

ANALYZIS OF EGNOS IONOSPHERIC MODEL

Introduction to Integrity

Integrity – the aim of the EGNOS analysis

„Integrity is the measure of the trust that can be placed in the correctness of the information supplied by a navigation system. Integrity includes the ability of the system to provide timely warnings to users when the system should not be used for navigation”
- *This definition was adapted from the 2008 US Federal Navigation Plan - Navipedia*

Introduction

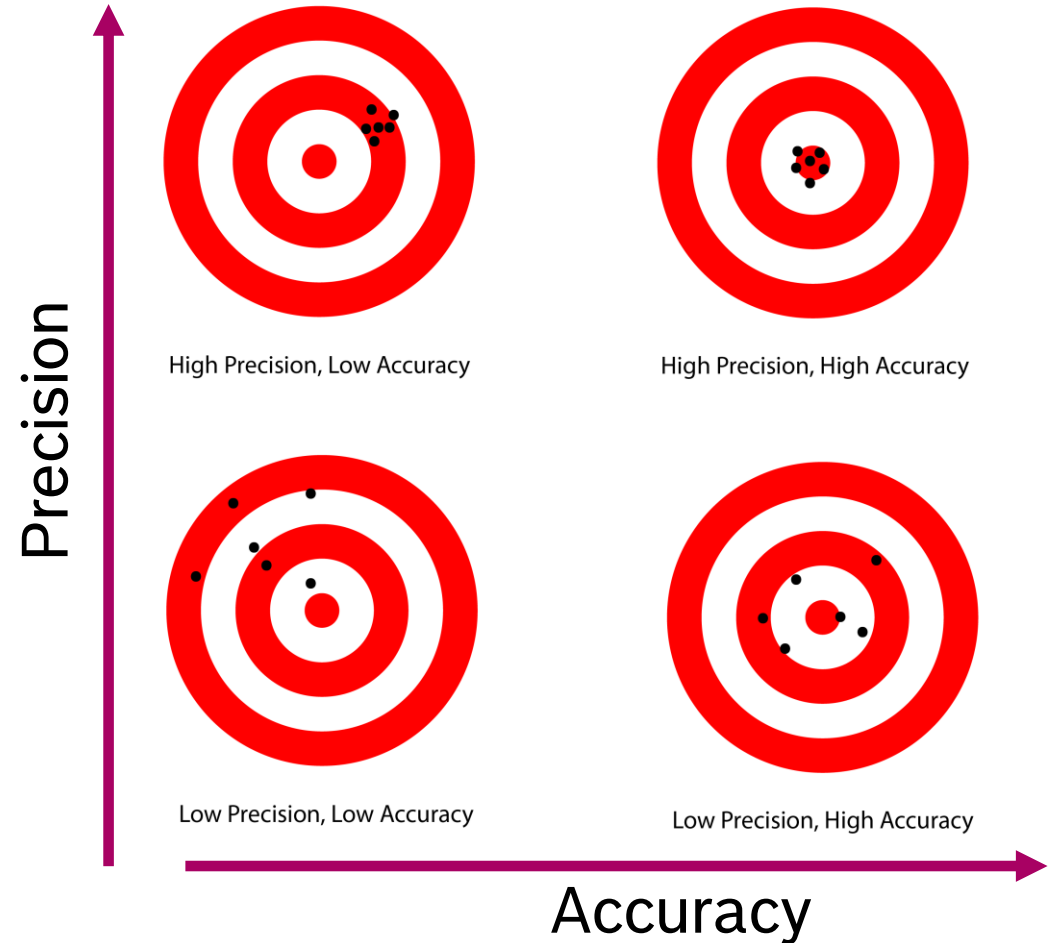
Definition of performance key parameter

Accuracy: Measure of navigation output **deviating from truth.**

Integrity: Ability of a system to **provide timely warnings** when the system should not be used for navigation. Integrity risk is the probability of an undetected, threatening navigation system problem.

Availability: Fraction of time navigation system is **usable as determined by compliance with accuracy, integrity and continuity requirements.**

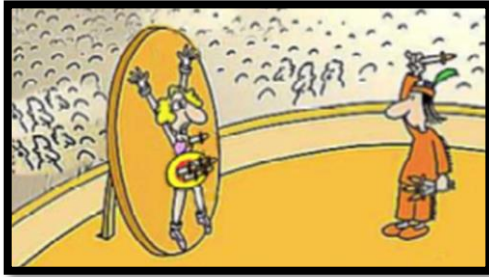
Continuity: Likelihood that the navigation **supports accuracy and integrity requirements for the duration of intended operation.** Continuity risk is the probability of a detected but unscheduled navigation interruption after initiation of an operation.



Introduction

System performance

- System performance:
Accuracy, Integrity, Availability, Continuity



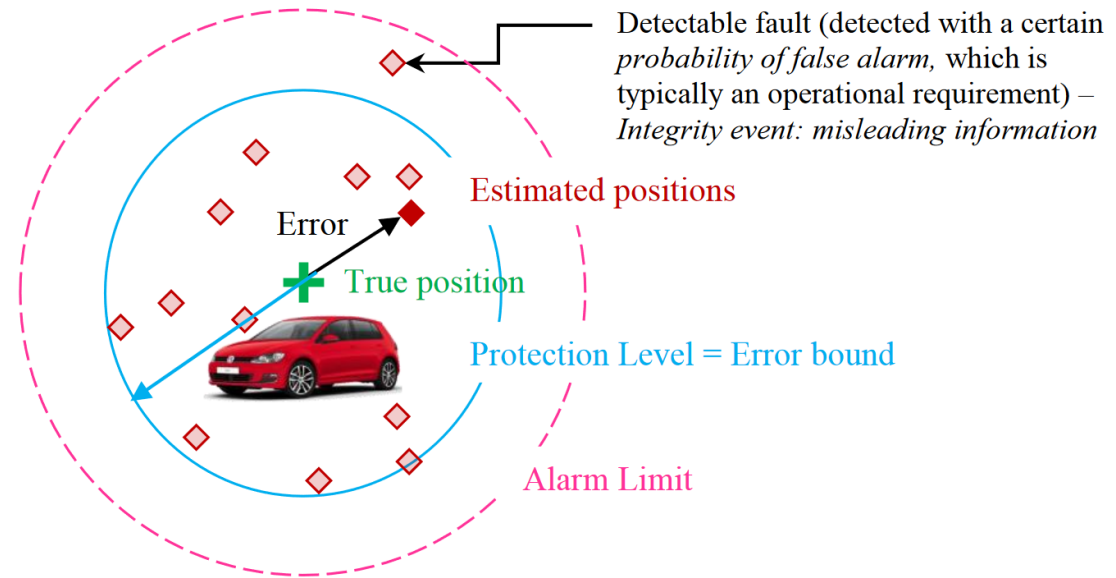
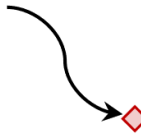
High accuracy, poor integrity

Integrity monitoring/mechanisms



Poor accuracy, high integrity

Detectable fault, but
hazardous condition –
Integrity event: hazardous
misleading information

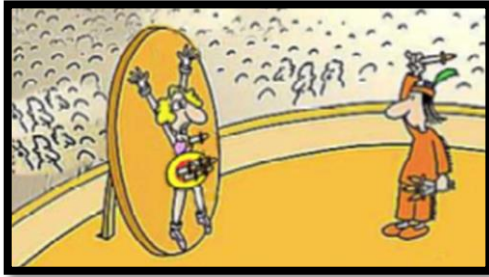


The concepts of Protection Level, Alarm Limit and integrity events.

Introduction

Stanford diagram

- System performance:
Accuracy, Integrity, Availability, Continuity

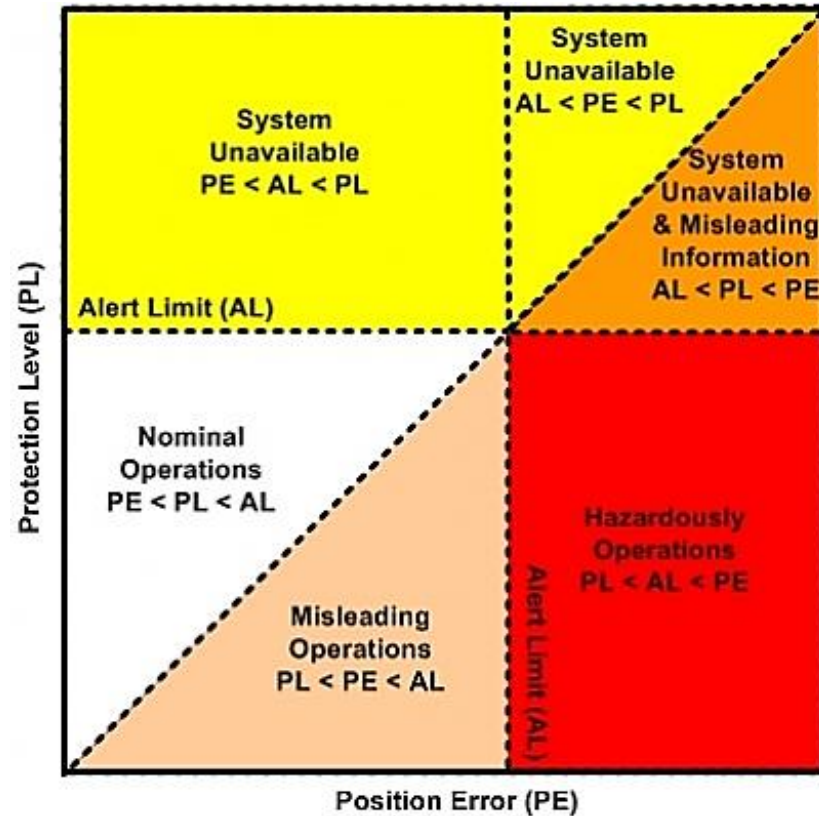


High accuracy, poor integrity

Integrity monitoring/mechanisms



Poor accuracy, high integrity



AL: Alert Limit
PL: Protection Level
PE: Position Estimation
HMI: Hazardously
Misleading Information

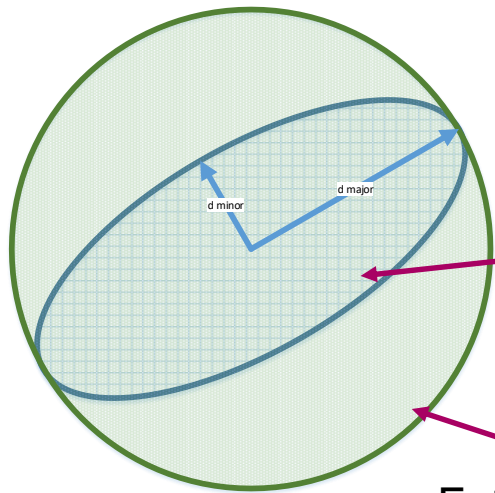
EGNOS

RTCA's protection level calculation

Horizontal protection level

$$HPL = K_H d_{major}$$

$K_H = 6.0$ – position probability 10^{-7}



Ellipsoid of projection matrix

Estimation of protection level

Variance of the residual after application of ionospheric error

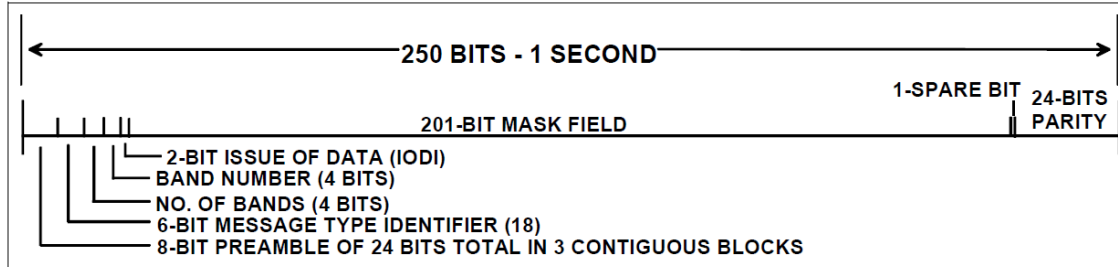
$$\sigma_i^2 = \sigma_{i,flt}^2 + \sigma_{i,UIRE}^2 + \sigma_{i,air}^2 + \sigma_{i,tropo}^2$$

Inverse cumulative function of 2D Gaussian

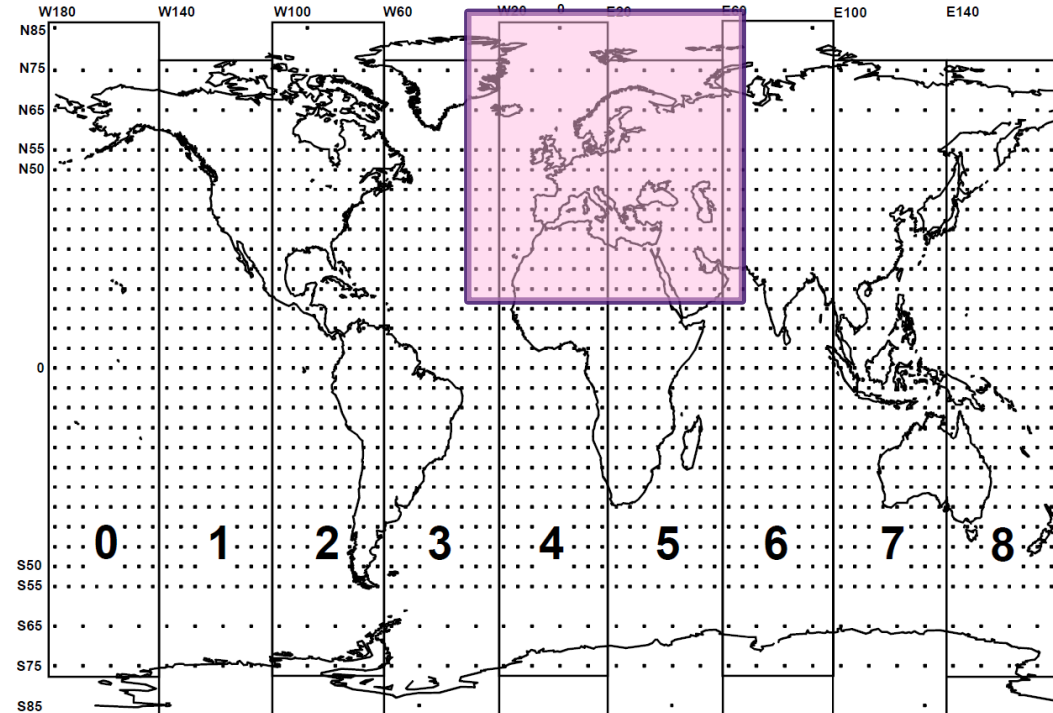
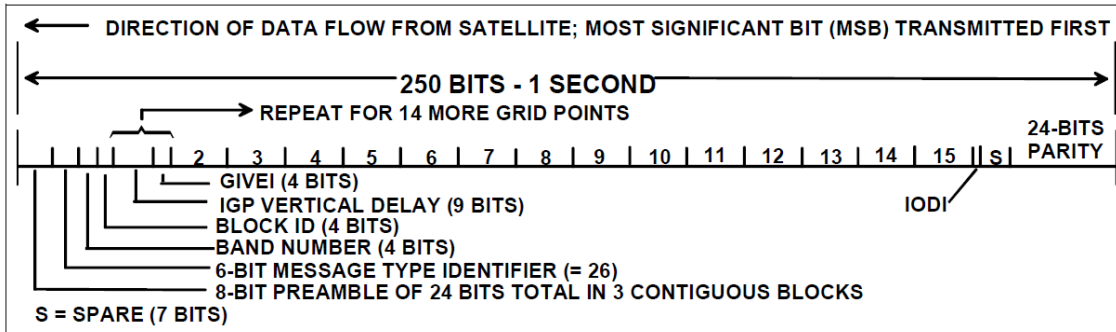
$$r = F^{-1}(p) = \sqrt{-2 \ln(1 - p)}$$

Ionospheric Grid Point and Message Formats

IGP Mask Message Format



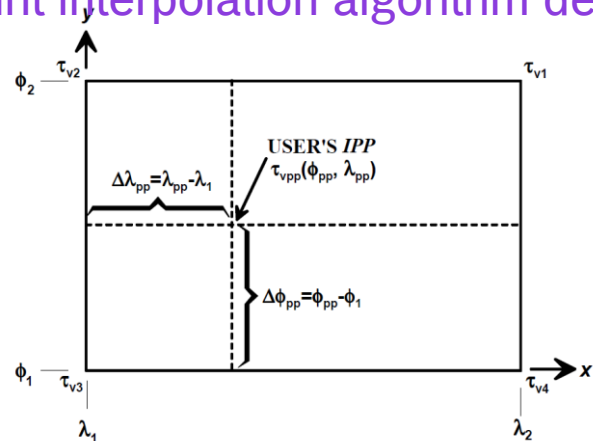
Ionospheric Delay Corrections Message Format



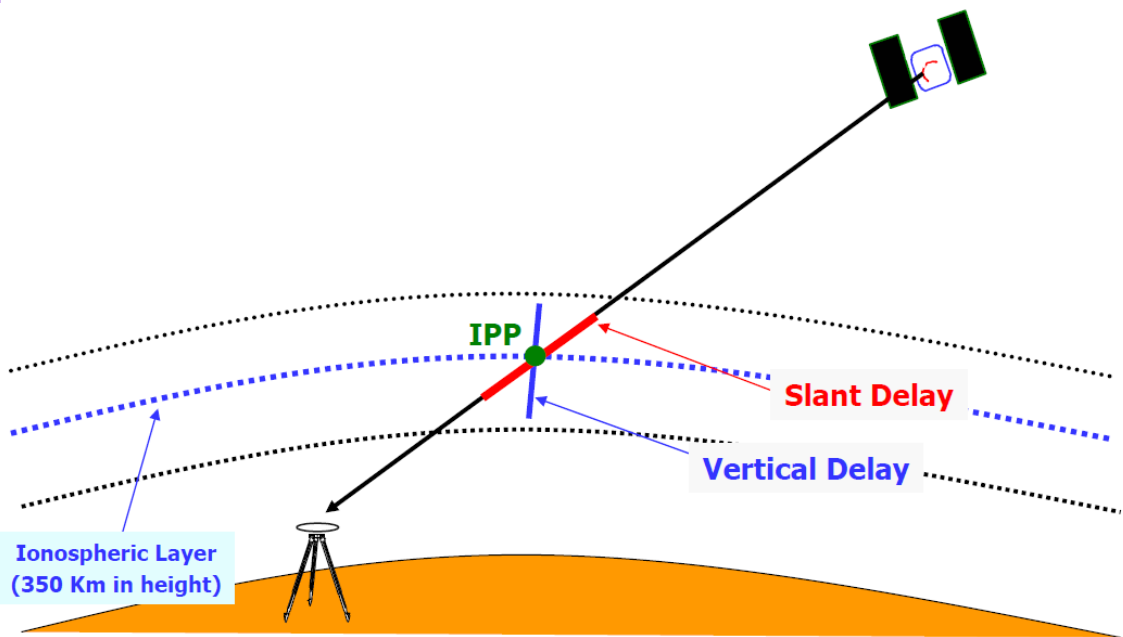
EGNOS

Ionospheric Variance

4 point interpolation algorithm definition



IONOSPHERIC PIERCE
POINTS (IPP)



GINA – Global Integrated Navigation Algorithm

EGNOS

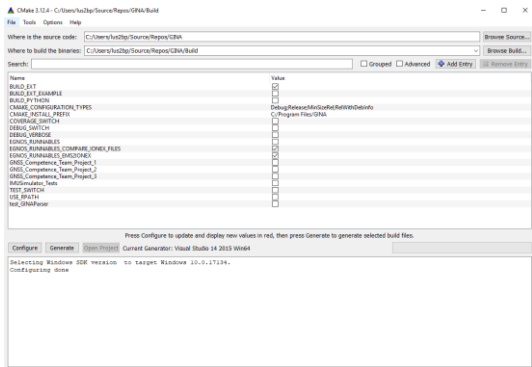
Cmake based C++ project

[Github Repo](#)

Evaulation done by Octova/Matlab

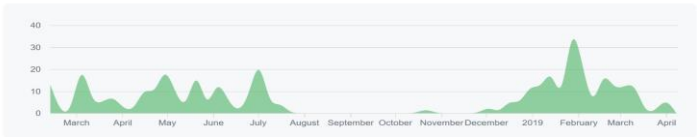
EGNOS is the latest module in GINA

It contains ~4000 Line of Code



Feb 18, 2018 – Apr 15, 2019

Contributions to master, excluding merge commits



GINA
GINAProject

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Popular repositories

GINA
Global Integrated Navigation Algorithm
C++ ★ 6 ♀ 1

162 contributions in the last year



Contribution activity

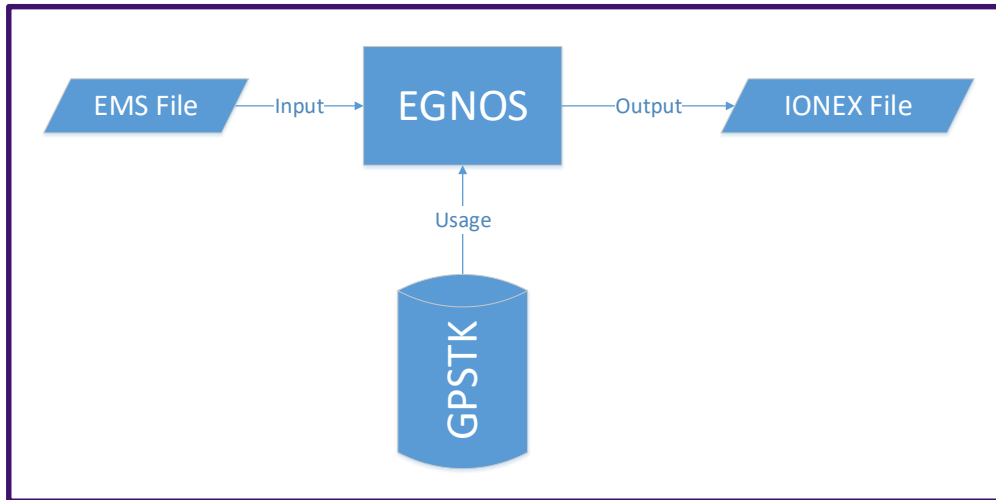
April 2019
Created 8 commits in 1 repository
GINAProject/GINA 8 commits



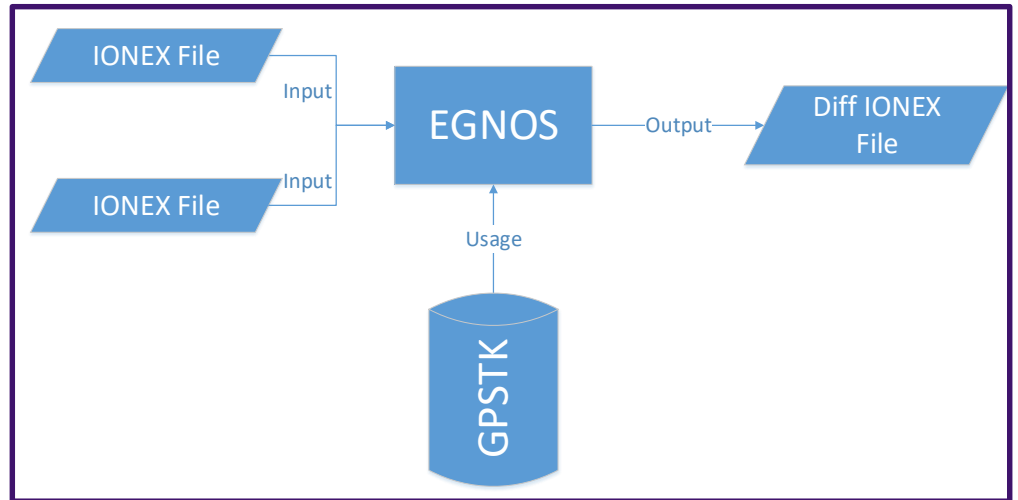
EGNOS - GINA

EGNOS Features – EMS Processing

Create Ionex file from EMS

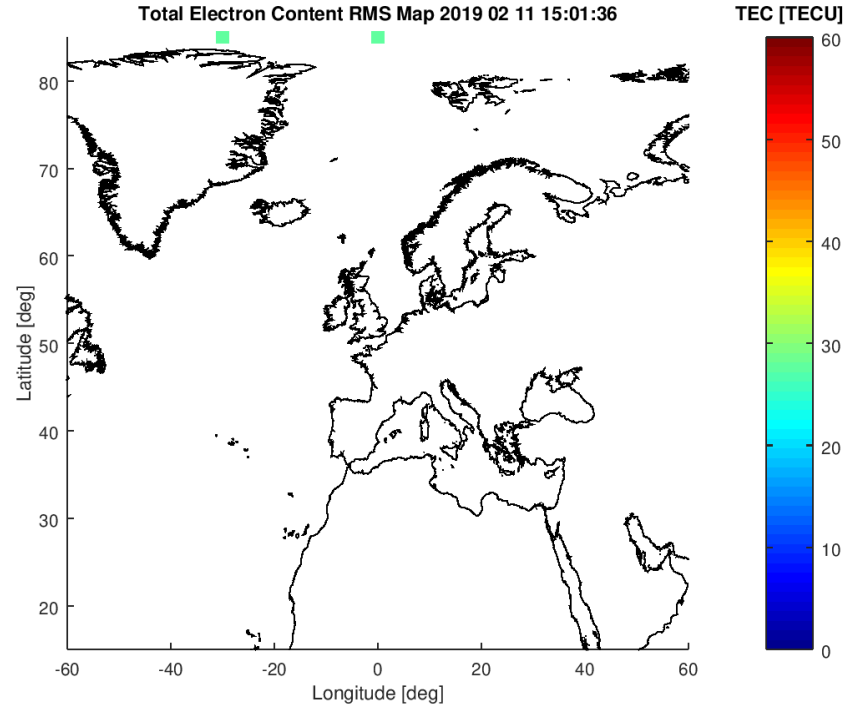
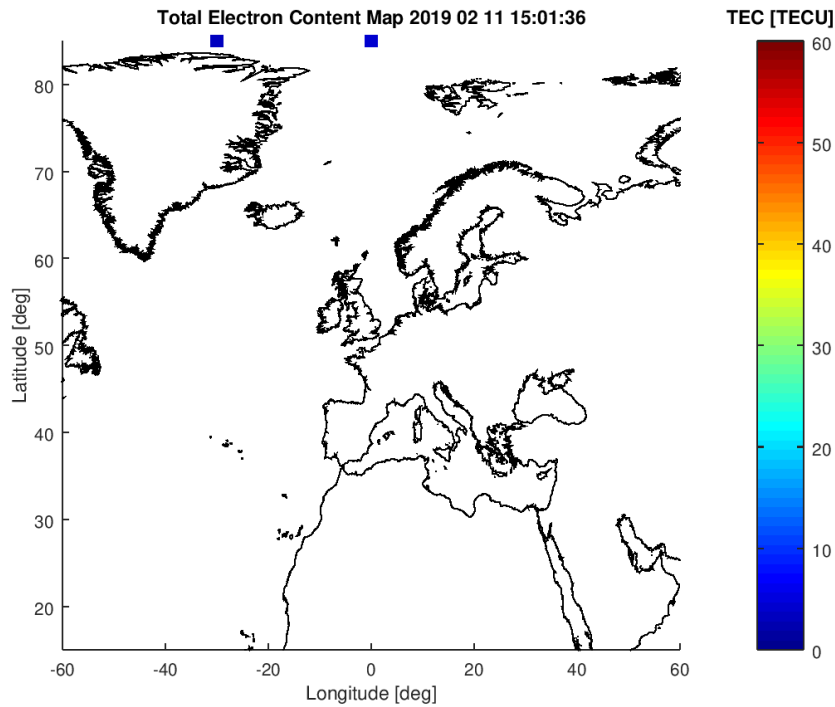


Difference of Ionex files



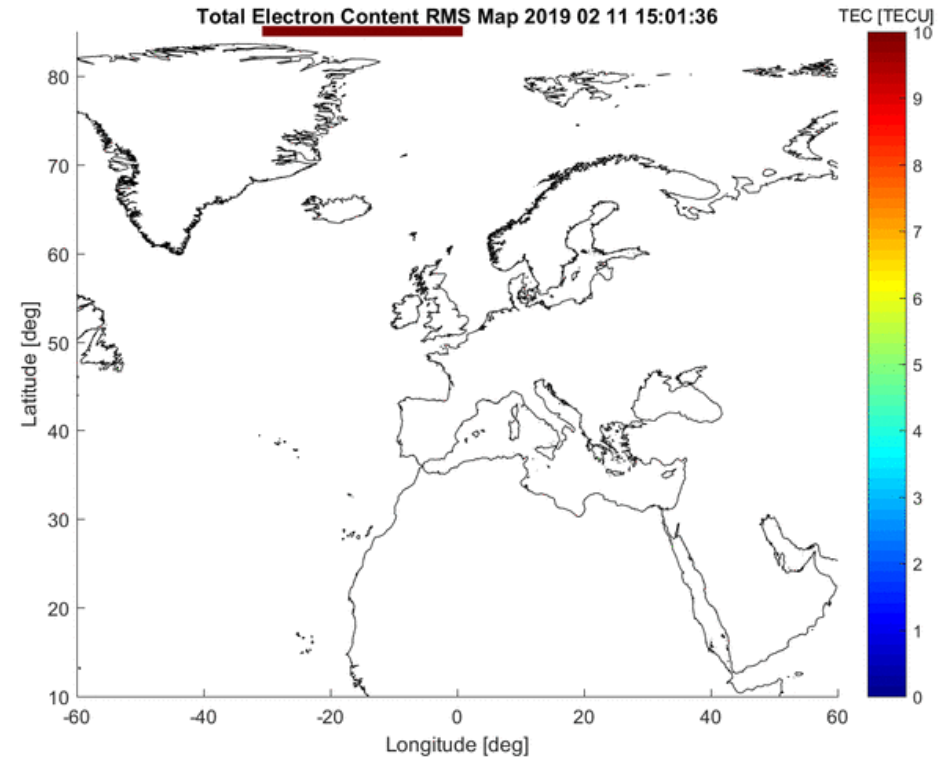
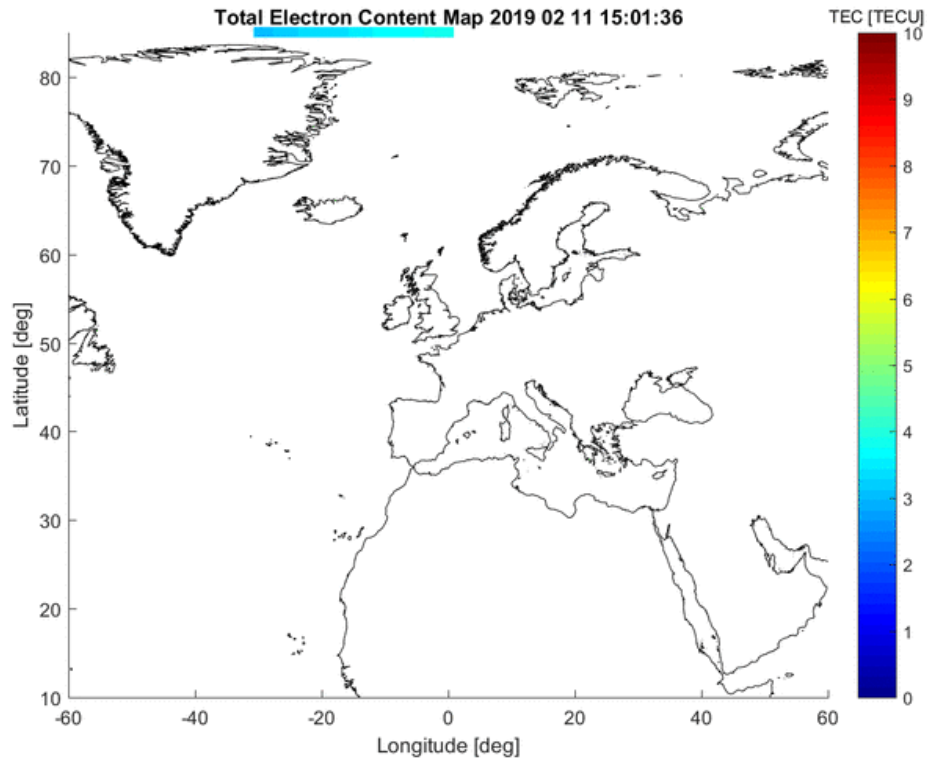
EGNOS - GINA

Raw measurement



EGNOS - GINA

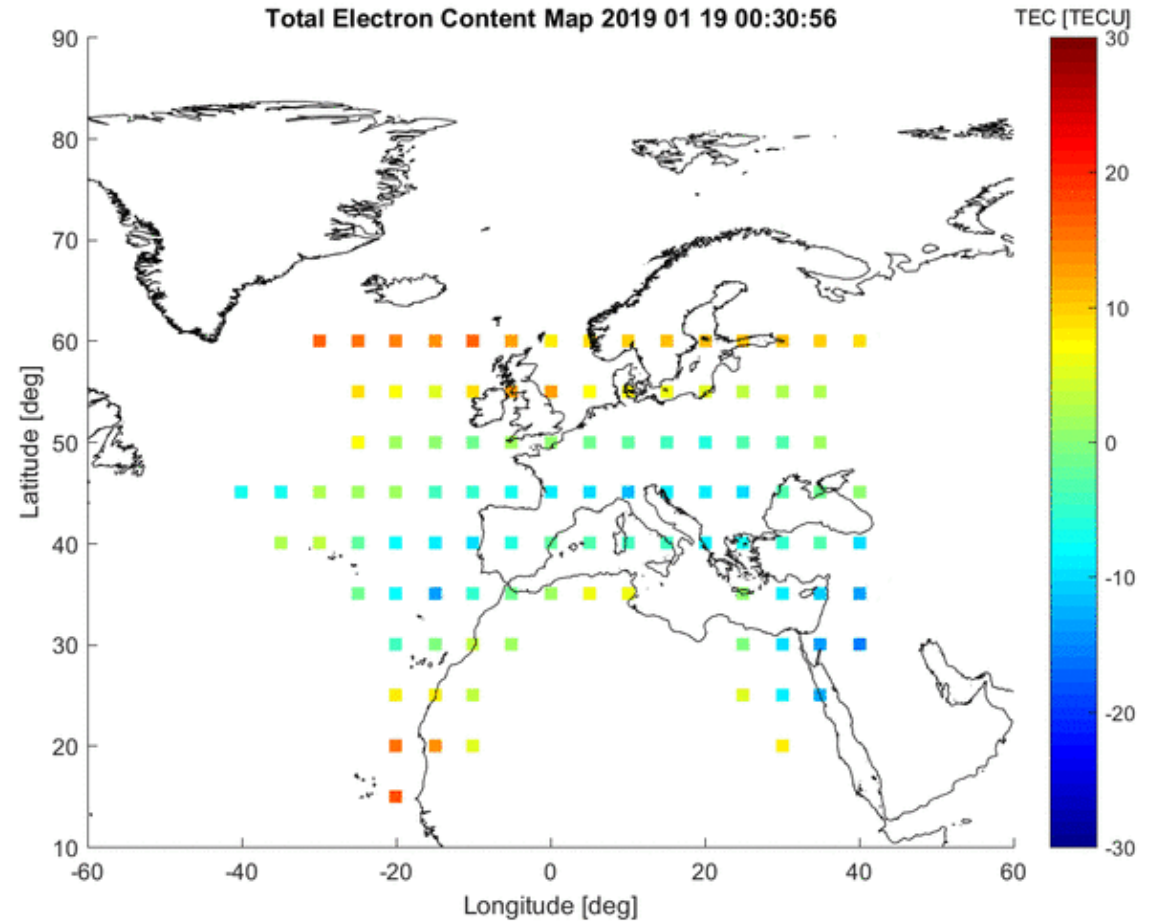
Interpolated IGPs



EGNOS - GINA

EMS – CODE's GIM

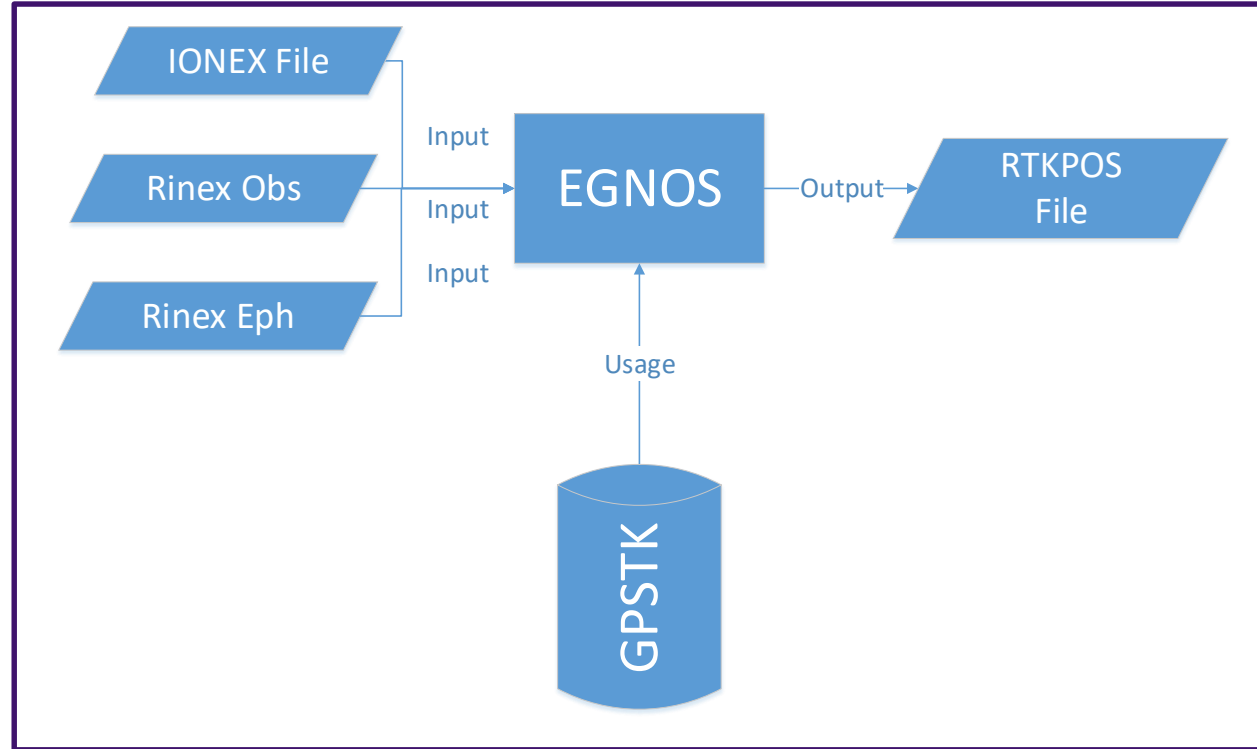
$$dTEC = TEC_1 - TEC_2$$
$$dRMS = \sqrt{RMS_1 + RMS_2}$$



EGNOS - GINA

EGNOS Features – Navigation Engine with EGNOS Ionosphere Model

Calculate position with IONEX correction

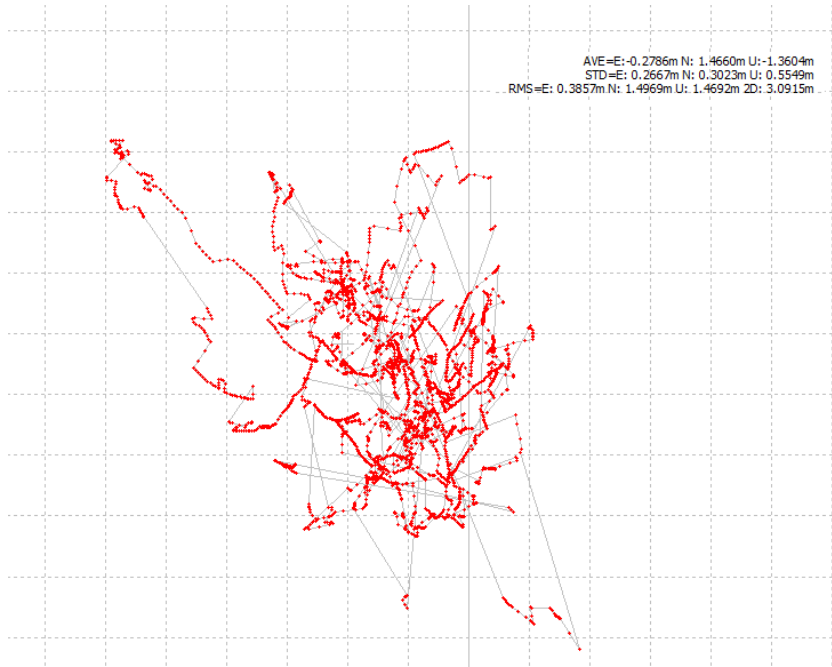


EGNOS - GINA

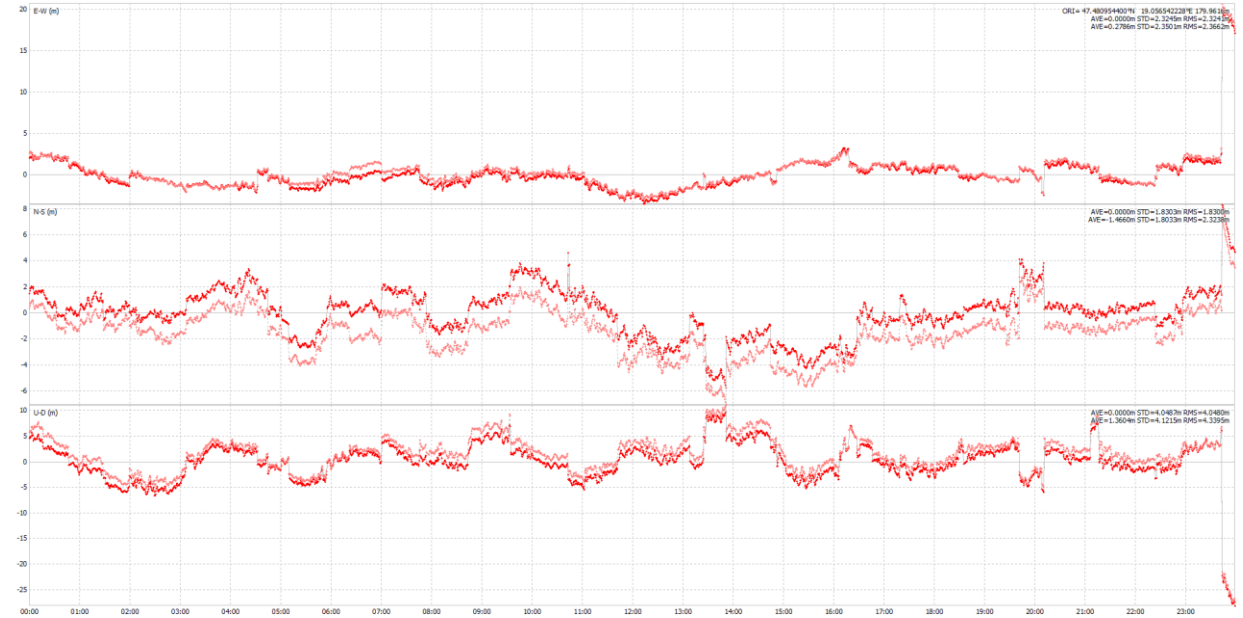
Navigation Engine with EGNOS Iono Model

Position Difference Ground Track

EMS-CODE

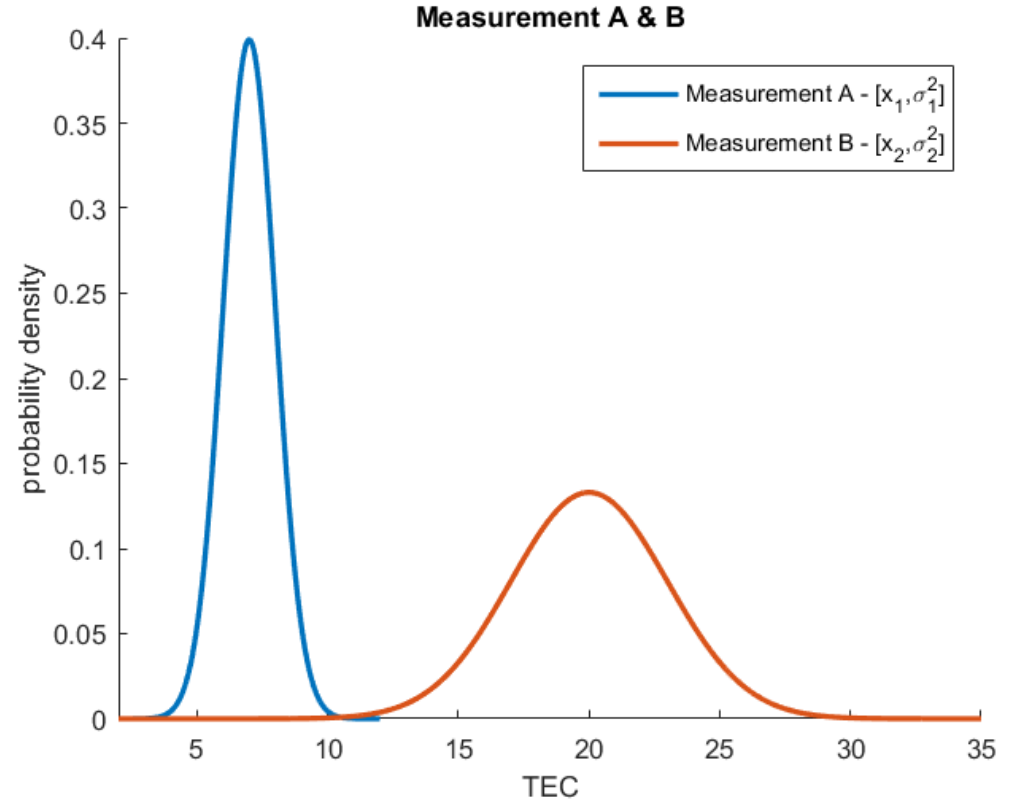
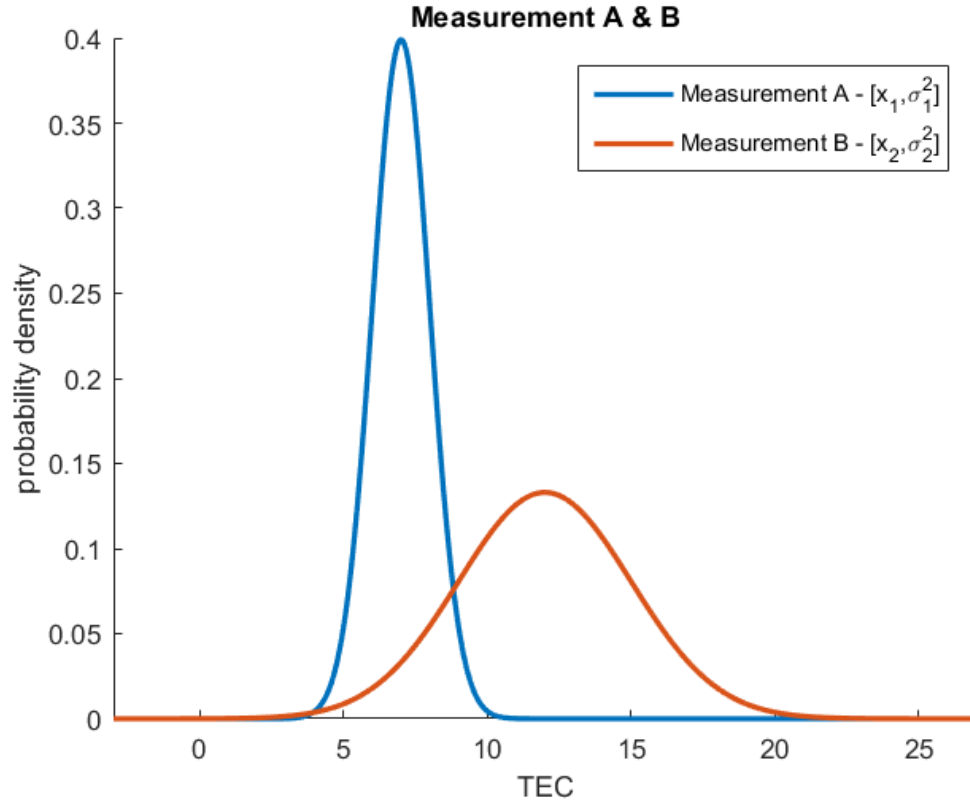


Position with EMS and with CODE



Bayesian statistic

Are the measurements consistent?

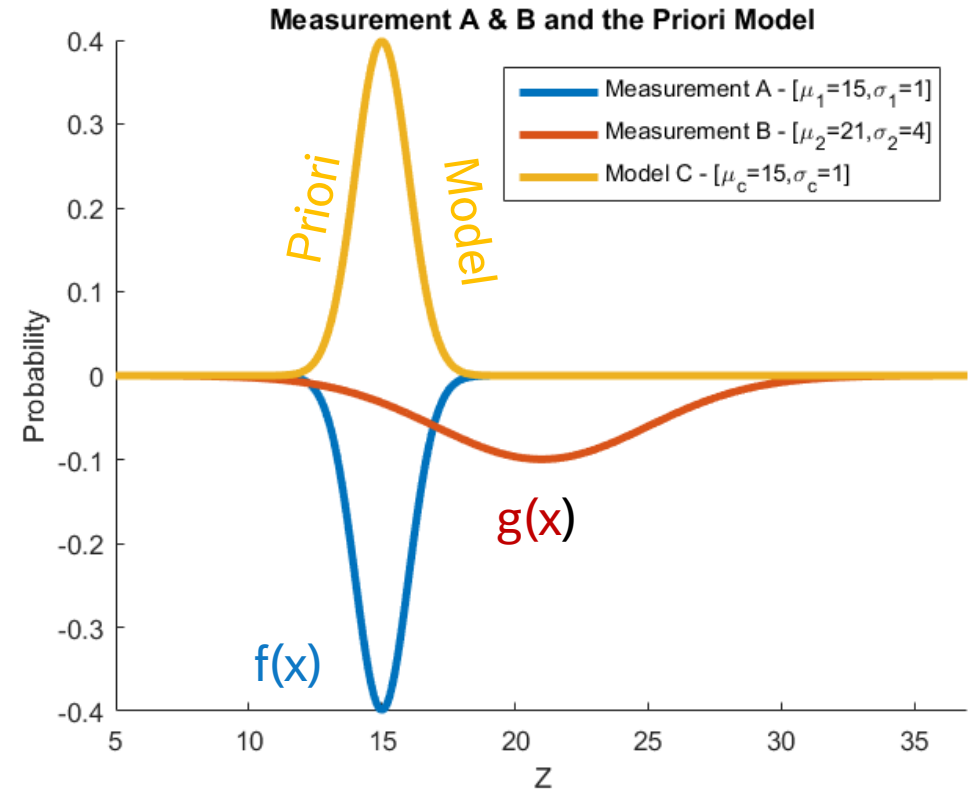


Likelihood of the measurements

Priori belief is the Measurement A

$$P(A = x_1 \text{ and } B = x_2 | \text{Priori is model C}) = ?$$

$$\begin{aligned} P(A = x_1 \text{ and } B = x_2 | \text{Priori is model A}) \\ &= P(A = x_1 | \text{Model A}) P(B = x_2 | \text{Model A}) \\ &= \sum_i P(A = x_1 | R = x_i) P(R = x_i) \sum_i P(B = x_2 | R = x_i) \end{aligned}$$

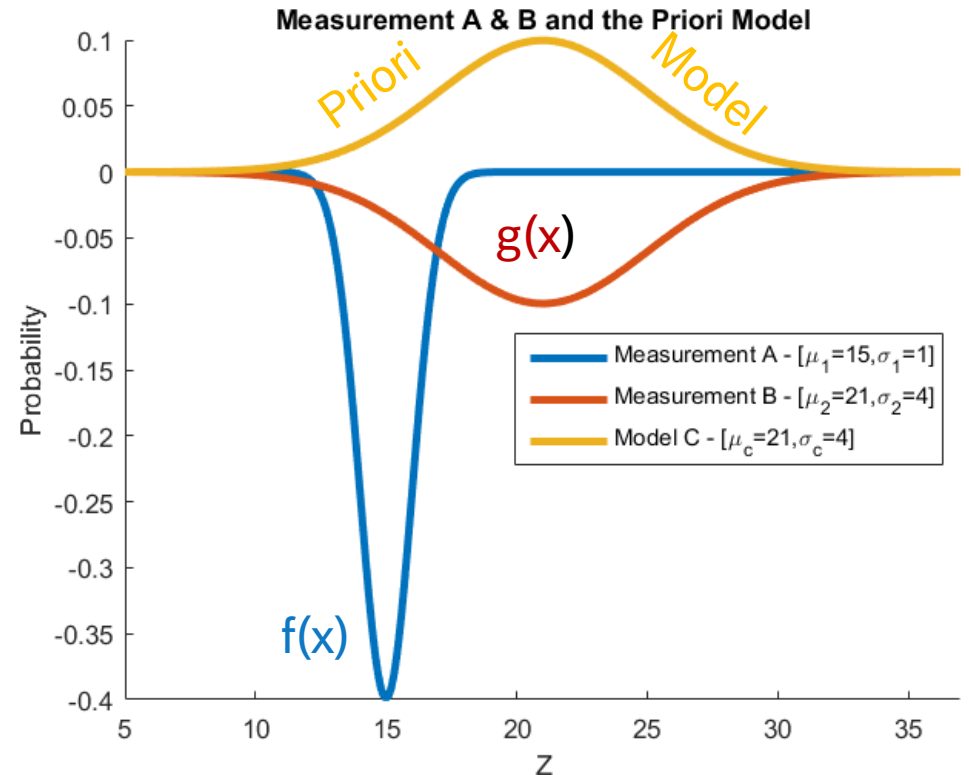


Likelihood of the measurements

Priori belief is the Measurement B

$$P(A = x_1 \text{ and } B = x_2 | \text{Priori is model C}) = ?$$

$$\begin{aligned} &P(A = x_1 \text{ and } B = x_2 | \text{Priori is model B}) \\ &= P(A = x_1 | \text{Model B})P(B = x_2 | \text{Model B}) \\ &= \sum_i P(A = x_1 | R = x_i)P(R = x_i) \sum_i P(B = x_2 | R \end{aligned}$$



Likelihood of the measurements

Which Priori Model would give the maximum likelihood?

$P(A = x_1 \text{ and } B = x_2 | \text{Priori is model C})$

Model C is a Gaussian with mean x_c and σ_c .

$$\begin{aligned} P(A = x_1 \text{ and } B = x_2 | \text{Model C}) &= P(A = x_1 | \text{Model C}) P(B = x_2 | \text{Model C}) \\ &= \sum_i P(A = x_1 | R = x_i) P(R = x_i) \sum_i P(B = x_2 | R = x_i) P(R = x_i) \\ &= \int_{-\infty}^{\infty} [x, \sigma_1]_{|x_1} [x_c, \sigma_c]_{|x} dx \int_{-\infty}^{\infty} [x, \sigma_2]_{|x_2} [x_c, \sigma_c]_{|x} dx \\ &= \int_{-\infty}^{\infty} [x_1, \sigma_1]_{|x} [x_c, \sigma_c]_{|x} dx \int_{-\infty}^{\infty} [x_2, \sigma_2]_{|x} [x_c, \sigma_c]_{|x} dx = [x_1, \sqrt{\sigma_1^2 + \sigma_c^2}]_{|x_c} [x_2, \sqrt{\sigma_2^2 + \sigma_c^2}]_{|x_c} \\ &= \left[\sqrt{\frac{(\sigma_1^2 + \sigma_c^2)(\sigma_2^2 + \sigma_c^2)}{(\sigma_1^2 + \sigma_c^2) + (\sigma_2^2 + \sigma_c^2)}}, \frac{x_2(\sigma_1^2 + \sigma_c^2) + x_1(\sigma_2^2 + \sigma_c^2)}{(\sigma_1^2 + \sigma_c^2) + (\sigma_2^2 + \sigma_c^2)} \right]_{|x_c} \end{aligned}$$

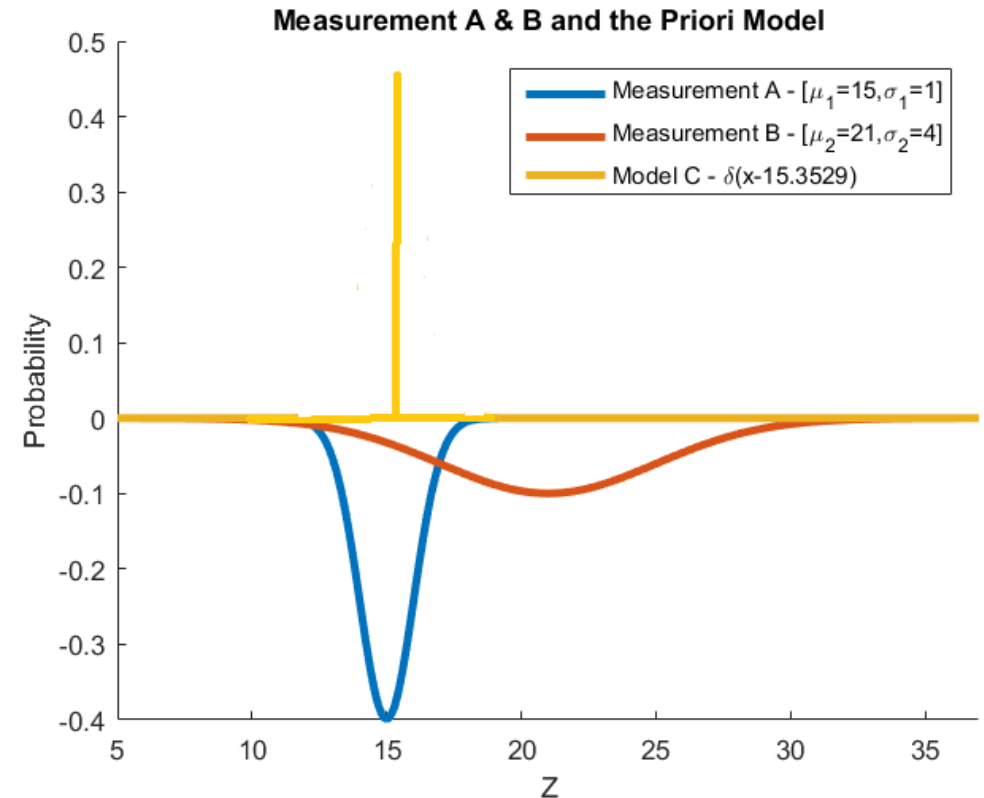
Likelihood of the measurements

Which Priori Model would give the maximum likelihood?

$$\min \left(\frac{(\sigma_1^2 + \sigma_c^2)(\sigma_2^2 + \sigma_c^2)}{(\sigma_1^2 + \sigma_c^2) + (\sigma_2^2 + \sigma_c^2)} \right) \Rightarrow \sigma_c^2 = 0$$

$$\frac{x_2(\sigma_1^2 + \sigma_c^2) + x_1(\sigma_2^2 + \sigma_c^2)}{(\sigma_1^2 + \sigma_c^2) + (\sigma_2^2 + \sigma_c^2)} = \frac{x_2(\sigma_1^2) + x_1(\sigma_2^2)}{(\sigma_1^2) + (\sigma_2^2)} = x_c$$

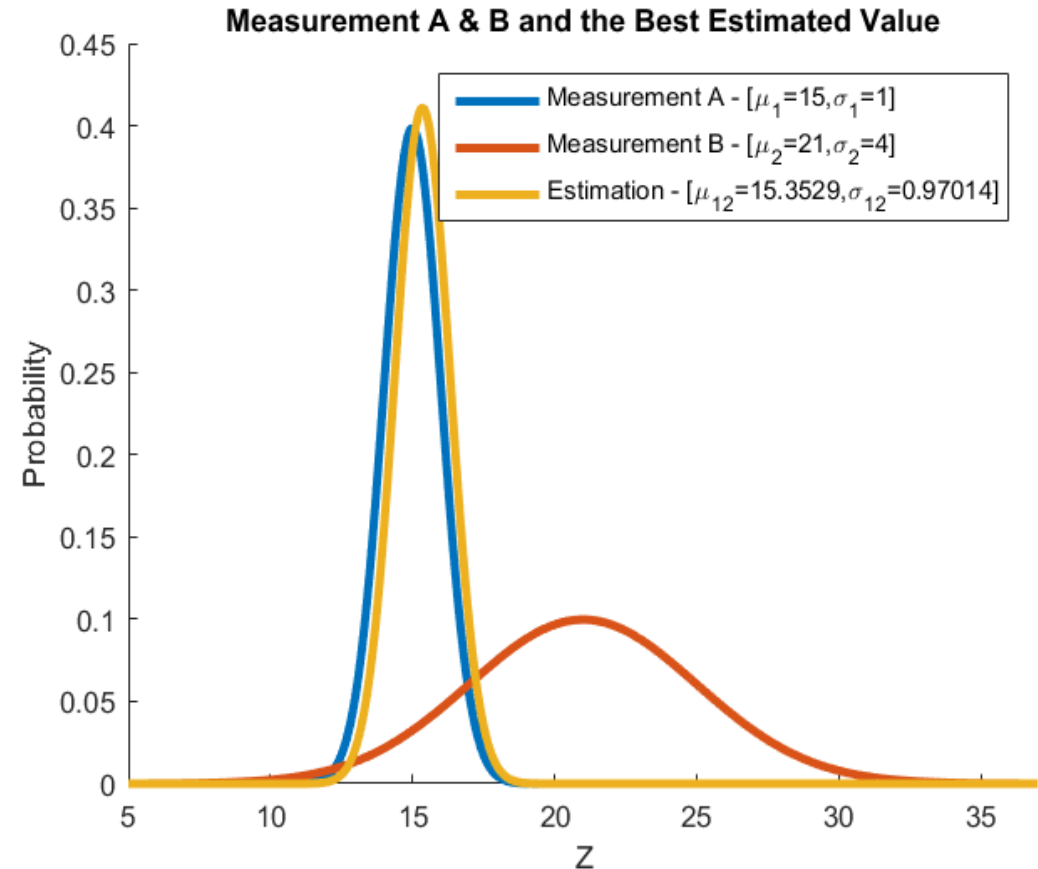
$$\lim_{\sigma \rightarrow \infty} \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1(x-\mu)^2}{2\sigma^2}} = \delta(x - \mu)$$



Maximum likelihood

But wouldn't we get the same result with a Kalman filter or LSQ?

$$\begin{aligned} &P(\text{Priori } D \text{ at } \mu | A = x_1 \text{ and } B = x_2) \\ &= \frac{[x_1, \sqrt{\sigma_1^2 + \sigma_2^2}]|_{x_2} [x_{12}, \sigma_{12}]|_{\mu}}{\int_{-\infty}^{\infty} [x_1, \sqrt{\sigma_1^2 + \sigma_2^2}]|_{x_2} [x_{12}, \sigma_{12}]|_x dx} \\ &= \frac{[x_1, \sqrt{\sigma_1^2 + \sigma_2^2}]|_{x_2} [x_{12}, \sigma_{12}]|_{\mu}}{[x_1, \sqrt{\sigma_1^2 + \sigma_2^2}]|_{x_2}} = [x_{12}, \sigma_{12}]|_{\mu} \end{aligned}$$



Next Steps Think Tank





THANK
YOU