enrichment. A ranked list of lipids was used for disease enrichment using the fast gene-set enrichment algorithm. This was implemented using the fgsea package in R (Korotkevich *et al.*, 2019).

Database contents

The current version of LipiDisease consists of a total of 709,038 PubMed records associated with 4,488 diseases and 2,771 lipids. Sterol lipids, polyketides, and fatty acyls are the most represented categories by those articles (Fig. S1). Among diseases, 23% of the PubMed articles discuss neoplasms with lipids. Figure S1 shows the top 10 diseases associated with any lipids as described in the literature.



Fig. S1. Percentage of PubMed articles annotated with (A) different lipid classes and (B) diseases.

Comparison with LipidPedia

LipidPedia is one of the few tools specialized in associating lipids with biomedical information (Kuo and Tseng, 2018). It performs full-text mining to extract lipid-relevant information including the biomedical associations. However, it lacks statistical filtering and only provides diseases associated with individual lipids. Hence, we obtained lipids associated with coronary disease, type II diabetes mellitus, dyslipidemias, heart failure, and hyperlipidemias from LipidPedia and compared with the results of "Diseases to Lipids" analysis from LipiDisease performed at the default settings (minimum 5 citations associated with diseases compared to LipiDisease (Table S2). This could be because LipidPedia performs full-text mining to find these associations, while LipiDisease only considers MesH-curated citations. Furthermore, LipidPedia lacks statistical filtering making LipiDisease results more stringent. The output of our tool LipiDisease facilitates manual lookup of the PubMed articles from which associations are derived (Figure S2). This is not possible with LipidPedia.

Disease	LipiDisease	LipidPedia	Common
Coronary disease	35	70	12
Diabetes Mellitus, type II	60	87	25
Dyslipidemias	13	7	2
Heart failure	21	36	4

26

73

10

Hyperlipidemias

Table S2. Comparison between LipiDisease and LipidPedia. The number of lipids associated with selected diseases obtained from both resources and overlap between them.

Α						В	
Lipid	Lipid_pubchem_cid	Disease	Citations	♦ p-value ♦	FDR 🔶	Effects of disulfiram and pyridoxine on serum cholesterol.	
Cholesterol				0.00	0.00	Major LP, Goyer PP. Cite Ann Intern Med. 1978 Jan;88(1):53-6. doi: 10.7326/0003-4819-88-1-53.	
Pravastatin	<u>54687</u>	Coronary Disease	534	0.00	0.00	PMID: 619758 Share	
Epicholesterol	5283629	Coronary Disease	5466	0.00	0.00	Cholesterol management in patients hospitalized for coronary heart disease.	
PGI2	5282411	Coronary Disease	322	4.32e-73	1.24e-72	2 Boekeloo BO, Becker DM, LeBailly A, Pearson TA. Cite Am J Prev Med. 1988 May-Jun;4(3):128-32.	
TXA2	<u>5280497</u>	Coronary Disease	194	1.19e-68	3.36e-68	PMID: 3395498 Share	
TXB2	<u>5283137</u>	Coronary Disease	195	1.02e-42	2.53e-42	[Impairment of cholesterol-acceptor function of high density lipoproteins in	
6-keto-PGF1alpha	5280888	Coronary Disease	118	2.56e-25	5.57e-25	3 patients with ischemic heart disease]. Cite Parfenova NS, Petrova-Maslakova LG, Kuznetsov AS, Joffe DV, Alksnis EG.	
alpha-Linolenic acid	<u>5280934</u>	Coronary Disease	<u>63</u>	5.68e-20	1.17e-19	Vopr Med Khim. 1988 Mar-Apr;34(2):42-6. Share PMID: 3400190 Russian.	
EPA	446284	Coronary Disease	114	2.06e-16	4.11e-16	Elevated chalasteral fast or fance?	
Linoelaidic acid	5282457	Coronary Disease	74	1.29e-11	2.41e-11	4 Tamura PY.	
Showing 1 to 10 of 35 entr	ries		Pre	vious 1 2 3	4 Next	Cite Hawaii Med J. 1966 Jun;47(6):264-6, 269. PMID: 3403248 No abstract available.	

Fig. S2. Output of the "Diseases to Lipids" analysis from LipiDisease. (A) top 10 lipids associated with Coronary Disease and (B) articles from PubMed associating the lipid Cholesterol with Coronary Disease.

References

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