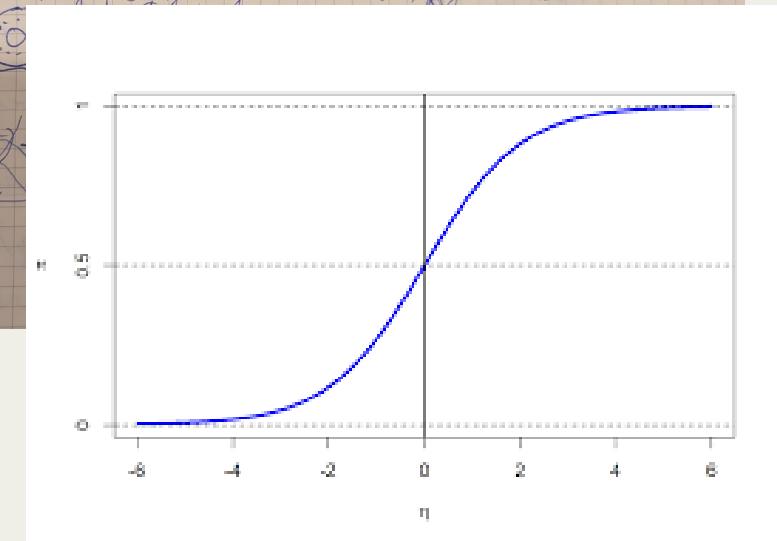
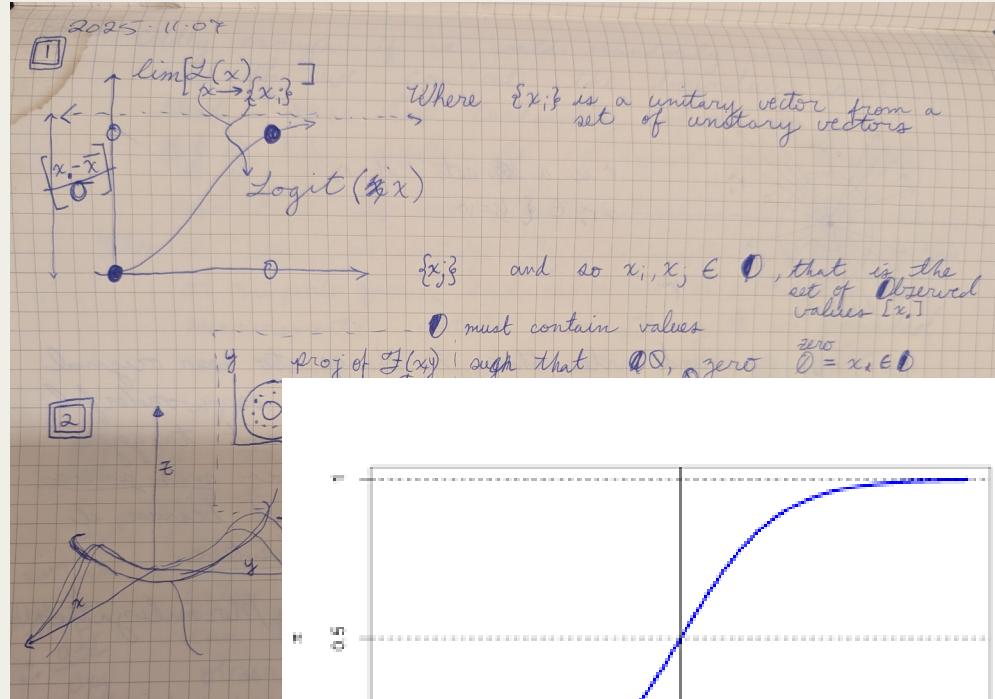


# FieldChartOCR

Extraction of Handwritten Charts and Tables

# Introduction

This presentation summarizes the FieldChartOCR project proposal, focusing on extracting and digitizing handwritten charts and tables. It addresses the challenges faced in manual chart digitization and outlines a novel solution. The goal is to provide a clear overview for supervisory-level stakeholders on project motivation, approach, and evaluation plans.



01

# Project Introduction

# **FieldChartOCR**

Hand-written Field Notes to Digital Charts

Presented by Chris Brechin

Date: November 12<sup>th</sup>, 2025

# Motivation and Problem Statement

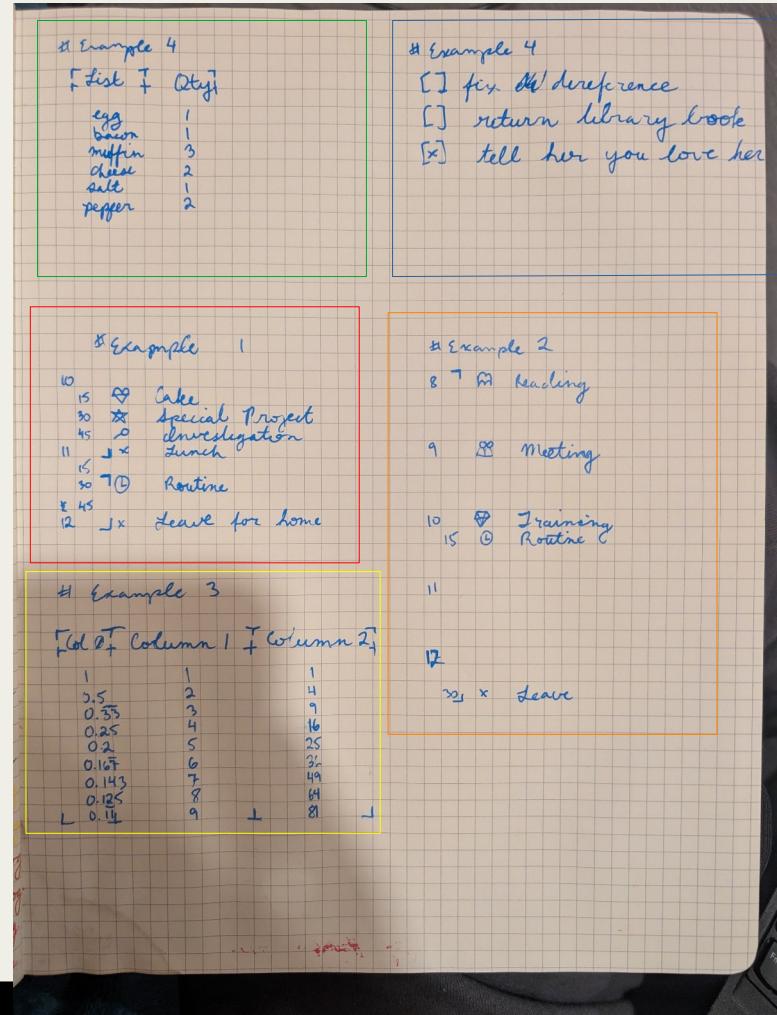
Digitizing handwritten charts is difficult due to varied handwriting styles, inconsistent layouts, and noise in scanned images. Manual transcription is slow and error-prone, causing significant pain points for data accessibility and analysis. **FieldChartOCR** aims to automate this process, improving accuracy and efficiency in converting handwritten visual data into machine-readable formats.

# Key Challenges in Digitizing Handwritten Charts

Handwritten charts are challenging to digitize due to inconsistent handwriting, diverse chart formats, and image noise. These factors lead to difficulties in automated recognition and interpretation. The lack of standardization increases error rates, making manual transcription time-consuming and costly. Addressing these challenges requires robust preprocessing, classification, and OCR techniques tailored specifically for handwritten data.

## Examples:

1. Coded Timesheet
2. Sparse, Coded Timesheet
3. Table with mixed, float and integer values
4. Table with mixed, string and integer values
5. Checklist

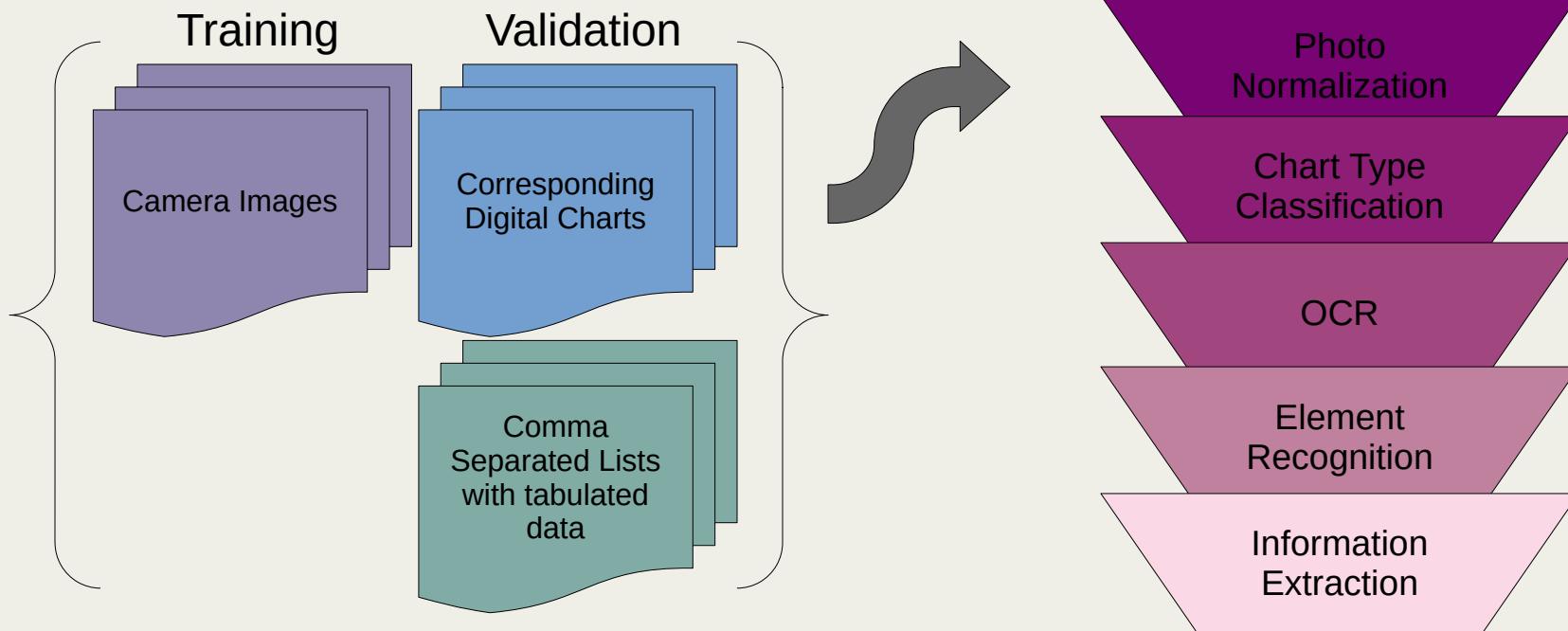


02

## Technical Approach and Planning

# Overview of FieldChartOCR Solution

FieldChartOCR employs a modular approach combining preprocessing, classification, and detection models. It will integrate OCR tailored for handwriting, alongside semantic assembly modules that reconstruct charts and tables accurately. I aim to automate the extraction pipeline, significantly reducing manual effort while maintaining high accuracy in digitizing complex handwritten visual data.

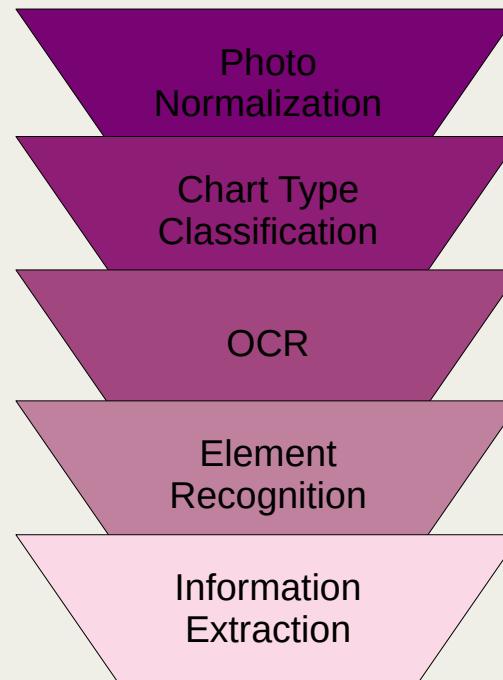


# System Architecture and Processing Pipeline

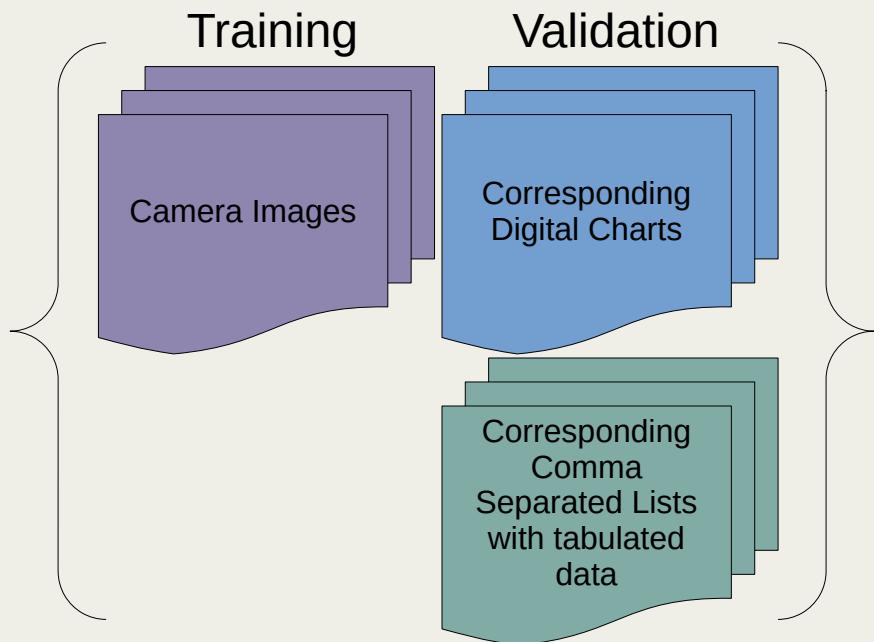
The pipeline consists of four key stages:

1. preprocessing to enhance image quality;
2. classification to identify chart types;
3. detection and OCR to extract handwritten text and graphical elements; and
4. semantic assembly to organize extracted data into structured formats.

Each stage leverages machine learning models for the specific purposes.



# Data Sources and Evaluation Metrics



Data sources include annotated datasets of handwritten charts and synthetic data generated to augment training diversity. Success will be measured using classification F1 scores, detection accuracy, Character Error Rate (CER) for OCR, and runtime performance. These metrics will guide iterative improvements, ensuring the system meets practical usability and reliability standards.

# Conclusions

**FieldChartOCR** targets the automated digitization of handwritten charts with a robust, multi-stage system architecture. Initial milestones focus on achieving an MVP with reliable extraction accuracy and efficiency. Future work will explore enhanced semantic analysis and broader dataset integration, aiming to deliver a scalable solution for diverse real-world applications.

# Thanks!

**Do you have any questions?**

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