# Technical Report: Al Models and Tools for No-Code Frontend Remix Platform

## **Overview**

This document presents a deep technical analysis of the AI models and training strategies relevant to building the "AI-Powered No-Code Frontend Remix Platform." It outlines the models used for image segmentation, OCR, screenshot-to-code translation, and text-based UI generation. Additionally, it explains which models can be fine-tuned or trained from scratch and how they can be integrated effectively.

# 1. Screenshot-to-Component Extraction

**Goal:** Automatically break down a webpage screenshot into meaningful visual components.

# A. Meta Segment Anything (SAM)

- Model Type: Foundation vision transformer for zero-shot segmentation
- Working: Produces binary masks for selectable regions in an image using sparse prompts
- Fine-tuning: Not needed. You can extend it by applying classifiers on top of masks
- Trainability: Not directly trainable; intended for generic use
- Integration: Use masks as bounding boxes and then classify each with an LLM or CNN

### B. Detectron2 (by Meta)

• Model Type: Object detection and instance segmentation framework

- Working: Outputs class-labeled bounding boxes and pixel-level masks
- **Fine-tuning:** YES, with your own dataset (e.g., UI screenshots labeled as buttons, inputs, cards)
- Train Strategy: Annotate with LabelMe/Roboflow → Train Detectron2 via PyTorch on a GPU
- Use Case Fit: Suitable if you want to custom-train on your Figma/screenshot dataset

### C. YOLOv8 (Ultralytics)

- Model Type: Real-time object detection
- Working: Returns (class, bounding\_box) tuples
- Fine-tuning: YES, very easy to fine-tune on a custom UI dataset
- Train Strategy: Use Roboflow to label your dataset and export it in YOLOv8 format
- **Use Case Fit:** Ideal for fast, component-aware UI segmentation with custom class types

# 2. OCR for UI Text Extraction

Goal: Extract all visible text from screenshots, such as form labels and headings

#### A. PaddleOCR

- Model Type: CNN + Transformer-based OCR
- Working: Detects and recognizes text lines, supports multiple languages
- Fine-tuning: YES, but pretrained models work very well on modern UI fonts
- Train Strategy: Label a dataset of text-region + transcriptions → retrain using PaddleOCR pipeline
- Use Case Fit: Best overall OCR for UI text; flexible and well-maintained

# B. docTR (Mindee)

- Model Type: Transformer-based document OCR
- Working: Layout-aware model that extracts structured text blocks
- Fine-tuning: YES, can be trained for domain-specific fonts or structured content
- Use Case Fit: Strong if UI has multi-line blocks or form-heavy structure

# C. TrOCR (Microsoft)

- Model Type: Vision Transformer + Seq2Seq Decoder (BERT-style)
- Working: End-to-end image-to-text transformer
- Fine-tuning: YES, supported by HuggingFace
- Use Case Fit: Good for stylized or handwritten text extraction

# 3. Vision-to-Code (Screenshot to JSX/Tailwind)

Goal: Convert an input screenshot into usable frontend code

#### A. GPT-4 Vision API

- **Type:** Proprietary OpenAl model (not trainable)
- Working: Takes screenshot and prompt (e.g. "Convert to Tailwind JSX")
- Strengths: State-of-the-art zero-shot accuracy
- Limitations: High cost, closed-source
- Use Case Fit: Best for production if budget allows

# B. BLIP-2 (Salesforce)

- Type: Open-source visual question answering and captioning model
- Working: Vision encoder + language decoder
- Trainability: YES. Fine-tune for structured image-to-markup captioning (using UI image + JSX pairs)
- Train Strategy: Collect dataset of (UI screenshot, JSX) → fine-tune decoder

• Use Case Fit: Best self-hosted alternative to GPT-4 Vision with effort

#### C. LLaVA

- Type: Visual Language Assistant (LLaMA + CLIP)
- Working: Input image and prompt → generates structured output
- Trainability: YES (HuggingFace versions available)
- Use Case Fit: Good for building custom models where you control both image and response quality

# 4. Prompt-to-Component/Text-to-Code Generation

**Goal:** User types "Create login form" → Code is generated

#### A. GPT-4 Chat API

- **Type:** Proprietary LLM
- Trainability: No. But can be improved via prompt engineering
- Prompt Tuning Strategy: Use function-calling and system prompts to enforce consistent JSX structure
- Use Case Fit: Best zero-shot generation of React components with Tailwind

# B. CodeLlama (Meta)

- Type: Open-source LLM for code
- **Trainability:** YES. Fine-tune on your own JSX/CSS projects
- Training Strategy: Provide (prompt, component code) as supervised data to the decoder
- Use Case Fit: Full control for academic/demo systems; ideal for FYP and local inference

# C. Vercel v0 (backend)

- Type: Service + open-source templates
- Trainability: No direct training

 Integration: Extract underlying prompt templates and replicate in CodeLlama for deeper control

# 5. Full Training Pipeline for Screenshot-to-Code

#### Step-by-Step:

- 1. **Collect Data:** Gather 2000–10,000 screenshots of UIs (can use open-source themes or auto-capture Figma frames)
- 2. **Label Components:** Using Roboflow or Label Studio, mark bounding boxes and component types (button, nav, input)
- 3. Extract Text Labels: Apply PaddleOCR or label manually
- 4. **Generate Code Targets:** For each screenshot, manually code the UI (or semi-automatically using GPT-4 and verify)
- 5. Train Detection Model: Fine-tune YOLOv8 or Detectron2 on visual segments
- 6. **Train Vision-to-Code Model:** Fine-tune BLIP-2 or LLaVA using screenshot → JSX pairs
- 7. **Train Prompt-to-Code Model:** Fine-tune CodeLlama with (prompt → component) pairs
- 8. **Evaluation:** Use BLEU or Tree Edit Distance to evaluate code accuracy; visually compare with screenshot

# 6. Deployment & Cost Strategy

- **Prototype Mode:** Use GPT-4 API + PaddleOCR + react-beautiful-dnd
- Academic Mode: Fine-tune YOLOv8 + CodeLlama + BLIP-2 (self-hosted)
- Production Mode: Hybrid: GPT-4 for vision, CodeLlama for text prompts,
  PaddleOCR for OCR