

# TEAM PIXEL-PAC

## REAL TIME MONOCULAR DEPTH ESTIMATION ON EDGE AI DEVICE

### *VIVEK CHOUDHRY:*

- Trained various hybrid models based on references from Adabins and DepthAnythingV2
- Depth Estimation Model : CNN based model with MULTI HEAD ATTENTION trained with KNOWLEDGE DISTILATION
- CI/CD Deployment : GITHUB AND DOCKER

### *SHWETANK SHEKHAR:*

- Successfully built, configured, and deployed a custom 6.15.0-rc7 PREEMPT RT kernel specifically for the Raspberry Pi 5 platform.
- Configured advanced kernel parameters for real-time performance, including NO\_HZ\_FULL, a 1000 Hz timer, the performance CPU governor, and PPS/NTP support features.
- Architected a complete inference pipeline that integrates pi camera input, GPIO-based visual and audio alerts, and live visual feedback on an HDMI display in Console Log Environment.

### *RATISH R.A:*

- Curated and preprocessed a 50GB KITTI subset with ground truth depth for scalable training.
- Built a multi-process, memory-efficient pipeline for real-time knowledge distillation.
- Developed a PyTorch–TensorFlow bridge for seamless cross-framework training.
- Exported the model for fast inference and visualization on new data.
- Designed and integrated a lightweight MobileNet-based encoder-decoder with a 64-bin AdaBins module for accurate and efficient monocular depth estimation.

### *AMRIT S S:*

- 3D CAD Design for Hardware Mounting. Ran fitment simulations to validate heat clearance.
- Performed Air circulation trajectories for best cooling.
- Compared multiple RTOS options, helped finalize and validate the PREEMPT RT Kernel.

# SOFTWARE

- Model deployed using Tensorflow-Lite
- Developed lightweight version of Adabins with an inference time on raspi 5 of 650ms.
- CNN based architecture with Multi head self attention in deep layers for inter-channel information relay
- Linear Regression to map the raw depth map to aligned depth map
- The inference time on raspberry pi for *tflite* model was 115ms/10.45FPS
- Model trained for 300 epochs using knowledge distillation for ~12000 images,
- Using silog loss and Learning rate scheduler results in the least score
- Pipeline for CI/CD deployment and unit testing with Github and Dockers

# HARDWARE

Model: Raspberry Pi 5 Model B Rev 1.1

- System on Chip (SoC): Broadcom BCM2712
- CPU: 2.4GHz quad-core 64-bit Arm Cortex-A76
- RAM: 8GB LPDDR4X-4267 SDRAM
- Storage: 64GB SanDisk microSD card
- Cooling: Active Cooler
- Camera: Raspberry Pi Camera Module 3
- Primary Display: HDMI Display
- Active Warning System: GPIO based Visual(LED) & Auditory(buzzer) warning system

## Operating System & Kernel Details

- Operating System: Raspberry Pi OS (64-bit), based on Debian Bookworm.
- Kernel Version String: 6.15.0-rc7-v8-16k-NTP+
- Kernel Type: Custom compiled with PREEMPT\_RT (Full Real-Time Preemption).
- Architecture: aarch64 (64-bit).
- Build Method: Natively compiled on the Raspberry Pi 5.
- Key Kernel Configurations :
  - Full Real-Time Preemption (CONFIG\_PREEMPT\_RT=y)
  - Timer Frequency set to 1000 Hz (CONFIG\_HZ\_1000=y)
  - Full Dynamic Ticks (CONFIG\_NO\_HZ\_FULL=y)
  - Default CPU Frequency Governor set to "performance"
  - Kernel PPS (Pulse Per Second) timing support (CONFIG\_NTP\_PPS=y)
  - PPS client support for GPIO (CONFIG\_PPS\_CLIENT\_GPIO=y)