## Supplementary Table

S2 Table: Characteristics of the included ML predictive models (Adopted and modified from [141])

	Advantages	Disadvantages	Diseases predicted using the model
Random forest	-Ability to handle datasets with	-Ignores the spatial locations of the	Alzheimer's disease [28][89] [90] [91] [92] [94] [98] [99]
(Number of studies:	a large number of predictor	observations so any spatial autocorrelation	[101]
58)	variables	in the data not accounted for by the	Anxiety [93]
		covariates.	Any cardiovascular disorder [29][55] [59] [60] [65]
			Any mental disorder [100]
			Asthma [113] [116]
			Atherosclerosis [64]
			Atrial fibrillation [53] [124]
			Chronic obstructive pulmonary disease [118]
			Chronic kidney disease [87]
			Colorectal cancer [37] [107] [131] [128] [130]
			Diabetes mellitus [40] [69] [74] [79]
			Diabetic polyneuropathy [78] [80]
			Epilepsy [111]
			Familial hypercholesterolaemia [31] [39] [82]
			Fibromyalgia [104]
			Gestational diabetes [81]
			Heart Failure [49] [57] [58]
			Hypertension [51] [62]
			Inflammatory bowel disease [67]
			Major depressive disorder [41]
			Myocardial infarction [43]
			Obesity [73]
			Pancreatic cancer [108]
			Post stroke spasticity [110]
			Post traumatic stress disorder [96]
			Preeclampsia [36]
			Primary Aldosteronism[70]
			Progressive supranuclear palsy [112]
			Rheumatoid arthritis [35]
			Stroke [48]
			Suicidality [85]
Support Vector	-Ability to provide non-linear	-Require knowledge about the kernel	Alzheimer's disease [30] [89] [94] [98] [101]
Machine (SVM)	solutions	employed to be able to achieve good	Any cardiovascular disorder [52] [55]
		performance	Any mental disorder [100]

(Number of studies:			Atrial fibrillation [53] [124]
30)			Chronic obstructive pulmonary disease [132]
,			Diabetes mellitus [72] [79]
			Diabetic foot [45]
			Diabetic polyneuropathy [77] [78] [80]
			Epilepsy [111]
			Heart failure [50] [57] [58] [66]
			Hypertension [51] [56]
			Influenza [46]
			Major depressive disorder [41]
			Post-partum depression [33]
			Post-traumatic stress disorder [96] [103]
			Preeclampsia [36]
Boosting models	-Ability to handle categorical	-Interpretability of ensemble can be	Alzheimer's disease [98]
(Number of studies:	features	questioned	Anxiety [93]
28)	-Few parameters to tune		Any cardiovascular disorder [55] [65] [60]
,	-Good performance in datasets		Any mental disorder [100]
	with large number of features		Asthma [115] [116]
	-		Atherosclerosis [64]
			Chronic obstructive pulmonary disease [118] [120]
			Colorectal cancer [37] [131] [128]
			Diabetes mellitus [74] [40]
			Diabetic polyneuropathy [80]
			Familial hypercholesterolemia [31]
			Heart failure [57] [66]
			Hypertension [51] [56]
			Lung cancer [109]
			Non-tuberculous mycobacterial lung disease [114]
			Obesity [73]
			Pancreatic cancer [108]
			Post-partum depression [33]
			Preeclampsia [36]
Decision tree	-Ability to provide non-linear	-Interpretability of ensemble can be	Alzheimer's disease [98] [101]
techniques	solutions	questioned	Ankylosing spondylitis [42] [106]
(Number of studies:			Any cardiovascular disorder [55]
25)			Any mental disorder [100]
			Chronic kidney disease [86]
			Colorectal cancer [107]
			COVID-19 [117]

			Diabetes mellitus [40] [68] [75]
			Diabetic polyneuropathy [77] [78] [80]
			Heart failure [57]
			Hypertension[51] [56]
			Lung cancer [109]
			Obesity [71]
			Post-partum depression [33]
			Preeclampsia [36]
			Rheumatoid arthritis [35] [42]
			Stroke [123]
Naïve Baves	-Good performance in small	-Assumption of independence between	Alzheimer's disease [89] [91] [94] [98]
(Number of studies:	datasets if conditional	features	Any cardiovascular disease [63]
8)	independent assumption holds		Enilensy [111]
0)	independent ussumption notus		Major depressive disorder [41]
			Obesity [71]
k-nearest neighbors	-Intuitive algorithm	-Number of neighbors must be defined by	Alzheimer's disease [98]
(Number of studies:	Intuitive argorithm	liser	Any cardiovascular disease [52] [55]
(i tumber of studies. 8)		-High relative computational complexity	Heart failure [50] [58]
0)		Thigh tenutive computational complexity	Hypertension [51]
			Diabetes mellitus [32] [72]
Least absolute	-Good performance with small	-Data assumptions are needed to be	Anxiety [93]
shrinkage and	datasets	complied	Any cardiovascular disease [29] [60]
selection operator	unioris	-Can only provide linear solutions	Atrial fibrillation [53] [124]
(LASSO)		Can only provide inical solutions	Back pain [105]
(Number of studies:			Diabetes mellitus [69] [75]
(14uniber of studies.			Diabetic nephropathy [76]
15)			Major depressive disorder [95]
			Obesity [73]
			Pancreatic cancer [108]
			Stroke [123]
			Suicidality [84]
			Systematic lunus erythematous [34]
Artificial Neural	-State-of-the-art results	-Many parameters to fine-tune	Alzheimer's disease [28] [89] [92] [97] [101] [102]
Network (ANN)	-Direct complex image	-I arge number of samples are required to	Any cardiovascular disorder [55] [65]
(Number of studies:	processing	achieve good performance	Any mental disorder [100]
35)	processing	acine ve 500a performance	Asthma [116]
55)			Atrial fibrillation [53] [124]
			Chronic kidney disease [86]
			Chronic kidney disease [86]

Chronic obstructive pulmonary disease [120]
Diabetes mellitus [47]
Diabetic polyneuropathy [78]
Diabetic retinopathy [44] [38] [125][126] [127]
Familial hypercholesterolemia [31]
Gastroesophageal reflux [26]
Heart failure [49] [50] [57]
Hypertension [54] [61]
Influenza [119]
Pancreatic cancer [108]
Post-traumatic stress disorder [96] [103]
Preeclampsia [36]
Psychosis [122]
Suicidality [83]