

# HTML Code Generation using CNN Algorithm

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**Abstract**— Website creation starts with creating rough draft of each web page either with the help of designing tools or drawing by hand. Equivalent code is then generated for draft of web page. Above process is carried out until desired web page is constructed and user is satisfied with the results. This process is repetitive, expensive and time consuming. Hence, proposed system aims to automate this process. Hand drawn image of form is given as an input and it is processed and various components are detected. Once the components are detected they are cropped and these components are recognized using deep learning CNN methods. On recognition of the component equivalent HTML code is constructed using HTML builder algorithm.

**Index Terms**—Object detection, object recognition, deep learning, HTML

## I. INTRODUCTION

The process of automatic code generation based on the hand drawn mockup images of Graphical User Interface created by designers is expensive in practice. There are various real time software applications that provide facility to generate code for user interfaces such as Android Studio, Xcode and Eclipse IDE. These software applications are quite expensive and cumbersome. It is not easy to convert user interface drawings into working user interface code as development process requires time and proper skills. At present programmer perform this task manually which is costly and error prone [3]. Developing GUI based application to automatically generate code will help developers to save time on GUI implementation.

Proposed system presents an approach that constructs HTML code for hand drawn GUI image. It involves

processes namely detection, dilation, erosion, classification and assembly of an image [4]. The initial process involves detecting borders of various GUI components such as text boxes, labels, drop down list, check box, button, etc. It involves 2 morphological processes: erosion and dilation. Dilation adds pixels to the image that expands the image and erosion removes pixel on object boundaries.

Once the GUI components are detected, they need to be classified. Thus second task is image classification. The proposed system is implemented using RCNN model. RCNN can detect multiple regions of image (up to 2000 different regions). Once object detection and image classification is performed, HTML code will be generated for specific component [3], [4], [6].

## II. PROBLEM STATEMENT

To generate HTML code from hand drawn image of web form using object detection and machine learning algorithms.

## III. LITERATURE SURVEY

*A. Generating Code from GUI Screenshots*  
Tony Beltramelli, Ulzard Technologies, Copenhagen, Denmark published a paper [6] dedicated to generating code from a GUI Screenshots. A system is used to transform GUI screenshots into computer code. Deep learning methods such as Convolutional and Recurrent Neural Networks were used for automatic generation of code from an input image with accuracy of 77% for platforms like iOS, Android and web-based technologies.

*B. Reverse Engineering Mobile Application*

Tuan Anh Nguyen, Cristoph Csallner, Computer Science and Engineering Department, The University of Texas at Arlington, USA published a paper [3] dedicated to Reverse Engineering Mobile Application. Computer vision and optical character recognition (OCR) techniques are used in REMAUI to identify various elements such as images, texts, containers and lists on a given input. 448 screenshots of iOS and Android applications were used in this system. User interfaces generated by REMAUI were similar to the original screenshots

### C. Machine Learning Based Prototyping of Graphical User Interfaces for Mobile Apps

Kevin Moran, Carlos Bernal Cárdenas, Michael Curcio, Richard Bonett, Denys Poshyvanyk published paper [4] dedicated to Machine Learning Based Prototyping of Graphical User Interfaces for Mobile Apps. This system is used to transform a GUI into code. Initially, components of a GUI were detected from an artifact. Then, GUI components were accurately classified into domain specific types (e.g. toggle-button) using automated dynamic analysis, software repository mining and deep convolutional neural networks. Finally, prototype application can be combined from a GUI structure generated using K-nearest neighbors algorithm. This was implemented for Android system.

## IV. EXISTING SYSTEM

The importance of Internet websites have increased due to progress in technology. Websites reflects states, institutions, communities and people which are used for product marketing or advertising purpose. Any website is the interface from which user interacts with the system. Hence, it's very important to design interactive web pages which attract attention of users. In existing system, software experts and designers along with the end-users work together for developing design for web page. Draft of design of user interface is created by designers either on paper or graphics tools. Then based on this draft the code for the web pages is developed by software experts. There may be changes in web pages according to the feedback received by the end users and this is repetitive task. Rewriting code for similar components is tedious task. Hence, to automate the production of web pages is all about Code Generation using CNN Algorithm.

## V. PROPOSED SYSTEM

The system is used to digitize the work of generation of web form. Hand-drawn image of a web form is given as an input to the system and equivalent HTML code is to be generated.

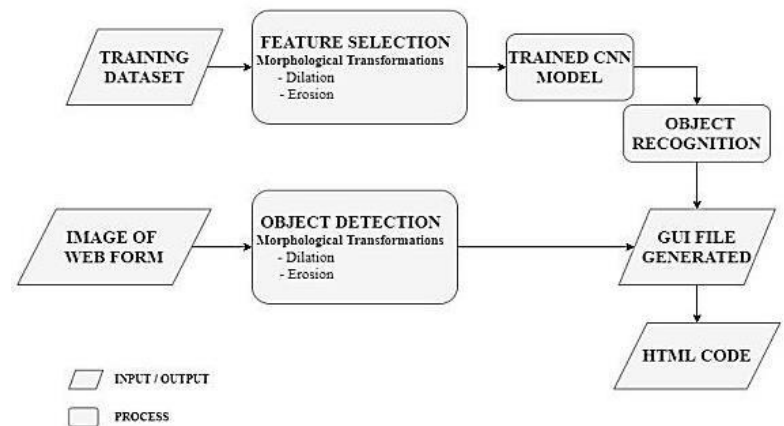


Fig 1: System Overview

### A. Object Detection

Dataset [5] consists of various components like textbox, buttons, and labels. These components in an image are detected using contour detection and morphological transformation. Contour detection is used to draw borders outside the component. Small objects are removed using morphological erosion so that only substantive objects are obtained. Gaps in objects are filled using morphological dilation so that objects become more visible. These techniques make object more visible and clear.

### B. Object Recognition

The dataset is trained using Convolutional Neural Network. Input image of web form is then given to this trained model which recognizes components in an image.

### C. Generate DSL Code with .gui File Extension

Figure 2 shows the content of DSL code with .gui file extension of an image of a web form. This .gui file is generated for the recognized image which contains information about various components in an image. This file gives a base to develop a HTML code for the given form.

```

<START>
header
{
  btn-inactive , btn-inactive
}
row
{
  quadruple {{ small-title , text , btn-orange }
  quadruple { small-title , text , btn-orange }
  quadruple { small-title , text , btn-orange }
  quadruple { small-title , text , btn-orange }
}
row
{
  quadruple { small-title , text , btn-orange }
  quadruple { small-title , text , btn-orange }
  quadruple { small-title , text , btn-orange }
  quadruple { small-title , text , btn-orange }
}
<END>

```

Fig 2: Format of .gui file

#### D. HTML Code Generation

HTML code is then generated using the .gui file generated in previous step for the recognized components. The aim is to encode the recognized components based on the web page. First, templates for a header and footer are created. Then detect number of items on each row and map the labels of the components with their codes. Finally, combine the header, body and footer.

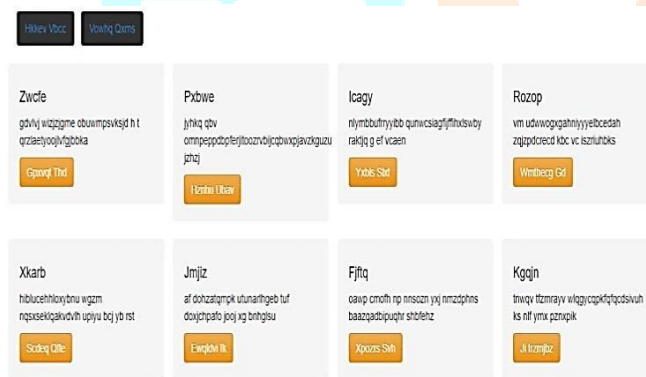


Fig 3: HTML code generated

#### VI. ADVANTAGES

1. Quick Development of front-end of website or mobile GUI.
2. Cost and time effective as time required to develop the front-end is minimized.
3. Easy to process digital image.

#### VII. LIMITATIONS

1. Input should be an image.

2. Attractive designs in input image are difficult to detect and create.
3. The system may not able to detect all GUI components.

#### VIII. CONCLUSION

Website is an important part in this growing world of technology as it is useful for product marketing and advertising purpose. Hence, it is important to build a website which will attract large number of users. In existing system building a website is a time consuming process. The proposed system takes hand drawn images of web form and convert it into HTML code. Hence, the system will consume less time and resources as compared to existing system.

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