NRI: ROBOTICS, SCIENCE AND TECHNOLOGY FOR FORESTRY (USDA/NIFA \#584401)

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## LARGE-SCALE MAPPING AND ODOMETRY



Semantic mapping


Over-canopy reconstruction


LiDAR odometry ( $\sim 0.01 \%$ drift)


## OUR VISION AND RESEARCH GOALS

| Sampling | Coverage Rate | Resolution | Labor |
| :--- | :--- | :--- | :--- |
| Over the canopy measurements | Fast | Low | Minimial |
| Under the canopy UAVs | Moderate | Medium | Modest |
| Ground measurements | Very slow, but necessary | High | High |

- Large-scale mapping using multiple UAVs
- Fine-grained semantic understanding of unstructured environments
- Pairing human-collected ground measurements with UAV data


## CURIOUS BUT BUSY BEE EXPLORER

Take controls that deteriorate the minimal sufficient representation; when you cannot do so you have learnt the scene

## PLATFORMS



## FALCON 450

~35 min flight time, LiDAR, stereo vision, flexible compute (NUC, or Xavier NX) SENSOR TOWER
LiDAR, RGB global shutter camera, RTK GPS, thermal camera; onboard compute for data acquisition and compression

MAP THE PENN CAMPUS

## FORESTRY ESTIMATES

Would like to accurately measuring the DBH (diameter at breast height), main stem taper profile, and height of the trees.


Measurement error with respect to ground truth obtained from Virginia Tech

1. Home-brew tool for labeling point clouds, train a semantic segmentation network from range images

2. Estimate control points and diameter in real time, accumulate profile estimates, re-cluster for correcting height



Catalogue the biodiversity and the carbon captured. The 3D map will be matched with Penn's existing database.
$\operatorname{argmin} \underset{\substack{p\left(x\left|\left.\right|^{t}, u^{t}\right), p(z \mid x), p(u \mid x)\right.}}{\mathbb{E}}\left[I_{g}\left(y_{t+1}(u) ; z\right)-I(z ; x)\right]$

## KEY CHALLENGES

Planning with very tight tolerances


Extreme contrast changes


