



IJRASET

International Journal For Research in
Applied Science and Engineering Technology



INTERNATIONAL JOURNAL FOR RESEARCH

IN APPLIED SCIENCE & ENGINEERING TECHNOLOGY

Volume: 11 Issue: IV Month of publication: April 2023

DOI: <https://doi.org/10.22214/ijraset.2023.51040>

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Computer Vision for Increased Efficiency of Delivery Drones Using NVIDIA Jetson Toolkit

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Abstract: While drone delivery seemed ludicrous a few years ago, it is now a reality in many parts of the United States and throughout the world. Favourable development conditions, because of the FAA's more relaxed and transparent standards, as well as a rise in demand for autonomous last-mile delivery, are propelling drone delivery technology ahead. Consumers who have had the opportunity to experience drone delivery say they prefer the comfort, speed, and positive environmental effect of drone deliveries. Thus, it is very crucial to implement computer vision to increase the efficiency and safety of these delivery drones – and we believe that Jetson toolkit is a fantastic employable solution.

Drones now have the processing power for improved object detection and tracking thanks to advances in computer vision and remote sensing. As a result, structured datasets can be used to teach deep learning-based unmanned aerial vehicles to detect objects while in surveillance mode. DNN-based drones have an important function: 3D object detection. After acquiring relevant data into the app, the camera with a trained computer vision model may scan an object utilising 3D object detection. This enables businesses to detect possible risks or things of concern in real time.

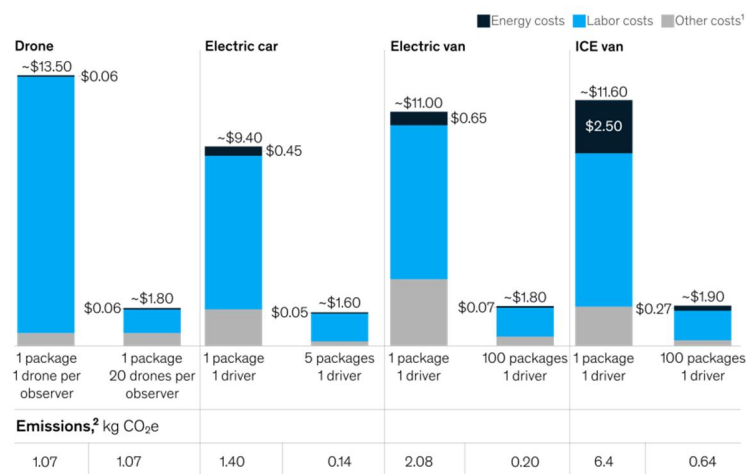
Keywords: Object Detection, DNN-based, Collision Avoidance, Delivery drones, NVIDIA Jetson.

I. INTRODUCTION

Increasing number of multinational companies such as Amazon, FedEx, Boeing, and Zipline, Walmart, etc., are using - or at least have started using last-mile delivery drones to reduce dependence on the traditional transportation systems, which have adverse effect on the environment – much larger than those of the drone delivery systems. Increasing fuel prices, and advancement in the speed and precision in the delivery, in near future ought to make UAVs relatively cost effective. Current cost analysis of UAVs versus traditional transportation is given by McKinsey and Co. [1] in the following graph:

Drones could become cost competitive with other transport modes.

Example breakdown: unit delivery costs and emissions for a five-mile delivery of a 216-cubic inch package (six inches per side)



¹Other costs include asset, maintenance, and insurance costs.
²Scope 2 and Scope 3.

C. Navigation Management System using Jetson Card

GPS can be a useful inclusion in a drone's navigational toolkit, but despite (pretty much) blanket coverage, there are areas where the technology just doesn't cut it. Researchers from GPU maker Nvidia are currently working on a navigation system that relies on visual recognition and computer learning to make sure drones don't get lost in the woods.^[4]

Rather of designing a flying robot from scratch, Nvidia's engineers used a commercially available drone. At the core of the navigation system is the company's Jetson TX1 machine learning module, which is fed visual input from two cameras. It was initially developed to go over woodland paths on rescue operations, such as checking for fallen hikers or storm damage, while still in the experimental stage of development. However, the team believes that the low-flying drone might be used anyplace GPS would be less than dependable or unavailable, such as canyons, dense metropolitan landscapes, or stock checks at a warehouse. The technology may potentially be modified to look for broken wires underwater. The drone's navigation system is also capable of avoiding obstacles and has been trained to follow train tracks. It has also been fitted in a wheeled robot for zipping around building halls. However, the major project testing ground has been woodland pathways, which can be more difficult to navigate than a more predictable urban context with consistent points of reference such as mailboxes and curb heights. In comparison, wooded environments provide little consistency - fluctuating light, a lack of markers, and trees of diverse shapes and sizes all contribute to pushing a camera-based navigation system to its limitations. The researchers claim the device has already completed the longest and most steady GPS-free flight to yet, flying autonomously for a kilometre (0.6 mi) through the centre of a woodland route while avoiding obstructions. The flying bot learned how to fly in the forest by viewing video collected by three GoPro cameras linked to a rig as Smolyanskiy went across eight kilometres of trails in the Pacific Northwest. TrailNet, the team's neural network, was also given video of trails in the Swiss Alps filmed by Zurich's Istituto Dalle Molle di Studi sull'Intelligenza Artificiale. The project's long-term objective is to be able to programme coordinates into a bot's camera-based navigation system and then have it navigate on its own. In the short term, software that robot builders may download and use to programme their own models to travel using solely visual data is being created.

III. NVIDIA JETSON

The navigation management system achieved by NVIDIA's Jetson TX1 embedded-AI card was truly a pioneering leap in the UAVs development during late 2017. Below we will discuss the latest benchmarks achieved by the latest Jetson hardware.

A. Hardware comparison of Jetson Modules

| Hardware feature \ Jetson module | Jetson TX2/TX2i | Jetson NX Xavier | Jetson AGX Xavier | Jetson AGX Orin | Hardware feature \ Jetson module |
|----------------------------------|--|---|---|---|----------------------------------|
| CPU (ARM) | 4-core ARM Cortex-A57 @ 2 GHz, 2-core Denver2 @ 2 GHz | 6-core ARM Carmel v8.2 | 8-core ARM Carmel v.8.2 @ 2.26 GHz | 12-core ARM Cortex-A78AE | CPU (ARM) |
| GPU | 256-core Pascal @ 1.3 GHz | 384-core Volta | 512-core Volta @ 1.37 GHz | 2048-core Ampere | GPU |
| Memory | 8 GB 128-bit LPDDR4, 58.3 GB/s | 16 GB 128-bit LPDDR4, 51.2GB/s | 16 GB 256-bit LPDDR4, 137 GB/s | 64 GB 256-bit LPDDR5, 205 GB/s | Memory |
| Storage | 32 GB eMMC | 16 GB eMMC | 32 GB eMMC | 64 GB eMMC | Storage |
| Tensor cores | -- | 48 | 64 | 64 | Tensor cores |
| Video encoding | 1x 4K60 (H.265) 3x 4K30 (H.265) 4x 1080p60 (H.265) | 2x 4K30 (H.265) 6x 1080p60 (H.265) | 4x 4K60 (H.265) 16x 1080p60 (H.265) 32x 1080p30 (H.265) | 2x 4K60, 4x 4K30, 8x 1080p60, 16x 1080p30 (H.265) H.264, AV1 | Video encoding |
| Video decoding | 2x 4K60 (H.265) 7x 1080p60 (H.265) 14x 1080p30 (H.265) | 2x 4K60 (H.265) 12x 1080p60 (H.265) 16x 1080p30 (H.265) | 2x 8K30 (H.265) 6x 4K60 (H.265) 26x 1080p60 (H.265) 72x 1080p30 (H.265) | 1x 8K30, 3x 4K60, 7x 4K30, 11x 1080p60, 22x 1080p30 (H.265) H.264, VP9, AV1 | Video decoding |
| PCI-Express lanes | 5 lanes PCIe Gen 2 | 1 x1 (PCIe Gen3)+ 1 x4 (PCIe Gen4) | 16 lanes PCIe Gen 4 | Up to 2x8, 1x4, 2x1 PCIe Gen 4 | PCI-Express lanes |

The above dataset is given by 'Fast Compression'^[6]

B. Jetson AGX Xavier and Jetson Xavier NX ML Performance Results for Comp Vision

| Model | Jetson Xavier NX (TensorRT) | Jetson AGX Xavier 32GB (TensorRT) |
|--------------------------------|-----------------------------|-----------------------------------|
| Image Classification ResNet-50 | 1245.10 | 2039.11 |
| Object Detection SSD-small | 1786.91 | 2833.59 |
| Object Detection SSD-Large | 36.97 | 55.16 |

The above test results were derived by us using the Jetpack v5.

IV. CONCLUSIONS AND THOUGHTS

Machine learning algorithms are taught by feeding them labelled picture data in order to recognise unlabelled input data. Such models are known as image classification models. Companies can use these models to determine which class an image belongs to. Keyword categorization and picture search are two examples of image classification applications. Drones equipped with deep neural networks (DNN) that have been trained with image classification models can assist organisations in identifying persons or objects in photographs provided to them. The photos may then be analysed and the results shared in real time. Drones now have the processing capability for improved object recognition and tracking thanks to advances in computer vision and remote sensing. As a result, structured datasets may be used to teach deep learning-based unmanned aerial vehicles to detect objects while in surveillance mode. DNN-based drones have an important function: 3D object detection. After downloading relevant data into the app, the camera with a trained computer vision model may scan an item utilising 3D object detection. This enables businesses to spot possible risks or things of concern in real time.

Jetson is used to deploy popular DNN models and ML frameworks to the edge for applications including as real-time classification and object identification, posture estimation, semantic segmentation, and natural language processing (NLP). Jetson results on the test cases ran by both the authors of the paper, i.e., Image Classification ResNet-50, Object Detection SSD-small, and Object Detection SSD-large, provide the feedback of top performance in these spaces than any other card in the industry. The above three mentioned tests were performed on Jetson Xavier NX and Jetson AGX Xavier 32GB cards, respectively. We truly believe that the implementation of these cards is one of the best options for product developers/enthusiasts in this space.

V. ACKNOWLEDGMENT

I (Ameya Shukla) would like to sincerely acknowledge my co-author Mr. Akul Athreye, for willingly accepting my offer of collaborating on this paper. I would also like to show sincere gratitude to both of our institutions which are Birla Open Minds International School, Hyderabad, and BITS Pilani Hyderabad Campus.

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