

The Harvard-Emory ECG Database

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ABSTRACT

The Harvard-Emory ECG Database (HEEDB) is a large collection of 12-lead electrocardiogram (ECG) recordings, developed through a collaboration between Harvard University and Emory University. The database consists of 10,608,417 unique ECG recordings from 1,818,247 patients from Massachusetts General Hospital (MGH) and 1,452,964 recordings from 552,481 patients from Emory University Hospital (EUH) collected in clinical settings as part of routine patient care since the early 1990s. Continuously updated with new data, the recordings consist of 10-second, 12-lead ECGs sampled at 250 and 500 Hz, and stored in WFDB format. Future updates will include demographic information such as age, sex, race, and ethnicity. Additionally, 12SL annotations—encompassing ECG diagnoses, morphology, and rhythms—are available for 10,471,531 recordings from MGH and 1,268,277 recordings from EUH. Shortly, ICD-9/10 codes, CPT codes, and medication history will also be made publicly available.

Background & Summary

The Harvard-Emory ECG database (HEEDB) is a large collection of 12-lead (I, II, III, aVR, aVL, aVF, V1, V2, V3, V4, V5, V6) electrocardiography (ECG) recordings, prepared through a collaboration between Harvard University and Emory University investigators. HEEDB comprises clinical ECGs obtained during routine care over several decades (from the 1990s to present), alongside associated covariates, including patient demographics, medications, diagnoses, and clinical outcomes which will be added in the future. These data were collected at Massachusetts General Hospital (MGH, Boston, MA, US) and Emory University Hospital (EUH, Atlanta, GA, US) and represent one of the largest ECG datasets available. Large-scale datasets with extended follow-up, such as this one, are uncommon, making this resource particularly valuable for conducting population-wide analyses of cardiac rhythms, their disturbances, and associated outcomes.

The HEEDB supports numerous applications, ranging from identifying early markers of cardiovascular disease to validating detection and prediction algorithms for clinical use. A major focus is the identification of ECG-derived biomarkers that may predict arrhythmias, sudden cardiac death, and other critical conditions. The inclusion of temporal data, with multiple ECG recordings for individual subjects over time, provides an opportunity to study the progression of cardiac abnormalities and their clinical implications. The scale and longitudinal nature of this database offer significant potential to advance both clinical research and computational cardiology, enabling new insights into cardiovascular disease.

Methods

The HEEDB contains 12-lead ECG recordings of 10 seconds duration sampled at 250 and 500 Hz. At the time of initial publication, the database includes 10,608,417 unique ECG recordings (10,771,552 ECGs total) from 1,818,247 patients from MGH. New ECGs will be added to the database periodically, including 1,452,964 recordings from 552,481 subjects recorded at EUH. All ECGs were collected in the course of routine clinical care using the MUSE ECG system. The MUSE system facilitates the acquisition, storage, and review of ECG data, providing consistent recordings across the dataset. The raw ECG waveforms, along with associated metadata, were stored in the XML format, which allows for both structured data

39 representation and interoperability with other systems. ECG waveforms were further de-identified following the Safe Harbor
40 method and converted to the WFDB (Waveform Database)^{1,2} and MATLAB (V4) compatible format.

41 Data Records

42 Each ECG recording includes one waveform data file (.mat) and one header file (.hea). The waveform data file can be read by
43 WFDB library functions, applications, and Toolbox, or be loaded to MATLAB directly. Most waveform files are synchronized
44 12-lead ECG signals recorded at 250 Hz (59% from MGH) and 500 Hz (41% from MGH, 100% from EUH) for a duration of
45 10 s with analog-to-digital converter (ADC) gain 1,000, stored in mV with byte offset of 24. The header file specifies the names
46 of the associated waveform files and their attributes such as sampling rate and units, including the channel names of the signal.
47 It contains line-oriented and field-oriented ASCII text and can be read by the WFDB library or generic text editors. The data
48 file contains 12-lead information encoded in 16 bits. The ECG database exhibits several common abnormalities that occur in
49 real ECG recordings, such as missing leads, incomplete time lengths, and various types of noise affecting the ECG recordings.

50 Metadata

51 The dataset includes general metadata for each subject, including a de-identified unique subject identifier, sex, age, race,
52 ethnicity, and a shifted acquisition date. All the dates are shifted between ± 90 days from the date of acquisition and the
53 shift is different for every subject/recording. Demographic characteristics for both data sources are summarized in Table 1.
54 To ensure confidentiality, individuals aged over 89 years at the time of data acquisition are uniformly recorded as 90 years
55 old. It is important to note that age data is missing for 15,466 recordings from MGH and 37,481 recordings from EUH due to
56 discrepancies in documenting either the acquisition date or birth date. The age distribution for both data sources is displayed in
57 the histogram in Figure 1.

58 ECG Annotations

59 The annotations in the dataset are generated using the Marquette 12SL ECG Analysis Program (GE Healthcare) version 4³.
60 These annotations consist of textual reports that describe the morphology, rhythm, and diagnostic information of the ECGs.
61 Each annotation includes statement numbers that correspond to human-readable diagnoses. The frequency of statements with
62 the highest occurrence is summarized in Table 2 for MGH and Table 3 for EUH.

63 The statement numbers include various components such as ECG diagnoses, morphology, rhythms, lead names, text
64 symbols (e.g., brackets, delimiters), measurements (e.g., “QTcB \geq 480 ms”), conjunctions (e.g., “with,” “or,” “and”), and
65 descriptive words like “present,” “accelerated,” and “blocked.” Together, these elements comprise the 12SL language model to
66 form human-readable diagnosis reports. For example, the following statement consists of the codes: 19, 177, 231, 178, 244,
67 390, 411, 700, 831, 1100, 1150, and 169, which map to the diagnosis: Sinus rhythm, with, premature ventricular complexes, or,
68 fusion complexes, Indeterminate axis, Pulmonary disease pattern, Septal infarct, , age undetermined, ST & T wave abnormality,
69 consider anterior ischemia, Abnormal ECG. Annotations are available for 10,471,531 ECG recordings, leaving 136,886
70 unlabeled recordings from MGH. For the EUH dataset, 1,268,277 recordings are annotated, with 184,687 remaining unlabeled.

71 Technical Validation

72 The ECG records were obtained in a clinical environment, during routine care guided by specialized personnel using the MUSE
73 ECG system facilitating acquisition, storage, and review of ECG data, enabling consistent recordings across the dataset. The
74 HEEDB data was checked for completeness and includes the non-preprocessed raw ECG waveforms in WFDB format coupled
75 with demographics and 12SL statements. Data was de-identified following the Safe Harbor method. ECG recordings contain
76 various types of standard ECG artifacts such as missing leads, missing parts of signals, or noise.

77 Usage Notes

78 The HEEDB dataset can be downloaded from the Brain Data Science Platform (BDSP) by following the instructions provided
79 at Harvard-Emory ECG database website^{4,5}(<https://bdsp.io/content/heedb/2.0/>). The data are made available
80 under an Attribution -NonCommercial - NoDerivatives (CC BY-NC-ND) license. This license enables reusers to copy and
81 distribute the material in any medium or format in **unadapted** form only, for **noncommercial purposes** only, and only so long
82 as **attribution is given to the creator**.

83 Code availability

84 The downloaded waveform data files can be accessed using WFDB library functions, applications, and toolboxes in Python⁶
85 and MATLAB⁷ or can be directly loaded into MATLAB. No custom code was developed for this study.

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102 **Author contributions statement**

103 Z.K. provided technical validation of the database, statistical analysis of the database, processing of the annotations and drafted
104 the manuscript; Q.L. contributed to statistical analysis of the data; C.R. contributed to data collection; J.M.V contributed to data
105 collection, developed infrastructure for data storage; M.G. contributed to data collection; A.G. conducted preliminary analyses
106 of the data; J.R. conducted preliminary analyses of the data; S.H. conducted preliminary analyses of the data; A.A. contributed
107 to data collection; D.E.A. provided insights into the clinical relevance of the data; J.X secured data annotations; A.P. secured
108 data annotations; R.S. contributed to the design and implementation of the data collection processes as well as overseeing the
109 technical validation of the database and contributed to the writing; M.A.R. contributed to the design and implementation of the
110 data collection processes, contributed to the writing and led; M.B.W. secured the funding, designed the database, contributed to
111 the design and implementation of the data collection processes, and contributed to the writing; G.D.C. secured the funding,
112 designed the database, provided intellectual input on processing and labeling, contributed to the manuscript and led the project.
113 All authors read the manuscript.

114 **Competing interests**

115 M.B.W. is a co-founder, scientific advisor, and consultant to Beacon Biosignals and has a personal equity interest in the
116 company. D.E.A. is the founder and Chief Medical Officer of AliveCor Inc. and holds personal equity in the company. J.X. is a
117 research fellow at Alivecor Inc. and also holds personal equity in the company. G.D.C. holds significant stock in AliveCor Inc.
118 The other authors report no competing interests.

119 **Figures & Tables**

HEEDB	MGH		Emory	
	Subjects Total (n=1,818,247)	Recordings Total (n=10,608,417)	Subjects Total (n=552,481)	Recordings Total (n=1,452,964)
Age				
Age, years (mean (std))	-	60.61 (18.3)	-	61.4 (16.1)
Unavailable	-	15,466 (0.2%)	-	37,481 (2.6%)
- [0 – 2)	-	33,907 (0.3%)	-	500 (0.03%)
- [2 – 10)	-	64,837 (0.6%)	-	215 (0.01%)
- [10 – 18)	-	114,391 (1.1%)	-	3,900 (0.3%)
- [18 – 40)	-	1,309,892 (12.4%)	-	160,398 (11.0%)
- [40 – 60)	-	3,027,640 (28.5%)	-	446,091 (30.7%)
- [60 – 89)	-	5,736,417 (54.1%)	-	776,837 (53.5%)
- >89	-	305,867 (2.9%)	-	27,542 (1.9%)
Sex				
Female	925,510 (51.1%)	-	284,979 (51.9%)	-
Male	886,155 (48.7%)	-	266,955 (48.3%)	-
Unavailable	3,582 (0.2%)	-	547 (0.1%)	-
Race				
White	1,291,671 (71.0%)	-	287,707 (52.1%)	-
Unavailable	347,943 (19.1%)	-	65,774 (11.9%)	-
Black or African American	120,207 (6.6%)	-	183,456 (33.2%)	-
Asian	49,748 (2.7%)	-	12,366 (2.2%)	-
Other	8,698 (0.4%)	-	3,178 (0.6%)	-
Ethnicity				
Non-Hispanic	844,564 (46.5%)	-	-	-
Unavailable	848,415 (46.7%)	-	-	-
Hispanic	125,268 (6.9%)	-	-	-
Education				
Unavailable	861,933 (47.4%)	-	-	-
Did not complete high school	97,953 (5.4%)	-	-	-
High school or equivalent	334,142 (18.4%)	-	-	-
Some college	82,648 (4.6%)	-	-	-
College and/or advanced degree	504,120 (27.7%)	-	-	-

Table 1. Available demographic information for the ECG records in the HEEDB dataset, including age, gender, race, ethnicity, and education level.

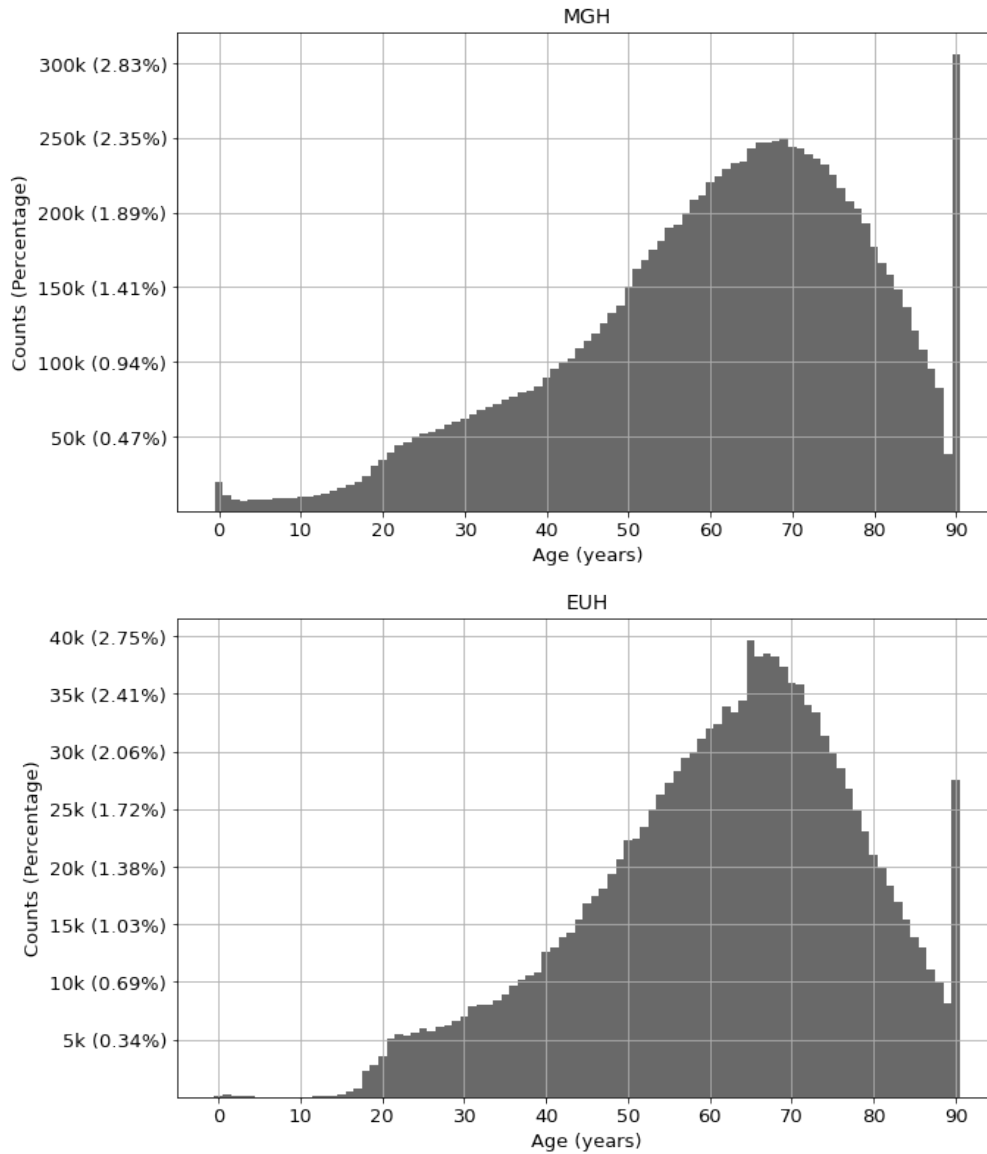


Figure 1. Histogram depicting the age distribution within the HEEDB dataset for Massachusetts General Hospital (MGH) recordings (top figure) and Emory University Hospital (EUH) recordings (bottom figure).

12SL Code	Statement	Count
1699	Abnormal ECG	7,361,517
22	Normal sinus rhythm	4,115,214
831	, age undetermined	2,170,034
1140	Nonspecific T wave abnormality	2,00,4701
19	Sinus rhythm	1,796,277
177	With	1,685,402
1684	Normal ECG	1,619,847
410	Low voltage QRS	1,330,445
244	Fusion complexes	1,288,987
390	Indeterminate axis	1,250,197
21	Sinus bradycardia	1,138,840
231	Premature ventricular complexes	1,130,305
1665	(1,129,759
1666)	1,129,759
411	Pulmonary disease pattern	927,790
23	Sinus tachycardia	902,408
372	Left axis deviation	851,802
1687	Otherwise normal ECG	826,410
780	Inferior infarct	798,145
161	Atrial fibrillation	767,742
299	Undetermined rhythm	732,518
542	Minimal voltage criteria for LVH, may be normal variant	705,711
1680	Possible	679,705
1141	Nonspecific ST and T wave abnormality	668,763
533	Cornell product	616,603
178	Or	601,429
530	R in aVL	591,281
101	With 1st degree AV block	575,095
1693	Borderline ECG	561,088
1100	ST	535,730
440	Right bundle branch block	514,185
740	Anterior infarct	474,416
1682	Cannot rule out	462,174
211	With occasional	432,974
380	Rightward axis	405,490
700	Septal infarct	393,446
1683	,	392,131
267	Junctional rhythm	387,313
1160	T wave abnormality, consider lateral ischemia	356,454
1146	QTcB \geq 480 ms	355,159
1673	*** Poor data quality, interpretation may be adversely affected	350,927
251	With sinus arrhythmia	333,597
222	Premature atrial complexes	324,974
541	Left ventricular hypertrophy	321,334
482	Nonspecific intraventricular block	319,421
33	Accelerated	277,351
445	Incomplete right bundle branch block	266,564
360	Left atrial enlargement	262,051
171	With rapid ventricular response	252,983
1143	Prolonged QT	250,326

Table 2. This table lists the 50 statements with the highest prevalence in the HEEDB for MGH data, including their corresponding 12SL codes, human-readable descriptions, and counts. The statements encompass various elements, including diagnoses, rhythms, conjunctions, ECG lead names, and brackets and other symbols, all of which contribute to the formation of a human-readable statement generated by the 12SL software.

12SL Code	Statement	Count
1699	Abnormal ECG	1,297,906
831	, age undetermined	1,143,262
22	Normal sinus rhythm	682,788
1680	Possible	651,936
700	Septal infarct	580,815
760	Lateral infarct	425,489
820	Anterolateral infarct	230,557
520	Right ventricular hypertrophy	200,551
445	Incomplete right bundle branch block	191,208
21	Sinus bradycardia	189,975
1682	Cannot rule out	188,883
410	Low voltage QRS	174,791
19	Sinus rhythm	170,730
372	Left axis deviation	147,536
740	Anterior infarct	127,294
780	Inferior infarct	120,966
810	Anteroseptal infarct	120,699
23	Sinus tachycardia	108,889
1666)	105,221
1665	(105,221
530	R in aVL	97,627
101	With 1st degree AV block	91,620
161	Atrial fibrillation	88,573
231	Premature ventricular complexes	78,863
542	Minimal voltage criteria for LVH, may be normal variant	73,897
177	With	72,794
482	Nonspecific intraventricular block	65,658
440	Right bundle branch block	64,964
1160	T wave abnormality, consider lateral ischemia	62,600
1100	ST	57,750
211	With occasional	56,935
251	With sinus arrhythmia	54,769
1170	T wave abnormality, consider inferior ischemia	50,993
222	Premature atrial complexes	50,500
1340	*** Critical test result:	41,691
1143	Prolonged QT	40,725
1140	Nonspecific T wave abnormality	36,538
470	Left anterior fascicular block	35,114
1684	Normal ECG	31,679
1150	T wave abnormality, consider anterior ischemia	29,813
541	Left ventricular hypertrophy	29,242
171	With rapid ventricular response	28,167
212	With frequent	23,383
181	With premature ventricular or aberrantly conducted complexes	21,063
1683	,	20,472
544	With repolarization abnormality	19,683
221	Premature supraventricular complexes	18,448
162	Atrial flutter	18,178

Table 3. This table lists the 50 statements with the highest prevalence in the HEEDB for Emory data, including their corresponding 12SL codes, human-readable descriptions and counts.