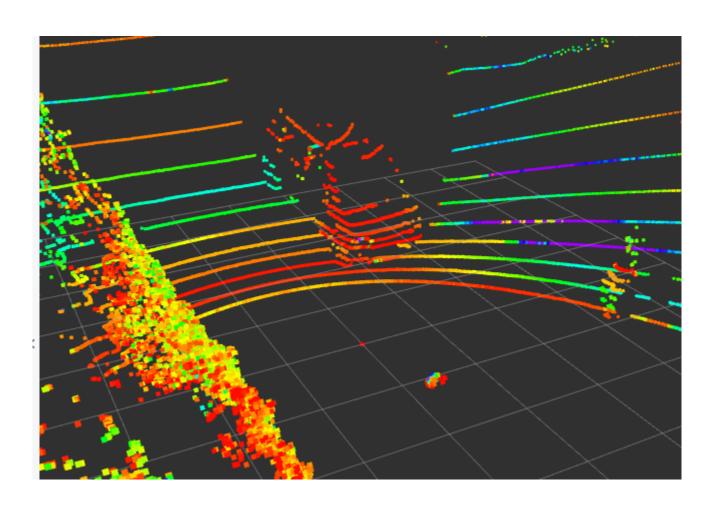
# Sensor fusion – GPS & LIDAR

# Setup



Aim: Get best possible estimate of distance b/w vehicles

# Setup



#### Sensor information and variables

$$f(x) = Distance(A_{GPS}, B_{GPS})$$

$$f(x) = \frac{1}{\sqrt{2\pi}\sigma_f} e^{-\frac{(x-\mu_f)^2}{2\sigma_f^2}}$$

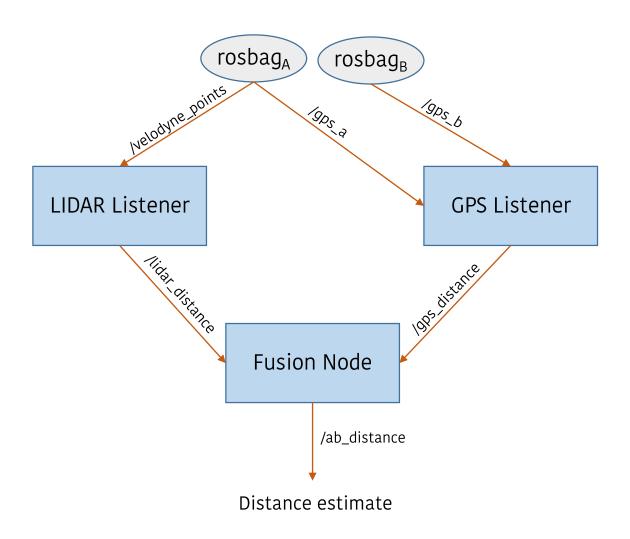
$$g(x) = A_{LIDAR\_READING\_OF\_B}$$

$$g(x) = \frac{1}{\sqrt{2\pi}\sigma_g} e^{-\frac{(x-\mu_g)^2}{2\sigma_g^2}}$$

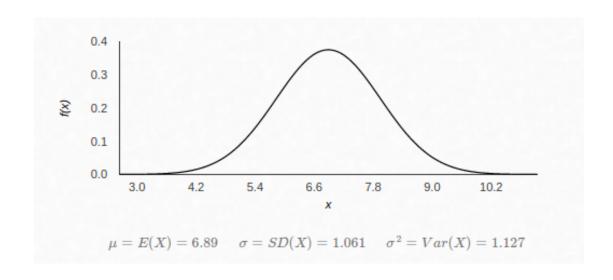
Best optimal estimate is a scaled Gaussian PDF obtained from the product of f(x) and g(x) with standard deviation and mean:

$$\sigma_{fg} = \sqrt{rac{\sigma_f^2 \sigma_g^2}{\sigma_f^2 + \sigma_g^2}} \quad ext{and} \quad \mu_{fg} = rac{\mu_f \sigma_g^2 + \mu_g \sigma_f^2}{\sigma_f^2 + \sigma_g^2}$$

## ROS Package

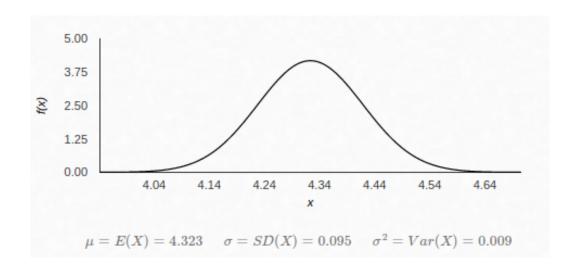


## Results - GPS



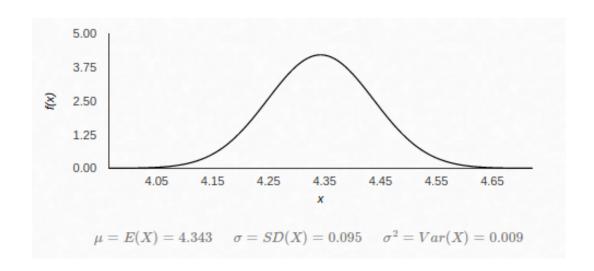
Mean = 6.89 Standard deviation = 1.061

### Results - LIDAR



Mean = 4.323 Standard deviation = 0.095

#### Results – Fused information



Mean = 4.343 Standard deviation = 0.0946

## Code and presentation

All code has been open-sourced at Github.

This presentation can be found online <u>here</u>.

#### References

- [1] Computing the distance between two locations on Earth from coordinates, John D Cook
- [2] Products and Convolutions of Gaussian Probability Density Functions, P.A. Bromiley
- [3] How a Kalman filter works, Tim Babb
- [4] Understanding the Basis of the Kalman Filter via a Simple and Intuitive Derivation, Ramsey Faragher