

sdmay23-05



Livestock Well-being Identification

# The Team

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Advisor/Client: Dr. Ashraf Gaffar, Teaching Professor [E CPE]

# Project Context

- 2021 USDA Statistics on farms in Iowa:
  - Over 30 million acres
  - 84,900 farms
  - About 42 million livestock
- Our goal: Identify illness in farm animals
- How? By training an AI model to identify patterns in the spectrogram and differentiate between healthy and sick animals

# Applications

- Identifying illness in farm animals
- Livestock farms
- Wilderness Conservation



# Functional Requirements

- Capture the sound of different sick and healthy animals
- Encode the sound waves
- Train the machine to recognize different sound patterns
- Model should reach an accuracy that is desired by the client (TBD)
- Model should be scalable (able to handle different sized datasets)

# Resource Requirements

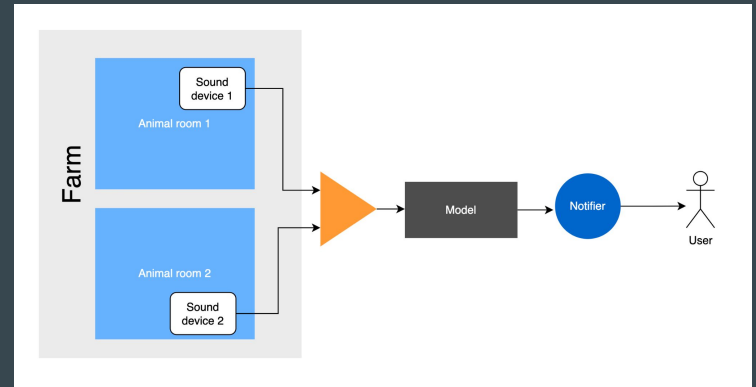
- Linux Virtual Machine
- GPU
- Keras framework
- Anaconda
- Various sounds of sick and healthy animals

# Design Overview

- Collecting sound recordings of healthy and sick animals
- Supervised Deep neural networks (DNNs)
- Validation strategies (cross-validation)

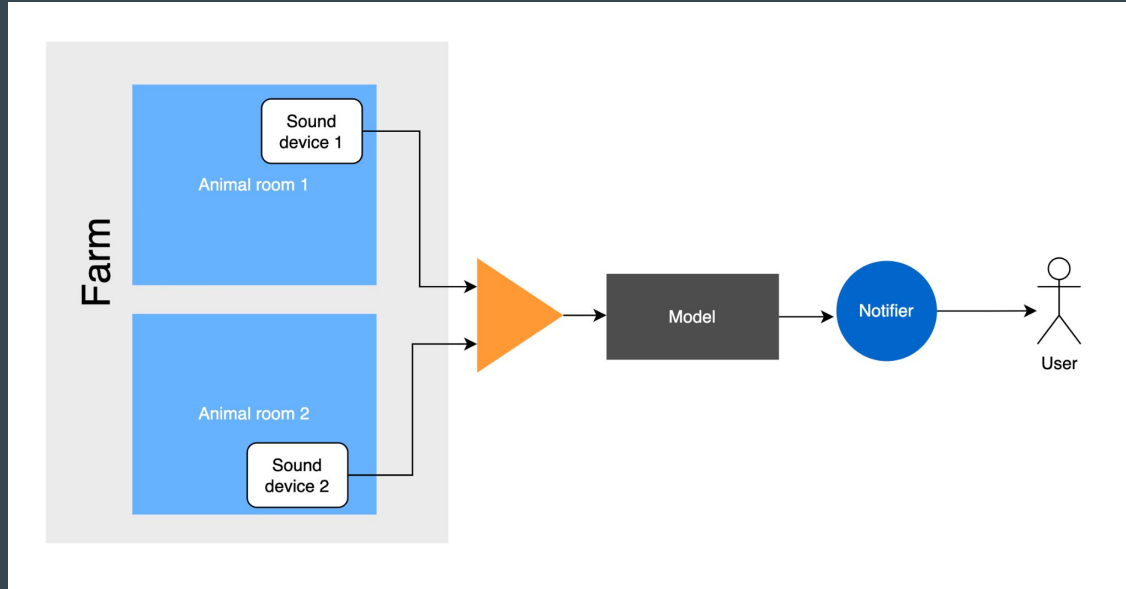
# Design Complexity

- Requires comprehensive understanding of sound frequencies, vocalizations and noise animals make in various health states
- Detailed knowledge about the animals natural behaviors
- Robust recognition capabilities
- Analyze various aspects of an audio sample (frequency spectrum , timbre)





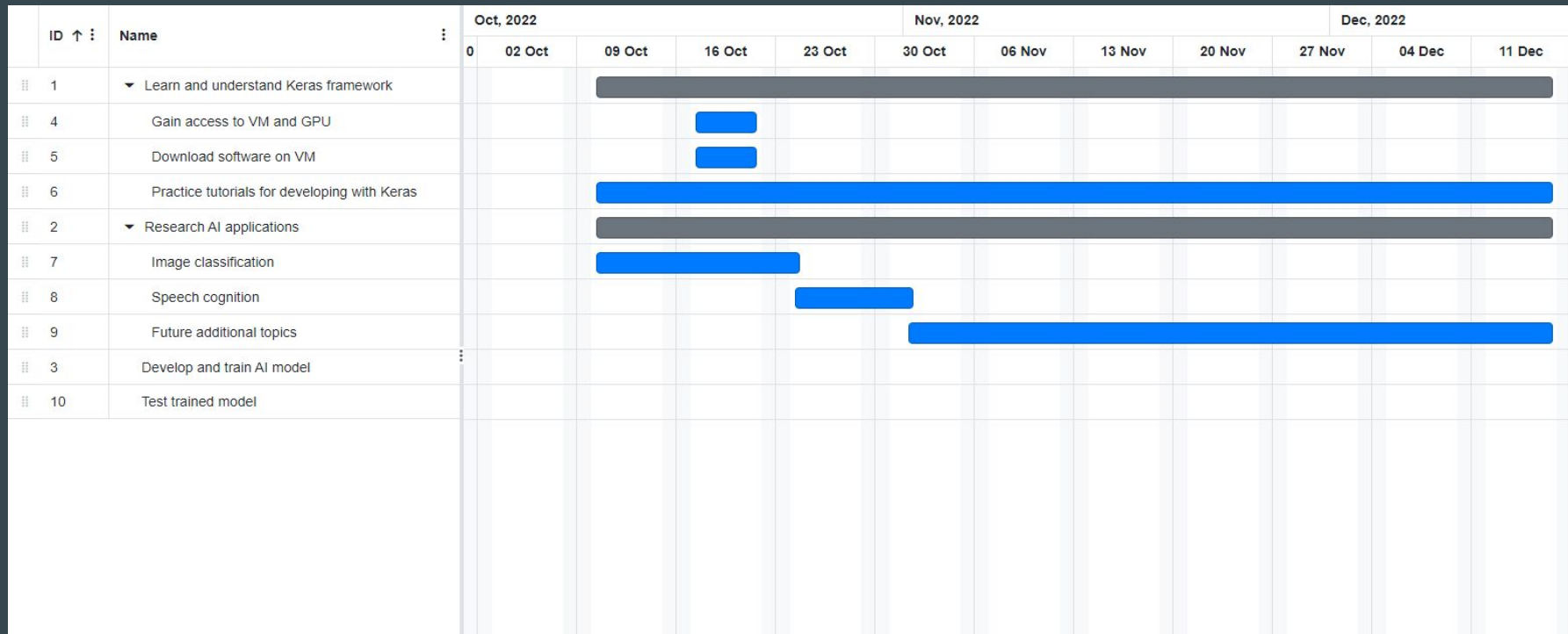
# Example Functionality



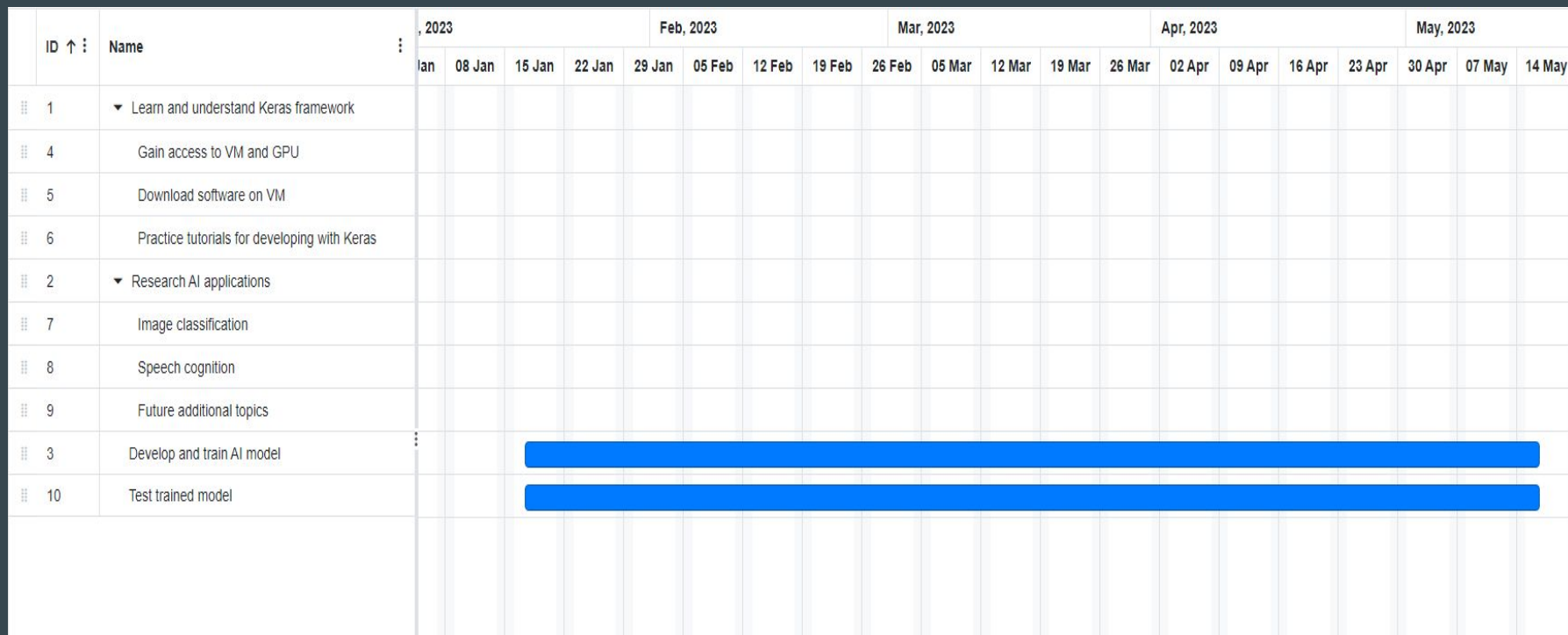
# Project Plan

- Analyze AI domain and available technologies
- Learn how to use Keras development framework
- Collect sound data of different sick and healthy animals
- Develop AI model to analyze animal sounds
- Adjust the model until the client's requirements are met

# Fall Semester Gantt Chart



# Spring Semester Gantt Chart



# Risks and Mitigation Plan

- Model might not be developed well enough to achieve desired accuracy
  - Risk factor  $> 0.5$
  - Mitigation plan: Improve code for the model, seek technical help from advisor
  
- Not enough sound data to achieve desired results
  - Risk factor  $> 0.5$
  - Mitigation plan: Work with advisor to collect additional data from other sources

# Test Plan

- It is not traditional testing.
- Measure and compare output.
- Measure the distance.
- Change the hyperparameters and train again.

# Testing Results

- Ensure that AI Model can identify the sick from healthy ones
- Testing Process includes Validation and Verification Process
- In Validation step, we will test AI Model with set of known data and use it to verify if our model identifies the sick animal correctly or not
- In Verification step, we will test AI Model with set of new data to ensure that the output is still accurate

# Testing Implementation

- We will receive the data from our client
- Convert the sounds to spectrograms
- Train the model using these sound datas
- Adjust the model until it meets our client expectations
- Ensure that the result is highly accurate at the end.



# Conclusions

- What we've done so far:
  - Researched image classification, speech cognition, and sound patterns AI
  - Learned how to use Keras and TensorFlow
  - Conceptualized design of our model
- Plans for next semester:
  - Develop solution
  - Train model
  - Adjust model

# Citations

- “2021 State Agriculture Overview.” USDA/NASS 2021 State Agriculture Overview for Iowa, [https://www.nass.usda.gov/Quick\\_Stats/Ag\\_Overview/stateOverview.php?state=IOWA](https://www.nass.usda.gov/Quick_Stats/Ag_Overview/stateOverview.php?state=IOWA).