

Cerebromicrovascular Disease in Elderly with Diabetes

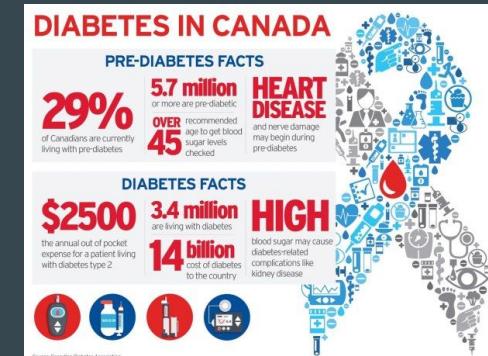
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Background

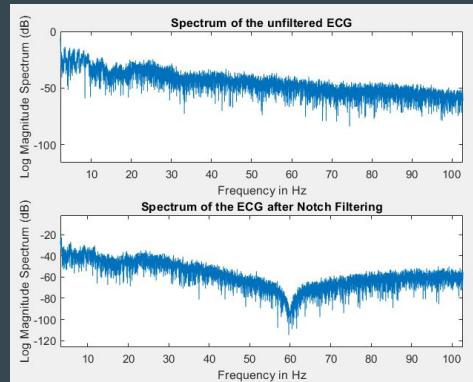
- Diabetes Mellitus (DM), prevalent in around 200 000 Canadians
- Cerebromicrovascular disease: small vessel disease
- DM alters blood barriers affecting microcirculatory regulation
- Dataset: 120 participants aged 55 to 75, with 60 diabetics and 60 control
- Analysis: 1 minute of 15 control signals and 18 diabetic signals
- In open literature, ML used to detect diabetes and pre-diabetes using ECG signals [1]
 - DiaBeats algorithm using 1262 subjects
 - 97.1% precision, 96.2% recall, and 96.8% accuracy



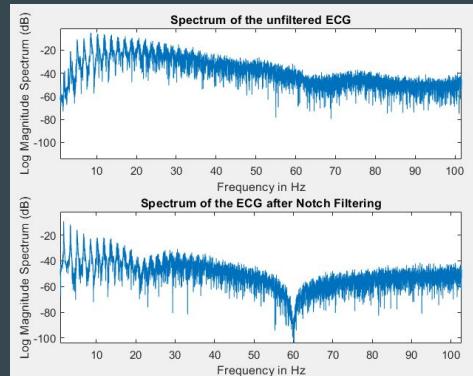
Pre-processing Methods

Method 1: Notch Filtering

- Used to Filter 60 Hz power-line interference (PLI)
- Occurs at 50 - 60 Hz
- Notch filter removes noise at 60 Hz point by adding a zero



Subject 4, Control



Subject 20, Diabetic

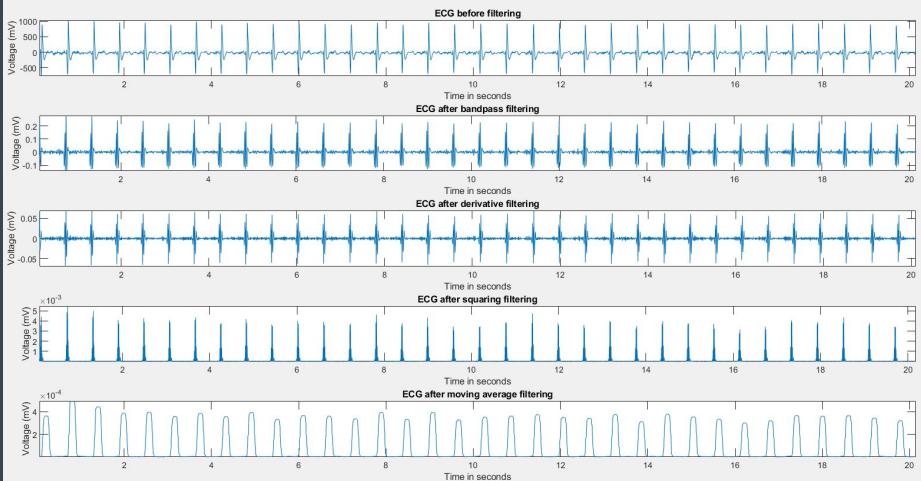
Pre-processing Methods

Method 2: Pan-Tompkins Algorithm

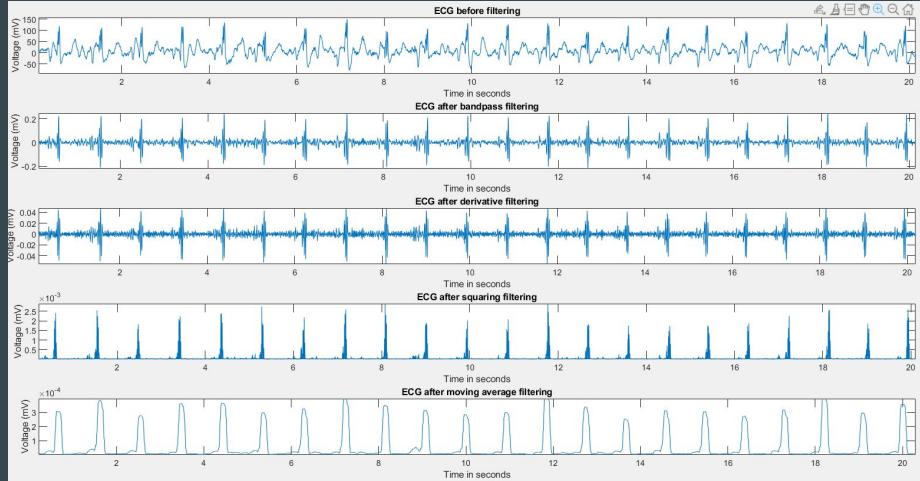
- Low-pass and high-pass filter cascaded for bandpass
- Differentiator to detect high vs. low peaks
- Squaring operator to amplify peaks and positive values
- Moving window integrator for definitive peaks

Pre-processing Methods

Method 2: Pan-Tompkins Algorithm



Subject 4, Control



Subject 20, Diabetic

Feature Analysis

Only the 6 important features were sent to the classification. These features are most affected in diabetic patients in ECG.

- 1) heart rate / mean RR
- 2) std from R-R
- 3) mean interval from S-T
- 4) std interval from S-T
- 5) mean interval Q-T
- 6) std interval Q-T

Method 1: Find the PQRST points (ECG toolbox by Rohan Sanghavi)

R Detection:

The moving window average plot with peak detection

S Detection:

The S point is located right after the R peak, minimum point is located between the R point and $0.1 * \text{sampling frequency}$.

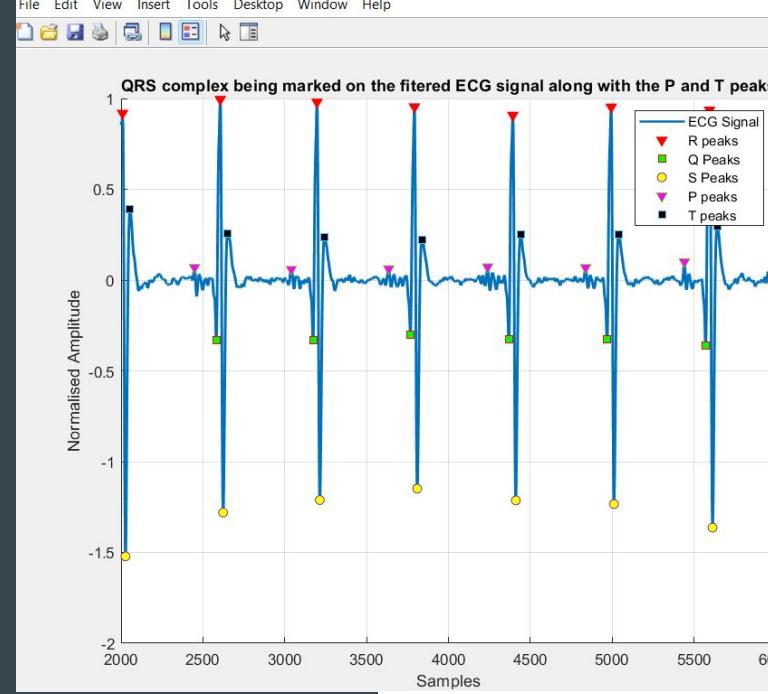
Q Detection:

The Q point is before the R peak, minimum point between the R point and $0.08 * \text{sampling frequency}$.

P and T Detection:

Searching half of the R-R distance before the R peak but after the S peak.

From here we are able to determine: 1) heart rate / mean RR 2) std from R-R



Heart Rate:

$$\text{HR (bpm)} = \frac{60}{\text{RR interval}} * 1000 \quad (9)$$

Mean:

$$\bar{X} = \frac{\sum X}{N} \quad (10)$$

Standard deviation:

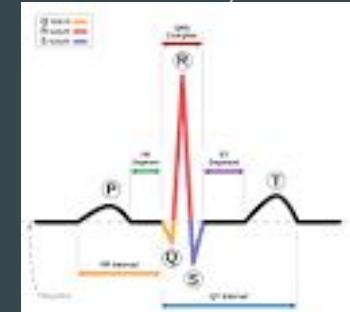
$$\text{SD} = \sqrt{\frac{\sum |x - \bar{x}|^2}{n}} \quad (11)$$

Method 2: Finding the PQRST intervals (ECG toolbox by Rohan Sanghavi)

- 1) Finds the closest R points and pairs them up with the closest P, Q, S, and T points.
- 2) Remove any points need that are not close to R peaks, max period length/2.
- 3) The intervals were calculated by finding the difference between the points.

From here we determine: mean interval from S-T , std interval from S-T, mean interval Q-T, and std interval Q-T

Limitations: Our findings estimated the separation between the peaks. P-T intervals, S-T intervals take into account the ends of curve of P and T.



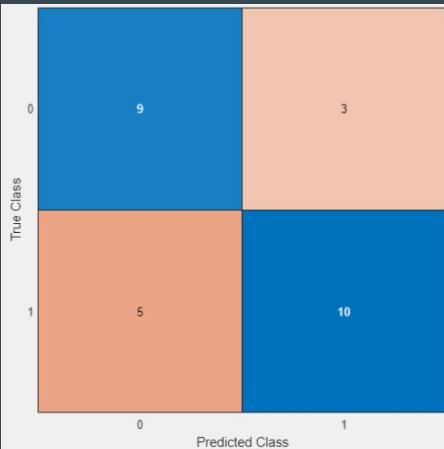
Machine Learning

5 Fold cross validation

3 Machine Learning Algorithms were trained and tested:

- Decision Tree: Branching nodal classification, 70.4%
- Linear Discriminant Analysis: Dimensional reduction linear combination, 63.0%
- Logistic Regression: Logit estimation linear combination, 77.8%

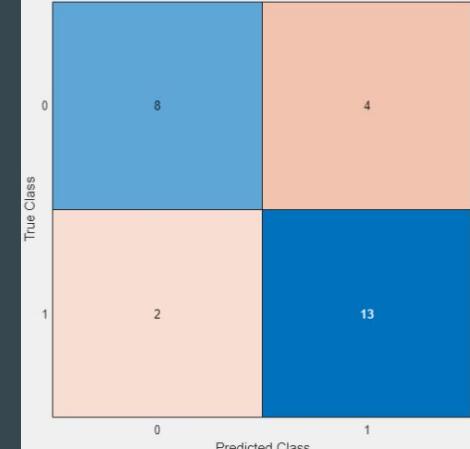
Decision Tree



Linear Discriminant

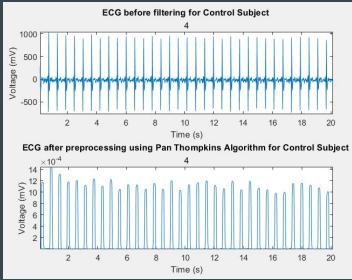


Logistic Regression

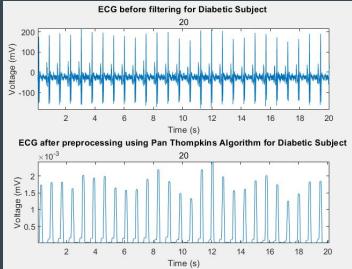


Results

Preprocessing

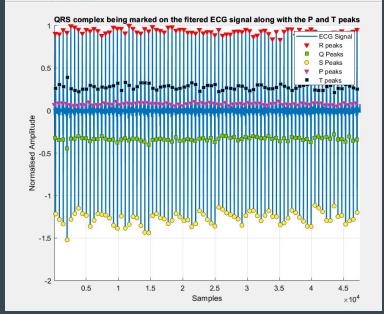


Subject 4, Control



Subject 20, Diabetic

Feature Analysis



Subject 4, Control

Machine Learning

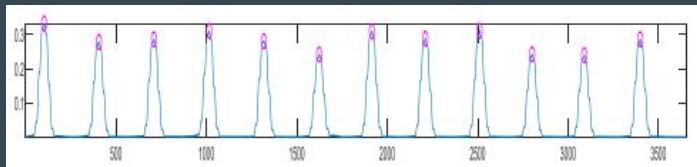
| Model | Validation % | Testing % |
|---------------------|--------------|-----------|
| Decision Tree | 70.4 | 66.7 |
| Linear Discriminant | 63.0 | 50 |
| Logistic Regression | 77.8 | 66.7 |

| A | B | C | D | E | F | G | H | | |
|----|----|----|---------|---------|---------|---------|---------|----------|---------|
| 1 | i | HS | meanST | stdST | RRmean | meanCT | RRstd | stdCT | |
| 2 | 1 | 0 | 56.4705 | 62.4222 | 51.9851 | 1025.15 | 130.178 | 35.598 | 32.648 |
| 3 | 2 | 0 | 73.9912 | 135.788 | 79.8089 | 810.897 | 221.682 | 219.24 | 111.141 |
| 4 | 4 | 0 | 81.4612 | 208.538 | 73.6854 | 756.547 | 287.713 | 13.0432 | 75.9643 |
| 5 | 5 | 0 | 87.3597 | 34.913 | 3.81845 | 687.603 | 72.8116 | 17.203 | 10.2646 |
| 6 | 6 | 0 | 106.973 | 59.4167 | 7.93226 | 560.892 | 105.619 | 8.99595 | 8.04341 |
| 7 | 7 | 0 | 87.3002 | 46.9412 | 7.65269 | 687.284 | 111.324 | 12.0677 | 33.0564 |
| 8 | 8 | 0 | 101.991 | 82.9024 | 71.5079 | 588.288 | 121.512 | 61.5647 | 72.9431 |
| 9 | 10 | 0 | 97.5756 | 31.6974 | 4.302 | 614.908 | 540.562 | 11.8526 | 12.0404 |
| 10 | 18 | 0 | 85.4901 | 238.153 | 82.4892 | 701.836 | 342.972 | 85.2491 | 83.7084 |
| 11 | 19 | 0 | 70.441 | 169.038 | 123.779 | 851.741 | 215.07 | 100.285 | 133.632 |
| 12 | 36 | 0 | 99.7995 | 29.2405 | 1.9365 | 601.205 | 70.2522 | 5.73972 | 1.86375 |
| 13 | 38 | 0 | 89.7647 | 63.8451 | 33.158 | 668.414 | 101.254 | 46.6819 | 35.1924 |
| 14 | 31 | 0 | 104.417 | 175.44 | 53.8752 | 574.617 | 314.728 | 36.4268 | 91.4701 |
| 15 | 26 | 0 | 105.749 | 149.824 | 65.6891 | 567.381 | 236.518 | 4.57048 | 64.9878 |
| 16 | 27 | 0 | 75.9638 | 22.541 | 1.50082 | 789.85 | 78.8852 | 25.7589 | 23.2114 |
| 17 | 28 | 1 | 102.771 | 26.3 | 0.89159 | 583.825 | 518.065 | 11.2134 | 0.9717 |
| 18 | 29 | 1 | 108.301 | 80.6353 | 51.765 | 553.976 | 130.459 | 6.52885 | 51.6581 |
| 19 | 30 | 1 | 95.1198 | 156.613 | 77.0845 | 630.784 | 239.613 | 10.9348 | 76.8888 |
| 20 | 32 | 1 | 83.054 | 201.18 | 95.3459 | 721.721 | 153.074 | 10.798 | 95.31 |
| 21 | 33 | 1 | 89.3889 | 201.394 | 95.3459 | 665.792 | 170.945 | 10.9196 | 221.094 |
| 22 | 34 | 1 | 89.3889 | 201.394 | 95.3459 | 665.792 | 170.945 | 10.9196 | 83.2529 |
| 23 | 3 | 1 | 84.8802 | 258.657 | 66.7121 | 707.545 | 364.985 | 34.3928 | 71.7335 |
| 24 | 9 | 1 | 94.052 | 122.48 | 78.9288 | 637.945 | 164.297 | 22.55634 | 79.306 |
| 25 | 11 | 1 | 78.632 | 65.9358 | 8.29353 | 763.048 | 644.254 | 14.4716 | 16.6893 |
| 26 | 12 | 1 | 89.705 | 179.761 | 59.9236 | 668.857 | 260 | 17.2091 | 65.1741 |
| 27 | 14 | 1 | 92.2497 | 43.0417 | 1.73936 | 650.408 | 89.25 | 10.7006 | 1.37123 |
| 28 | 15 | 1 | 80.5997 | 38.381 | 3.46144 | 744.419 | 79.1587 | 23.2517 | 1.09554 |

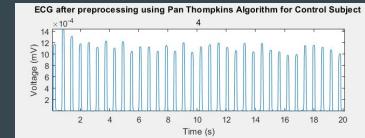
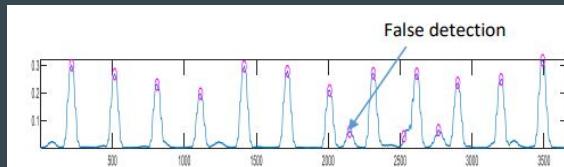
Discussion of Results

Preprocessing:

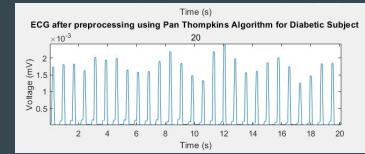
- Notch filter effective but no visible noticeable change
- Control ECG resembled signal of open literature
- Diabetic had false peak detection due to lots of noise



Open Literature



Subject 4, Control



Subject 20, Diabetic

Discussion of Results

Feature Extraction Result:

Heart Rate / RR-mean and standard deviation R-R

Research shows that a normal elder aged 45 to 70 has an average heart rate between 60 to 100 bpm [8].

Patients with diabetes have an increased heart rate reaching more than 86 bpm. [9]

Our similar result

Diabetes(126 bpm - 65 bpm).

Control (106.9 bpm - 58.47 bpm)

| Label | HR (bpm) | meanST (ms) | std_ST (ms) | RR_mean (ms) | mean_QT (ms) | RR_std (ms) | std_QT (ms) |
|-------|------------|-------------|-------------|--------------|--------------|-------------|-------------|
| 0 | 87.2596617 | 34.9130435 | 3.8184503 | 687.6029412 | 72.8115942 | 17.2030366 | 10.2645526 |
| 0 | 106.972548 | 59.4166667 | 7.93225584 | 560.8915663 | 105.6190476 | 8.99595037 | 8.04341347 |
| 1 | 84.812919 | 168 | 82.9954362 | 707.4393939 | 234.5522388 | 4.93068504 | 82.1429433 |
| 1 | 101.795378 | 108.3125 | 77.2240852 | 589.4177215 | 161.3625 | 17.1206934 | 76.3289423 |

2) mean interval from S-T / std interval from S-T

According to Shlomo Stern, patients with diabetes have high changes in S-T intervals [10].

This conclusion is supported by our results where the standard deviation of S-T intervals is way higher than the standard deviation of the S-T intervals on the constant patient.

4) mean interval Q-T/ std interval Q-T

According to Shlomo Stern, patients with diabetes have long Q-T intervals [11].

The results reflect this where mean Q-T is higher in comparison to normal patients.

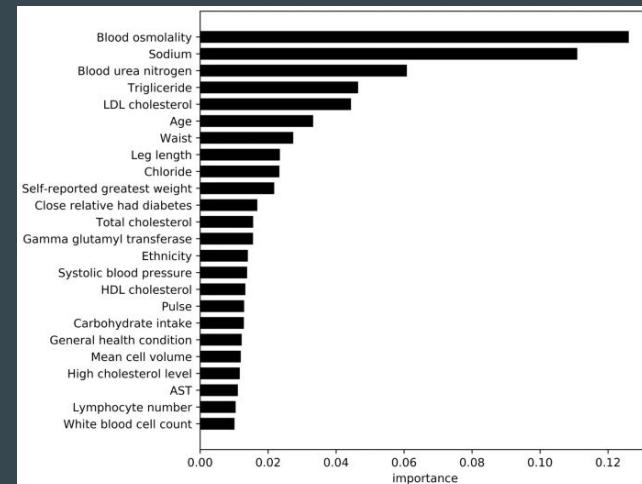
Discussion of results

- XGBoost-based DiaBeats model trained on ECG signal extracted features
 - Diabetes classification prediction accuracy of 96.8%
- Quan Zou decision tree and random forest classifier models

| Dataset | Classifier | ACC |
|---------|------------|--------|
| Luzhou | RF | 0.8084 |
| | J48 | 0.7853 |

Predicting Diabetes
w/ ML (2018, Zou)

- An Dinh weighted feature extraction
 - Logistic regression: 82.7%
 - SVM: 84.9%



Optimal Features (2019, Dinh)

References

- [1] A. R. Kulkarni, A. A. Patel, K. V. Pipal, S. G. Jaiswal, M. T. Jaisinghani, V. Thulkar, L. Gajbhiye, P. Gondane, A. B. Patel, M. Mamtani, and H. Kulkarni, "Machine-learning algorithm to non-invasively detect diabetes and pre-diabetes from Electrocardiogram," *BMJ Innovations*, 06-Jul-2022. [Online]. Available: <https://innovations.bmjjournals.org/content/early/2022/07/06/bmjinnov-2021-000759>. [Accessed: 29-Nov-2022].
- [2] P. H. A. of Canada, "Government of Canada," *Canada.ca*, 17-Aug-2020. [Online]. Available: <https://www.canada.ca/en/public-health/services/publications/diseases-conditions/diabetes-canada-highlights-chronic-disease-surveillance-system.html>. [Accessed: 28-Nov-2022].
- [3] C.-L. J. D. E. Z. A. P. B; "Cerebral Small Vessel Disease: A Review," *Advances in clinical and experimental medicine : official organ Wroclaw Medical University*. [Online]. Available: [https://pubmed.ncbi.nlm.nih.gov/33768739/#:~:text=Cerebral%20small%20vessel%20disease%20\(CSVD,over%2080%20years%20of%20age](https://pubmed.ncbi.nlm.nih.gov/33768739/#:~:text=Cerebral%20small%20vessel%20disease%20(CSVD,over%2080%20years%20of%20age). [Accessed: 28-Nov-2022].
- [4] V. Novak and R. Quispe, "Cerebromicrovascular disease in elderly with diabetes," *Cerebromicrovascular Disease in Elderly with Diabetes v1.0.1*, 05-Aug-2022. [Online]. Available: <https://doi.org/10.13026/00bm-0x81>. [Accessed: 28-Nov-2022].
- [5] M. A. Z. Farhal, R. Ikeura², S. Hayakawa², and S. Tsutsumi², "IOPscience," *Journal of Physics: Conference Series*, 01-Jun-2020. [Online]. Available: <https://iopscience.iop.org/article/10.1088/1742-6596/1532/1/012022#:~:text=The%20Pan%2DTompkins%20Algorithm%20is,quality%20clinical%20ECG%20signal%20data>. [Accessed: 28-Nov-2022].
- [6] <https://www.frontiersin.org/articles/10.3389/fgene.2018.00515/full>
- [7] <https://bmcmedinformdecismak.biomedcentral.com/articles/10.1186/s12911-019-0918-5>
- [8] A. J. Rodriguez, "Normal heart rate for elderly adults," *Griswold Home Care*, 07-Sep-2021. [Online]. Available: <https://www.griswoldhomecare.com/blog/2021/september/normal-heart-rate-for-elderly-adults/#:~:text=A%20normal%20heart%20rate%20for,heart%20rate%20with%20regular%20exercise>. [Accessed: 28-Nov-2022].
- [9] Organization, Heilio, and ImageObject, "Higher heart rates related to type 2 diabetes risk in normal-weight adults," *Heilio*, 18-Feb-2019. [Online]. Available: <https://www.heilio.com/news/endocrinology/20190215/higher-heart-rates-related-to-type-2-diabetes-risk-in-normalweight-adults#:~:text=The%20researchers%20found%20that%20heart,%3D%201.24%3B%2095%25%20CI%2C>. [Accessed: 28-Nov-2022].
- [10] S. Stern, Shlomo Stern From The Hebrew University of Jerusalem, S. Scalarowsky, Samuel Scalarowsky From The Hebrew University of Jerusalem, and C. to S. Stern, "The ECG in diabetes mellitus," *Circulation*, 20-Oct-2009. [Online]. Available: <https://www.ahajournals.org/doi/full/10.1161/circulationaha.109.897496>. [Accessed: 28-Nov-2022].
- [11] M. D. Lawrence Rosenthal, "Normal electrocardiography (ECG) intervals," *Normal Electrocardiography Intervals*, 16-Jul-2021. [Online]. Available: <https://emedicine.medscape.com/article/2172196-overview>. [Accessed: 28-Nov-2022].