# **CV Approval Cover Sheet**

Complete this sheet and place it atop your completed forms before you scan and submit them.

Name:	Aditya Ker	Aditya Kendre						
Email:	kendreadi	tya@gmail.d	com					
Building:	CVHS	Grade:	12					
Sponsor:	Michael Fl	oreck	Email:	: mfloreck@cv	vschools.o	rg		
Required	forms: <i>Thes</i>	se forms m	nust be comp	pleted for A	LL projec	ts.		
Sponso	r initials							
	1: Adult S	ponsor Chec	<b>cklist</b> - Checklist	t and Sponsor S	Signature			
title,			– Student and F terials, procedu	-				search plan (i.e.
	<b>1B: Appro</b> required.	oval form - Si	ignatures of stu	dent and parer	nt/guardian,	as well as	signatures o	of SRC or IRB if
	<b>3: Risk Assessment form</b> - required by the CV IRB in order to determine if the project involves hazardous chemicals (not found in a typical high school chemistry laboratory setting), hazardous activities or devices (i.e. weapons), and microorganism that are not exempt from pre-approval.							
Potential <sup>1</sup>	forms: <i>Dete</i>	ermined w	ith your spo	nsor by usii	ng the <u>Fo</u>	<u>rm Wiza</u>	<u>rd.</u>	
	1C: Regul	ated Researc	ch / Institution	<b>setting</b> - for p	rojects not	completed	at home or	school.
			<b>form -</b> may be r zardous biologic					
		-	s form - ALL pro ed Consent - re	•	•	•	humans.	
	5A: Verte	brate Anima	I - ALL non-exer	mpt vertebrate	projects co	nducted at	home/scho	ool/field site
	5B: Vertel	orate Animal	l - for vertebrate	e projects cond	lucted in a F	Regulated R	esearch Ins	stitution
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		numan and ot						(including prima ood products and
	the same f	ield of study	arch Progressio as a previous pr included with th	roject for this st				

#### **ISEF Sample Abstract & Certification**

Category Generative Adversarial Networks for PCG Arrhythmia Detection Pick one only-mark an "X" in box at right Aditya Kendre **Animal Sciences** Behavioral and Social Cumberland Valley HS, Mechanicsburg, PA, USA Sciences Biochemistry With the rapid growth of computational power and complex algorithms, we Biomedical and Health Sciences propose a novel approach to detect arrhythmias in Phonocardiograms (PCGs). Biomedical Engineering Typically, Electrocardiograms are used to diagnose arrhythmias, requiring Cellular & Molecular Biology medical-grade equipment to accurately recognize cardiac illnesses. PCGs provide Chemistry ease of access to everyone who has a device capable of recording audio, allowing Computational Biology and medical professionals to treat arrhythmias in the developmental stages. The new **Bioinformatics** design comprises two subsystems; one is based on the relationship between Earth & Environmental Sciences Electrocardiograms (ECGs) and PCGs, and the other between PCGs and **Embedded Systems** arrhythmias. The association between ECGs and PCGs is amended to translate Energy: Sustainable from one space to another, where ECGs become dimensionally reduced, then Materials and Design reconstructed into a PCG signal. The second subsystem uses a Generative **Engineering Mechanics** Adversarial Networks (GAN), in which both arbitrary PCG signals are generated, **Environmental Engineering** and preexisting ECG datasets are recreated into PCG signals (using subsystem Materials Science one). These signals are fed into a classifier that detects if an arrhythmia is Mathematics present. This proposed system's advantage is that PCG data is more readily Microbiology available than ECG data; hence, more heart diagnostics can be made. Physics and Astronomy **Plant Sciences** Robotics & Intelligent Machines Systems Software Translational Medical Science 1. As a part of this research project, the student directly handled, manipulated, or interacted with (check all that apply): potentially hazardous biological agents human participants vertebrate animals microorganisms rDNA tissue This abstract describes only procedures performed by me/us, reflects my/our own independent research, and represents one year's work only. I/We worked or used equipment in a regulated research institution or industrial setting. This project is a continuation of previous research. ✓ yes ✓ no My display board includes non-published photographs/visual depictions of humans (other than FOR ISEF myself): OFFICIAL USE ✓ yes ONLY 6. I/We hereby certify that the abstract and responses to the above statements are correct and properly reflect my/our own work.

✓ yes

# **Checklist for Adult Sponsor (1)**

This completed form is required for ALL projects.

To be completed by the Adult Sponsor in collaboration with the student researcher(s):

Student's Name(s): Aditya Kendi	ře					
Project Title: Generative Adversarial Ne						
1. I have reviewed the ISEF Rules a	and Guidelines, including the so	ience fair ethics st	atement.			
2. I have reviewed the student's co	ompleted Student Checklist (1A)	) and Research Plan	n/Project Summary.			
3.  I have worked with the student	and we have discussed the pos	sible risks involved	l in the project.			
4. The project involves one or mor Humans Vertebrate Animals	Poten	prior approval by a ntially Hazardous Bi Microorganisms				
	<b>₽</b> R		ŕ	nent)		
see full text of the rules.)  Human Participants Form (4  Sample of Informed Conse	t includes the use of one or mo signed inventions/prototypes. (F 4) or appropriate Institutional IR nt Form (when applicable and/o (when applicable and/or requir	Requires prior appr B documentation or required by the II	roval by an Institutional Review	Board (IRB);		
Vertebrate Animals (Requires prior approval, see full text of the rules.)  Vertebrate Animal Form (5A)-for projects conducted in a school/home/field research site (SRC prior approval required.)  Vertebrate Animal Form (5B)-for projects conducted at a Regulated Research Institution. (Institutional Animal Care and Use Committee (IACUC) approval required prior experimentation.)  Qualified Scientist Form (2) (Required for all vertebrate animal projects at a regulated research site or when applicable)						
Human and Vertebrate Anir fresh or frozen tissue, prima Qualified Scientist Form (2) The following are exempt fresimilar microorganisms, for	gical Agents Risk Assessment F mal Tissue Form (6B)-to be com ary cell cultures, blood, blood p	orm (6A) upleted in addition roducts and body isk Assessment For uposting, fuel produ	to Form 6A when project involfluids.  rm 3: projects involving protistuction or other non-culturing e	ves the use of s, archae and xperiments,		
Hazardous Chemicals, Activities and Devices (No SRC prior approval required, see full text of the rules.) Risk Assessment Form (3) Qualified Scientist Form (2) (required for projects involving DEA-controlled substances or when applicable)						
Other Risk Assessment Form (3)						
I attest to the information checked above and that I have read and agree to abide by the science fair ethics statement.						
Mike Floreck			10/30/20			
Adult Sponsor's Printed Name	Signature		Date of Review (mm/d	d/yy)		
(717) 506-3413	mfloreck@cvschools	s.org				
Phone	Email	9				

## **Student Checklist (1A)**

This form is required for ALL projects.

1.	a. Student/Team Leader: Aditya Kendre Grade: 12
	Email: kendreaditya@gmail.com Phone: (717) 622-1281
	b. Team Member: c. Team Member:
2.	Title of Project:
	Generative Adversarial Networks for PCG Arrhythmia Detection
3.	School: Cumberland Valley High School School Phone: (717) 506-3413
	School Address: 6746 Carlisle Pike
	Mechanicsburg, PA 17050
4.	Adult Sponsor: Mike Floreck Phone/Email: mfloreck@cvschools.org
	Does this project need SRC/IRB/IACUC or other pre-approval? ☐Yes • No Tentative start date:
6.	Is this a continuation/progression from a previous year?   ■Yes ■No If Yes:
	a. Attach the previous year's Abstract <b>and</b> Research Plan/Project Summary
	b. Explain how this project is new and different from previous years on Continuation/Research Progression Form (7)
7.	This year's laboratory experiment/data collection:
	10/30/20 03/01/21
	Actual Start Date: (mm/dd/yy) End Date: (mm/dd/yy)
Q.	Source of Data:
0.	Collected self/mentor  Other Describe/url:  Physionet Database
9. I	List name and address of all non-home and non-school work site(s):
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Ado	dress:
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em	· · · · · · · · · · · · · · · · · · ·
10.	Complete a Research Plan/Project Summary following the Research Plan/Project Summary instructions

and attach to this form.

11. An abstract is required for all projects after experimentation.

## **Research Plan/Project Summary Instructions**

# A complete Research Plan/Project Summary is required for ALL projects and must accompany Student Checklist (1A).

- All projects must have a Research Plan/Project Summary
  - a. Written prior to experimentation following the instructions below to detail the rationale, research question(s), methodology, and risk assessment of the proposed research.
  - b. If changes are made during the research, such changes can be added to the original research plan as an addendum, recognizing that some changes may require returning to the IRB or SRC for appropriate review and approvals. If no additional approvals are required, this addendum serves as a project summary to explain research that was conducted.
  - c. If no changes are made from the original research plan, no project summary is required.
- Some studies, such as an engineering design or mathematics projects, will be less detailed in the initial project plan and will change through the course of research. If such changes occur, a project summary that explains what was done is required and can be appended to the original research plan.
- The Research Plan/Project Summary should include the following:
  - a. **RATIONALE:** Include a brief synopsis of the background that supports your research problem and explain why this research is important and if applicable, explain any societal impact of your research.
  - b. RESEARCH QUESTION(S), HYPOTHESIS(ES), ENGINEERING GOAL(S), EXPECTED OUTCOMES: How is this based on the rationale described above?
  - c. Describe the following in detail:
- **Procedures:** Detail all procedures and experimental design including methods for data collection, and when applicable, the source of data used. Describe only your project. Do not include work done by mentor or others.
- Risk and Safety: Identify any potential risks and safety precautions needed.
- Data Analysis: Describe the procedures you will use to analyze the data/results.
  - d. **BIBLIOGRAPHY:** List major references (e.g. science journal articles, books, internet sites) from your literature review. If you plan to use vertebrate animals, one of these references must be an animal care reference.

# Items 1–4 below are subject-specific guidelines for additional items to be included in your research plan/project summary as applicable.

#### 1. Human participants research:

- a. Participants: Describe age range, gender, racial/ethnic composition of participants. Identify vulnerable populations (minors, pregnant women, prisoners, mentally disabled or economically disadvantaged).
- b. Recruitment: Where will you find your participants? How will they be invited to participate?
- c. Methods: What will participants be asked to do? Will you use any surveys, questionnaires or tests? If yes and not your own, how did you obtain? Did it require permissions? If so, explain. What is the frequency and length of time involved for each subject?
- **d. Risk Assessment:** What are the risks or potential discomforts (physical, psychological, time involved, social, legal, etc.) to participants? How will you minimize risks? List any benefits to society or participants.
- e. Protection of Privacy: Will identifiable information (e.g., names, telephone numbers, birth dates, email addresses) be collected? Will data be confidential/anonymous? If anonymous, describe how the data will be collected. If not anonymous, what procedures are in place for safeguarding confidentiality? Where will data be stored? Who will have access to the data? What will you do with the data after the study?
- f. Informed Consent Process: Describe how you will inform participants about the purpose of the study, what they will be asked to do, that their participation is voluntary and they have the right to stop at any time.

#### 2. Vertebrate animal research:

- a. Discuss potential ALTERNATIVES to vertebrate animal use and present justification for use of vertebrates.
- b. Explain potential impact or contribution of this research.
- c. Detail all procedures to be used, including methods used to minimize potential discomfort, distress, pain and injury to the animals and detailed chemical concentrations and drug dosages.
- d. Detail animal numbers, species, strain, sex, age, source, etc., include justification of the numbers planned.
- e. Describe housing and oversight of daily care.
- f. Discuss disposition of the animals at the end of the study.

#### Potentially hazardous biological agents research:

- a. Give source of the organism and describe BSL assessment process and BSL determination.
- b. Detail safety precautions and discuss methods of disposal.

#### 4. Hazardous chemicals, activities & devices:

- Describe Risk Assessment process, supervision, safety precautions and methods of disposal.
- Material Safety Data Sheets are not necessary to submit with paperwork.

# GENERATIVE ADVERSARIAL NETWORKS FOR PCG ARRHYTHMIA DETECTION

#### Aditya Kendre

Cumberland Valley High School Mechanicsburg, PA 17050

January 3, 2021

#### **ABSTRACT**

With the rapid growth of computational power and complex algorithms, we propose a novel approach to detect arrhythmias in Phonocardiograms (PCGs). Typically, Electrocardiograms are used to diagnose arrhythmias; requiring medical grade equipment to accurately recognize cardiac illnesses (Rajpurkar et al., 2017). PCGs, however, provide ease of access to everyone who has a device capable of recording audio, allowing medical professionals to treat arrhythmias in the developmental stages. The new design comprises two subsystems; one is based on the relationship between Electrocardiograms (ECGs) and PCGs, and the other between PCGs and arrhythmias. The association between ECGs and PCGs is amended to translate from one space to another, where ECGs become dimensionally reduced, then reconstructed into a PCG signal. The second subsystem uses a Generative Adversarial Networks (GAN), in which both arbitrary PCG signals are generated, and preexisting ECG datasets are recreated into PCG signals (using subsystem one). These signals are fed into a classifier that detects if an arrhythmia is present. This proposed system's advantage is that PCG data is more readily available than ECG data; hence, more heart diagnostics can be made.

#### 1 Introduction

**Problem Statement.** Every physical examination done with a stethoscope should aim to diagnose any arrhythmias present within a patient.

**Question.** Is it possible to create a model capable of surpassing the accuracy of Cardiologists in identifying heart arrhythmias in Phonocardiograms?

**Hypothesis.** It is possible to exceed the accuracy of Cardiologists when compared to that of a Generative Adversarial Network's, to identify heart arrhythmias in Phonocardiograms.

#### Materials List. Computer.

Electrocardiograms have created a profound impact in the field of cardiology, specifically in recognizing heart arrhythmias, a problem with the rhythm of one's heartbeat. Noninvasive arrhythmia analysis is based on multiple electrodes that reflect the electrical activity on ECGs. However, with the recent surge of heart-related medical cases, it is getting difficult to diagnose heart conditions at an early stage. As most treatments rely on detecting the disease in it's infancy stages. Traditionally, arrhythmias are diagnosed by cardiologists by analyzing ECG recordings (Jordaens, 2018). Some clinics have adopted a new technique in which ECG and PCG signals are simultaneously recorded and then computationally analyzed. This, however, still requires an instrument capable of recording ECG data. Such instruments are only available during scheduled appointments, often which are recommended by physicians. If a physician fails to detect symptoms of arrhythmia, a patient may never receive a diagnosis. One study found 44% of cardiologists were not able to detect common cardiac events with stethoscopes (Mangione et al., 1993); in another study, delays in cardiac-related illness diagnosis and treatment impacted procedural success rates by as much as 24% (Bunch et al., 2013). We propose a method where it is now possible to accurately detect arrhythmias with only PCG recordings. This provides an opportunity for physicians to check for potential developments of cardiac arrhythmias at every physical exam accurately.

Current PCG arrhythmia diagnosis methods only recognize between Normal and Abnormal (binary classification), providing minimal information about what is present within the PCG signal (Aziz et al., 2020). This is because no PCG datasets exist that include more than 3 classes of arrhythmia. Therefore, it is necessary to transform pre-existing ECG datasets with multiple classes to PCG signals. This enables models to detect a larger range of arrhythmia without explicitly collecting new PCG recordings. Currently, no technology attempts to construct PCG signals from existing ECG data.

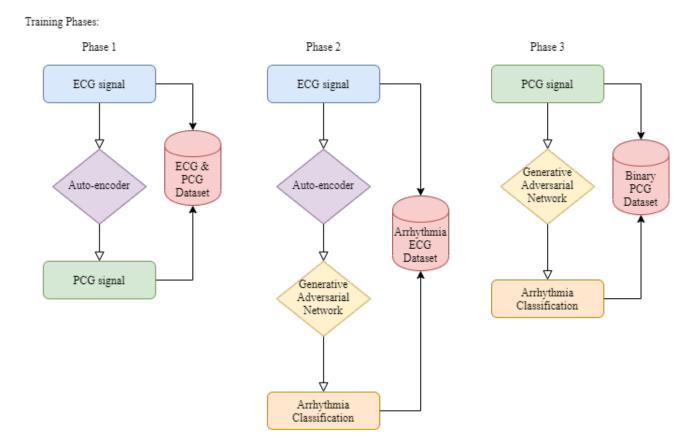
#### 2 Methodology

#### 2.1 Approach

The model contains two sub-models, an Autoencoder (AE), and a Generative Adversarial Network (GAN). The AE is responsible for extracting relevant features from an ECG signal and constructing a PCG signal from the latent features. The GAN is responsible for extracting relevant features and classifying the PCG signals.

The training phase involves 3 stages: AE training, AE+GAN training, and GAN fine-tuning. During training phases, all datasets will follow the following split: 70% - training, 15% - validation, 15% - testing; this cross-validation step validates that both models are not overfitting during the training phase. The first stage involves training the AE with a supervised dataset of ECG and PCG signals (Liu et al, 2016). The second stage involves training both the AE and the

GAN with a supervised dataset of arrhythmias within ECGs (Goldberger et al., 2017). During the training process, the AE model will be frozen (the weights and biases of the AE model won't be trained) as this process is already done in the preceding stage. The last stage is fine-tuning the GAN on a binary supervised dataset of PCG signals (Normal vs Abnormal). This validates the model's metrics in the previous step.



#### 2.2 Data Analysis

While testing and training, the model will be validated against with metrics such as recall, precision, accuracy, loss, FBeta, F1 score, and ROC/AUC score. These tests will ensure that the model is accurately predicting the classes, and identifying important features within the datasets. Each step in the training phase will represent a milestone and an accuracy of 97% will mark the completion criteria.

#### 2.3 Potential Problems

**Overfitting:** One of the largest problems in Deep Learning overall, which possesses a threat to our model is overfitting. Overfitting typically happens when the model metrics of the training and validation set diverge. This suggests that the model is not generalizing, but rather memorizing the training dataset. To combat overfitting, researchers typically implement data argumentation techniques to reinforce important features in a dataset.

**Domain Shift:** A domain shift occurs when a source dataset performs well but on a different dataset distribution, the performance drastically decreases. Typically, domain adaptation is often used to improve performance on target datasets. This is done by training the model itself on multiple datasets to improve the model's capacity to generalize.

**Traning Time:** With large multi-model architectures, it becomes tough to train models on a single GPU. This can happen for a number of reasons, but the main reason is because the model takes up too much memory of the GPU. Generally, parallel processing is used to split tasks and assign them to different GPUs. For instance, the AE model will run on a single GPU, while the GAN will run on another GPU.

#### References

- [1] Jordaens, L. (2018). Aclinical approach to arrhythmias revisited in 2018. Netherlands Heart Journal. doi:10.1007/s12471-018-1089-1
- [2] Mangione, S. (1993). The Teaching and Practice of Cardiac Auscultation during Internal Medicine and Cardiology Training: A Nationwide Survey. Annals of Internal Medicine, 119(1), 47. doi:10.7326/0003-4819-119-1-199307010-00009
- [3] Bunch, T. J., May, H. T., Bair, T. L., Johnson, D. L., Weiss, J. P., Crandall, B. G., ... Day, J. D. (2013). Increasing time between first diagnosis of atrial fibrillation and catheter ablation adversely affects long-term outcomes. Heart Rhythm, 10(9), 1257–1262. doi:10.1016/j.hrthm.2013.05.013
- [4] Aziz, S., Khan, M. U., Alhaisoni, M., Akram, T., Altaf, M. (2020). Phonocardiogram Signal Processing for Automatic Diagnosis of Congenital Heart Disorders through Fusion of Temporal and Cepstral Features. Sensors, 20(13), 3790. doi:10.3390/s20133790
- [5] Felipe Alonso "Detection of life threatening arrhythmias using feature selection and support vector machines", IEEE Transactions on Biomedical Engineering, Vol 61No.3, pp.832-840, March 2014.

## **Approval Form (1B)**

A completed form is required for each student, including all team members.

## 1. To Be Completed by Student and Parent

- a. Student Acknowledgment:
  - I understand the risks and possible dangers to me of the proposed research plan.
  - I have read the ISEF Rules and Guidelines and will adhere to all International Rules when conducting this research.
  - I have read and will abide by the science fair ethics statement.

Aditya Kendre Student's Printed Name  Signature  b. Parent/Guardian Approval: I have read and unconcented Plan/Project Summary. I consent to read and unconcented Plan/Project Summary.			10/:	10/29/20		
		erstand the risks	Mus) and possible d	Acknowledged (mm/dd/yy) t be prior to experimentation.) angers involved in the		
Nivrutti Kendre	no de la composition della com	y cruid participal		29/20		
Parent/Guardian's Printed Name Signatu		•		Date Acknowledged (mm/dd/yy) (Must be prior to experimentation.)		
2. To be completed by the loca (Required for projects requiring pr	ior SRC/IRB APP		or 2b as appro	priate.)		
<ul> <li>a. Required for projects that need prior SRC/IRB approval BEFORE experimentation (humans, vertebrates or potentially hazardous biological agents).</li> <li>The SRC/IRB has carefully studied this project's Research Plan/Project Summary and all the required forms are included. My signature indicates approval of the Research Plan/Project Summary before the student begins experimentation.</li> </ul>		Research approval This project w (not home or by the proper complies with	Research Institutions with no prior fair SRC/IRB			
SRC/IRB Chair's Printed Name		SRC Chair's Pr	inted Name			
	oval (mm/dd/yy) o experimentation.)	ONO Chairs Fi	mileu Name			
(index be prior to	o experimentation.)	Signature		Date of Signature (mm/dd/yy) (May be after experimentation)		
. Final ISEF Affiliated Fair SRC						
<b>SRC Approval After Experimentation and B</b> I certify that this project adheres to the appr				th all ISEF Rules.		
Regional SRC Chair's Printed Name	Signature		Date	e of Approval (mm/dd/yy)		

Signature

(where applicable)

Date of Approval (mm/dd/yy)

## **Risk Assessment Form (3)**

Must be completed before experimentation.

St	udent's Name(s) Aditya Kendre
Tit	tle of Project Generative Adversarial Networks for PCG Arrhythmia Detection
	be completed by the Student Researcher(s) in collaboration with Designated Supervisor/Qualified eientist: (All questions must be answered; additional page(s) may be attached.)
1.	List all hazardous chemicals, activities, or devices that will be used; identify microorganisms exempt from pre-approval (see Potentially Hazardous Biological Agent rules).  The only device used in this research project is a laptop.
2.	Identify and assess the risks and hazards involved in this project.  N/A
3.	Describe the safety precautions and procedures that will be used to reduce the risks.  N/A
4.	Describe the disposal procedures that will be used (when applicable).  N/A
5.	List the source(s) of safety information.  N/A
I F	To be completed and signed by the Designated Supervisor (or Qualified Scientist, when applicable): agree with the risk assessment and safety precautions and procedures described above. I certify that I have reviewed the Research Plan/Project Summary and the International Rules, including the science fair ethics statement and will provide lirect supervision.
[	Designated Supervisor's Printed Name Signature Date of Review (mm/dd/yy)
_ F	Phone or email contact information
_	

## **Continuation/Research Progression Projects Form (7)**

Required for projects that are a continuation/progression in the same field of study as a previous project. This form must be accompanied by the previous year's abstract and Research Plan/Project Summary.

Student's Name(s) Aditya Kendre

To be completed by Student Researcher: List all components of the current project that make it new and different from previous research. The information must be on the form; use an additional form for previous year and earlier projects.

Components	Current Research Project	Previous Research Project: Year: 19-20
1. Title	Generative Adversarial Networks for PCG Arrhythmia Detection	ECG-Based Abnormal Heartbeat Classification: A Deep Learning Approach for Arrhythmia Detection
2. Change in goal/ purpose/objec- tive	To create a lightweight, precise, and accurate model for predicting heart arrhythmias in Phonocardiograms using a Generative Adversarial Network capable of exceeding Cardiologists' accuracy.	To create a model capable of surpassing the accuracy of Cardiologists in identifying heart arrhythmias in Electrocardiograms.
3. Changes in methodology	A Generative Adversarial Networks comprises of two models: a generator model and a classifier model (which contains a Convolutional Neural Network). The generator creates artificial PCG data to deceive the classifier into predicting the data is a real PCG signal while simultaneously being fed true PCG data from a dataset.	A Convolutional Neural Network extracts latent features from an electrocardiogram database following a fully-connected Linear layer that predicts whether an arrhythmia is present within the electrocardiogram, based upon the features extracted by the CNN.
4. Variable studied	Manipulated variables include: Learning Rate, Batch size, Number of Epochs, Hidden Layers, Hidden Units, Activations Functions, and level of Data Augmentation.  Responding variables include: Loss, Accuracy, Recall, Precision, F-Beta Score, F1 Score, and ROC and AUC.	Manipulated variables include: Number of layers, Hidden Units, and the level of Data Augmentation.  Responding variables include: Loss and Accuracy.
5. Additional changes	Conversion between ECG and PCG signals using an Autoencoder.	ECG signal with a one-dimensional CNN.

Attached are:  Abstract and Research Pla	an/Project Summary, Year	
I hereby certify that the above board properly reflect work of	ve information is correct and that the done only in the current year.	current year Abstract & Certification and project display
Aditya Kendre	aditye	10/30/20
Student's Printed Name(s)	Signature	Date of Signature (mm/dd/yy)

#### OFFICIAL ABSTRACT and CERTIFICATION

_		1271201117101 4114 62					
A Deep Learning Approach for Arrhythmia Detection  Aditya Kendre Cumberland Valley High School, Mechanicsburg PA, Adams County  Early detection of cardiac arrhythmia has the potential to prevent the millions of						Category Pick one only — mark an "X" in box at right  Animal Sciences  Behavioral & Social Sciences  Biochemistry  Biomedical & Health Sciences  Biomedical Engineering  Cellular & Molecular Biology  Chemistry  Computational Biology & Bioinformatics  Earth & Environmental Sciences  Embedded Systems Energy: Sustainable Materials and Design  Engineering Mechanics  Environmental Engineering  Materials Science	
Early detection of cardiac arrhythmia has the potential to prevent the millions of moralities that the disease causes globally. However, there are few automated systems to identify arrhythmia. A significant impediment in achieving successful methods include the lack of a large training dataset. Despite this difficulty, processes like data augmentation allow for anincreased amount and diversity of data. Here, the electrocardiogram (ECG) datasets were obtained from the PhysioNet database. The dataset was used to train a Convolutional Neural Network (CNN) on classifying cardiac arrhythmia. Experimental results illustrate advantages such as better responsiveness and higher accuracy of deep learning-based models when compared to the traditional analysis on ECGs.							
1.	As a part of this research proj interacted with (check ALL tha	•	/ handled, m	nanipula	ited, or	Mathematics Microbiology	
	☐ human participants	☐ potentially hazardo	ous biologica	al agent	S	Physics & Astronomy	
	□ vertebrate animals	☐ microorganisms	☐ rDNA		☐ tissue	Plant Sciences	
2.	I/we worked or used equipme					Robotics & Intelligent Machines Systems Software	
	or industrial setting:					Translational Medical	
3.	This project is a continuation of	of previous research.	[	⊐ Yes	■ No	Sciences	
4.	My display board includes nor depictions of humans (other t		hs/visual [	⊐ Yes	■ No		
5.	This abstract describes only p reflects my/our own independ work only	•	•	■ Yes ear's	□No		
6.	I/we hereby certify that the ab above statements are correct	•		■ Yes ork.	□No	\	
an	is stamp or embossed seal atte d state laws and regulations a en obtained including the final	nd that all appropriate	reviews and	l approv	vals have		

ECG-Based Abnormal Heartbeat Classification: A Deep Learning Approach for Arrhythmia

Detection

Aditya Kendre

Cumberland Valley High School

#### Rationale

Electrocardiograms (ECG) have created a profound impact in the field of cardiology, specify in recognizing of heart arrhythmias. Non-invasive arrhythmia analysis is based on 10 electrodes that reflect the electrical activity on ECGs. An estimated three million cases of arrhythmia occur in the United States yearly (Mayo Clinic). Diagnosing this disease early is the key to one's wellness, yet 18% of cardiologists misinterpreted ECGs containing atrial fibrillation (Anh et al, 2006). With the recent advancements in technology, Machine Learning algorithms such as Deep Neural Networks (DNNs), allow a computer to learn features and identify patterns within a given dataset. On the basic level, DNNs receive input data, and through a series of weights and biases, outputs a confidence value in all possible labels of the dataset, similar to a human's neural network. Furtherance in the accuracy of abnormal heartbeat classification will allow cardiologists to accurately, and efficiently recognizing arrhythmia before becoming prevalent in one's wellbeing.

#### Research

Research Question: This research project will examine whether a classifier will be able to accurately identify abnormal heartbeat in ECGs.

Hypothesis: If an image classifier received a supervised dataset of heart arrhythmia of ECGs, then the image classifier will allow an accurate identification of arrhythmia.

Expectation: The image classifier should reach an accuracy of above 82%.

## Procedure:

- 1. Gather a dataset of annotated ECGs
- 2. Determine type of classifier used to learn dataset features
- 3. Analyze results using Gradient Decent and Mean Loss function

### Risks and Safety:

This research project involves no risks or safety concerns.

#### References

- Alfaras, Miquel, Soriano, & Silvia. (2019, July 3). A Fast Machine Learning Model for ECG-Based Heartbeat Classification and Arrhythmia Detection. Retrieved October 30, 2019, from https://www.frontiersin.org/articles/10.3389/fphy.2019.00103/full.
- Mayo Clinic. (2019, April 2). Heart arrhythmia. Retrieved October 30, 2019, from https://www.mayoclinic.org/diseases-conditions/heart-arrhythmia/symptoms-causes/syc-20350668?utm\_source=Google&utm\_medium=abstract&utm\_content=Cardiac-arrhythmia&utm\_campaign=Knowledge-panel.
- Srinivasan, N. T., & Schilling, R. J. (2018, June). Sudden Cardiac Death and Arrhythmias.

  Retrieved October 30, 2019, from https://www.ncbi.nlm.nih.gov/pmc/articles/PMC6020177/.