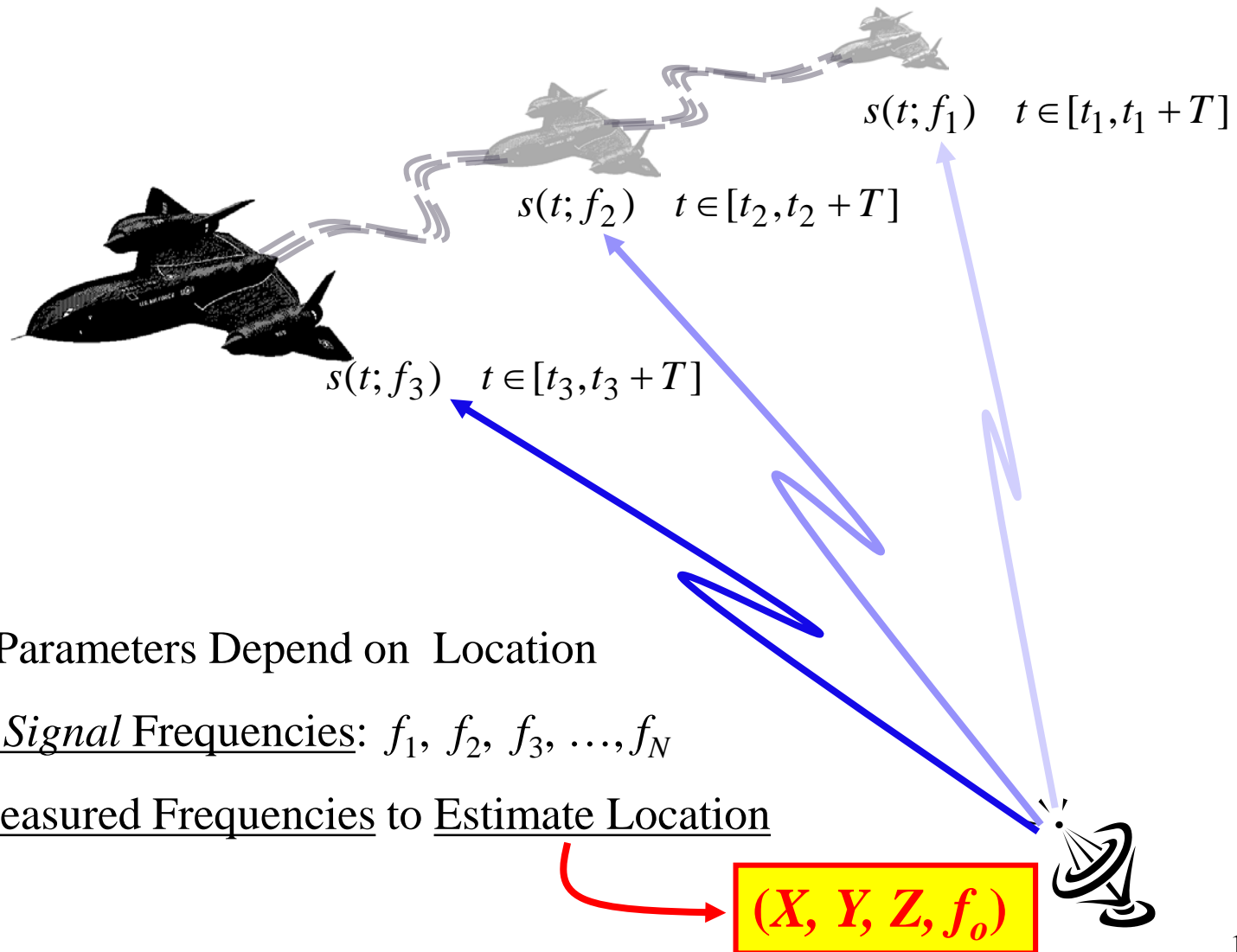


CRLB Example:

Single-Rx Emitter Location via Doppler



- Received Signal Parameters Depend on Location
 - Estimate Rx Signal Frequencies: $f_1, f_2, f_3, \dots, f_N$
 - Then Use Measured Frequencies to Estimate Location

Problem Background

Radar to be Located: at Unknown Location (X, Y, Z)

Transmits Radar Signal at Unknown Carrier Frequency f_o

Signal is intercepted by airborne receiver with:

Known (Navigation Data):

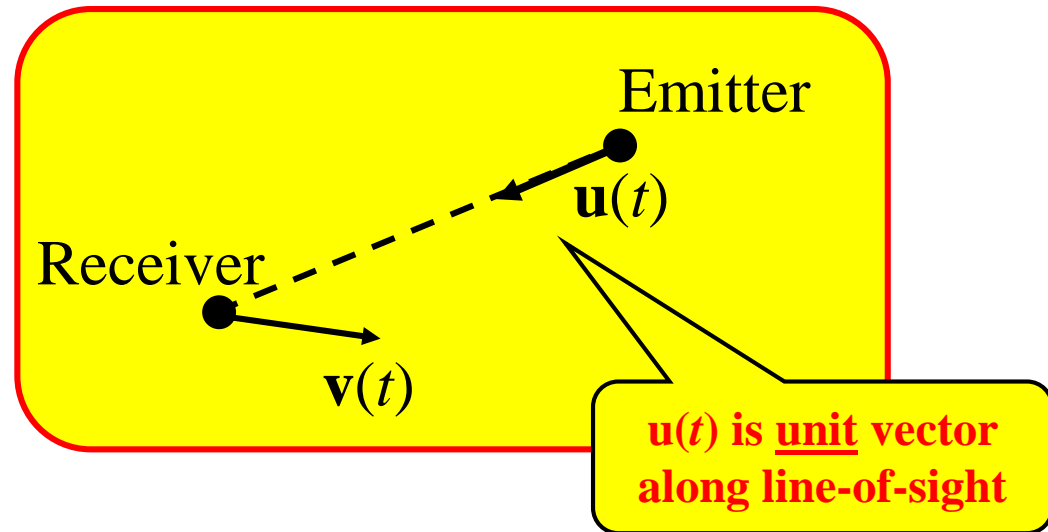
Antenna Positions: $(X_p(t), Y_p(t), Z_p(t))$

Antenna Velocities: $(V_x(t), V_y(t), V_z(t))$

Goal: Estimate Parameter Vector $\mathbf{x} = [X \ Y \ Z \ f_o]^T$

Physics of Problem

Relative motion between emitter and receiver causes a Doppler shift of the carrier frequency:



$$f(t, \mathbf{x}) = f_o - \frac{f_o}{c} \mathbf{v}(t) \bullet \mathbf{u}(t)$$

$$= f_o - \frac{f_o}{c} \left[\frac{V_x(t)(X_p(t) - X) + V_y(t)(Y_p(t) - Y) + V_z(t)(Z_p(t) - Z)}{\sqrt{(X_p(t) - X)^2 + (Y_p(t) - Y)^2 + (Z_p(t) - Z)^2}} \right].$$

Because we estimate the frequency there is an error added:

$$\tilde{f}(t_i, \mathbf{x}) = f(t_i, \mathbf{x}) + v(t_i)$$

Estimation Problem Statement

Vector-Valued function of a Vector

Given:

Data Vector: $\tilde{\mathbf{f}}(\mathbf{x}) = [\tilde{f}(t_1, \mathbf{x}) \tilde{f}(t_2, \mathbf{x}) \cdots \tilde{f}(t_N, \mathbf{x})]^T$

Navigation Info: $X_p(t_1), X_p(t_2), \cdots, X_p(t_N)$

$Y_p(t_1), Y_p(t_2), \cdots, Y_p(t_N)$

$Z_p(t_1), Z_p(t_2), \cdots, Z_p(t_N)$

$V_x(t_1), V_x(t_2), \cdots, V_x(t_N)$

$V_y(t_1), V_y(t_2), \cdots, V_y(t_N)$

$V_z(t_1), V_z(t_2), \cdots, V_z(t_N)$

Estimate:

Parameter Vector: $[X \ Y \ Z \ f_o]^T$

Right now only want to consider the CRLB

The CRLB

Note that this is a “signal” plus noise scenario:

- The “signal” is the noise-free frequency values
- The “noise” is the error made in measuring frequency

Assume zero-mean Gaussian noise with covariance matrix \mathbf{C} :

- Can use the “General Gaussian Case” of the CRLB
- Of course validity of this depends on how closely the errors of the frequency estimator really do follow this

Our data vector is distributed according to: $\tilde{\mathbf{f}}(\mathbf{x}) \sim \mathcal{N}(\mathbf{f}(\mathbf{x}), \mathbf{C})$

Only need the first term in the CRLB equation:

$$[\mathbf{J}]_{ij} = \left[\frac{\partial \mathbf{f}(\mathbf{x})}{\partial x_i} \right]^T \mathbf{C}^{-1} \left[\frac{\partial \mathbf{f}(\mathbf{x})}{\partial x_j} \right]$$

Only Mean Shows
Dependence on
parameter \mathbf{x} !

I use \mathbf{J} for the FIM instead of \mathbf{I} to avoid confusion with the identity matrix.

Convenient Form for FIM

Called “The Jacobian” of $\mathbf{f}(\mathbf{x})$

To put this into an easier form to look at... Define a matrix \mathbf{H} :

$$\mathbf{H} = \left. \frac{\partial}{\partial \mathbf{x}} \mathbf{f}(\mathbf{x}) \right|_{\mathbf{x}=\text{true value}} = [\mathbf{h}_1 \mid \mathbf{h}_2 \mid \mathbf{h}_3 \mid \mathbf{h}_4]$$

where

$$\mathbf{h}_j = \left[\begin{array}{c} \frac{\partial}{\partial x_j} f(t_1, \mathbf{x}) \\ \frac{\partial}{\partial x_j} f(t_2, \mathbf{x}) \\ \vdots \\ \frac{\partial}{\partial x_j} f(t_N, \mathbf{x}) \end{array} \right]_{\mathbf{x}=\text{true value}}$$

Then it is east to verify that the FIM becomes:

$$\mathbf{J} = \mathbf{H}^T \mathbf{C}^{-1} \mathbf{H}$$

CRLB Matrix

The Cramer-Rao bound covariance matrix then is:

$$\begin{aligned}\mathbf{C}_{CRB}(\mathbf{x}) &= \mathbf{J}^{-1} \\ &= \left[\mathbf{H}^T \mathbf{C}^{-1} \mathbf{H} \right]^{-1}\end{aligned}$$

A closed-form expression for the partial derivatives needed for \mathbf{H} can be computed in terms of an arbitrary set of navigation data – see “Reading Version” of these notes (on BlackBoard).

Given: Emitter Location & Platform Trajectory and Measurement Cov \mathbf{C}

- Compute Matrix \mathbf{H}
- Compute the CRLB covariance matrix $\mathbf{C}_{CRB}(\mathbf{x})$
- Compute eigen-analysis of $\mathbf{C}_{CRB}(\mathbf{x})$
- Determine the 4-D error ellipsoid.

Can't really plot a 4-D ellipsoid!!!

But... it is possible to project this 4-D ellipsoid down into a 2-D ellipse so that you can see the effect of geometry.

Projection of Error Ellipsoids

See also “Slice Of Error Ellipsoids”

A zero-mean Gaussian vector of two vectors \mathbf{x} & \mathbf{y} : $\boldsymbol{\theta} = \begin{bmatrix} \mathbf{x}^T & \mathbf{y}^T \end{bmatrix}^T$

The the PDF is:

$$p(\boldsymbol{\theta}) = \frac{1}{(2\pi)^{N/2} \sqrt{\det(\mathbf{C}_{\boldsymbol{\theta}})}} \exp\left\{-\frac{1}{2} \boldsymbol{\theta}^T \mathbf{C}_{\boldsymbol{\theta}}^{-1} \boldsymbol{\theta}\right\} \quad \mathbf{C}_{\boldsymbol{\theta}} = \begin{bmatrix} \mathbf{C}_{\mathbf{x}} & \mathbf{C}_{\mathbf{xy}} \\ \mathbf{C}_{\mathbf{yx}} & \mathbf{C}_{\mathbf{y}} \end{bmatrix}$$

The quadratic form in the exponential defines an ellipse:

$$\boldsymbol{\theta}^T \mathbf{C}_{\boldsymbol{\theta}}^{-1} \boldsymbol{\theta} = k$$

Can choose k to make size of ellipsoid such that $\boldsymbol{\theta}$ falls inside the ellipsoid with a desired probability

Q: If we are given the covariance $\mathbf{C}_{\boldsymbol{\theta}}$ how is \mathbf{x} alone is distributed?

A: Extract the sub-matrix $\mathbf{C}_{\mathbf{x}}$ out of $\mathbf{C}_{\boldsymbol{\theta}}$

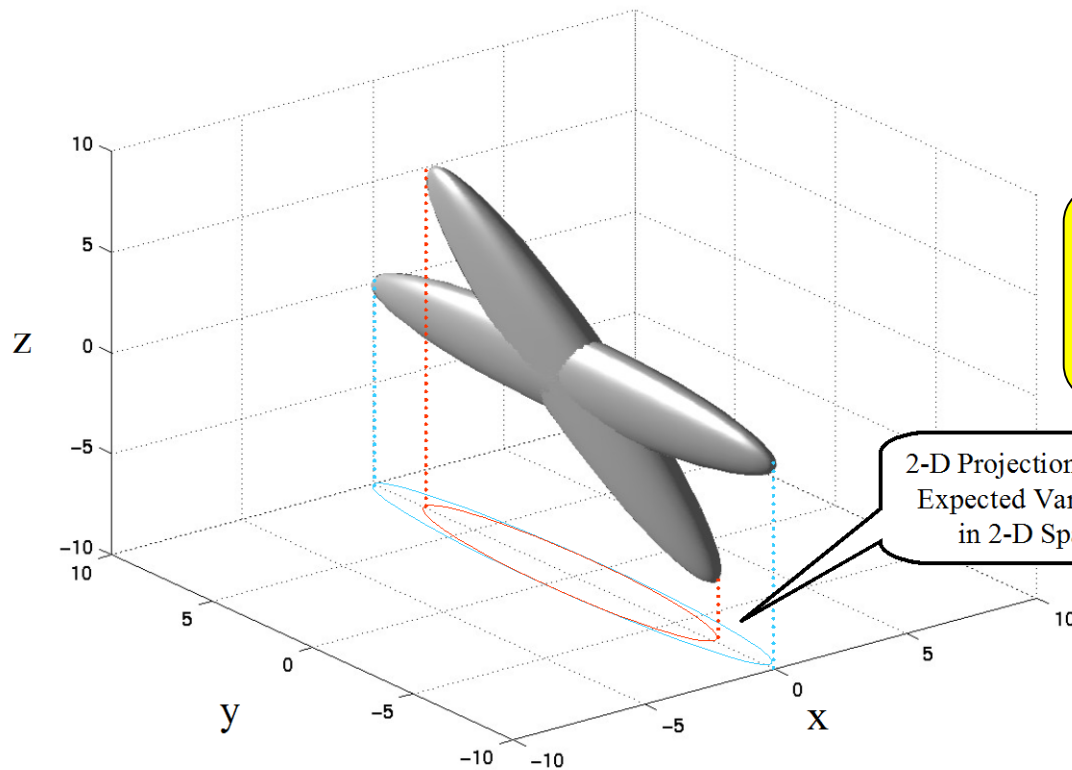
Projection Example

Full Vector: $\boldsymbol{\theta} = [x \ y \ z]^T$

Sub-Vector: $\mathbf{x} = [x \ y]^T$

We want to project the 3-D ellipsoid for $\boldsymbol{\theta}$
...down into a 2-D ellipse for \mathbf{x}

Projections of 3-D Ellipsoids onto 2-D Space



2-D ellipse still shows
the full range of
variations of x and y

2-D Projections Show
Expected Variation
in 2-D Space

Finding Projections

To find the projection of the CRLB ellipse:

1. Invert the FIM to get \mathbf{C}_{CRB}
2. Select the submatrix $\mathbf{C}_{CRB,sub}$ from \mathbf{C}_{CRB}
3. Invert $\mathbf{C}_{CRB,sub}$ to get \mathbf{J}_{proj}
4. Compute the ellipse for the quadratic form of \mathbf{J}_{proj}

Mathematically:

$$\mathbf{C}_{CRB,sub} = \mathbf{P}\mathbf{C}_{CRB}\mathbf{P}^T$$
$$= \mathbf{P}\mathbf{J}^{-1}\mathbf{P}^T$$



$$\mathbf{J}_{proj} = \left(\mathbf{P}\mathbf{J}^{-1}\mathbf{P}^T\right)^{-1}$$

\mathbf{P} is a matrix formed from the identity matrix:

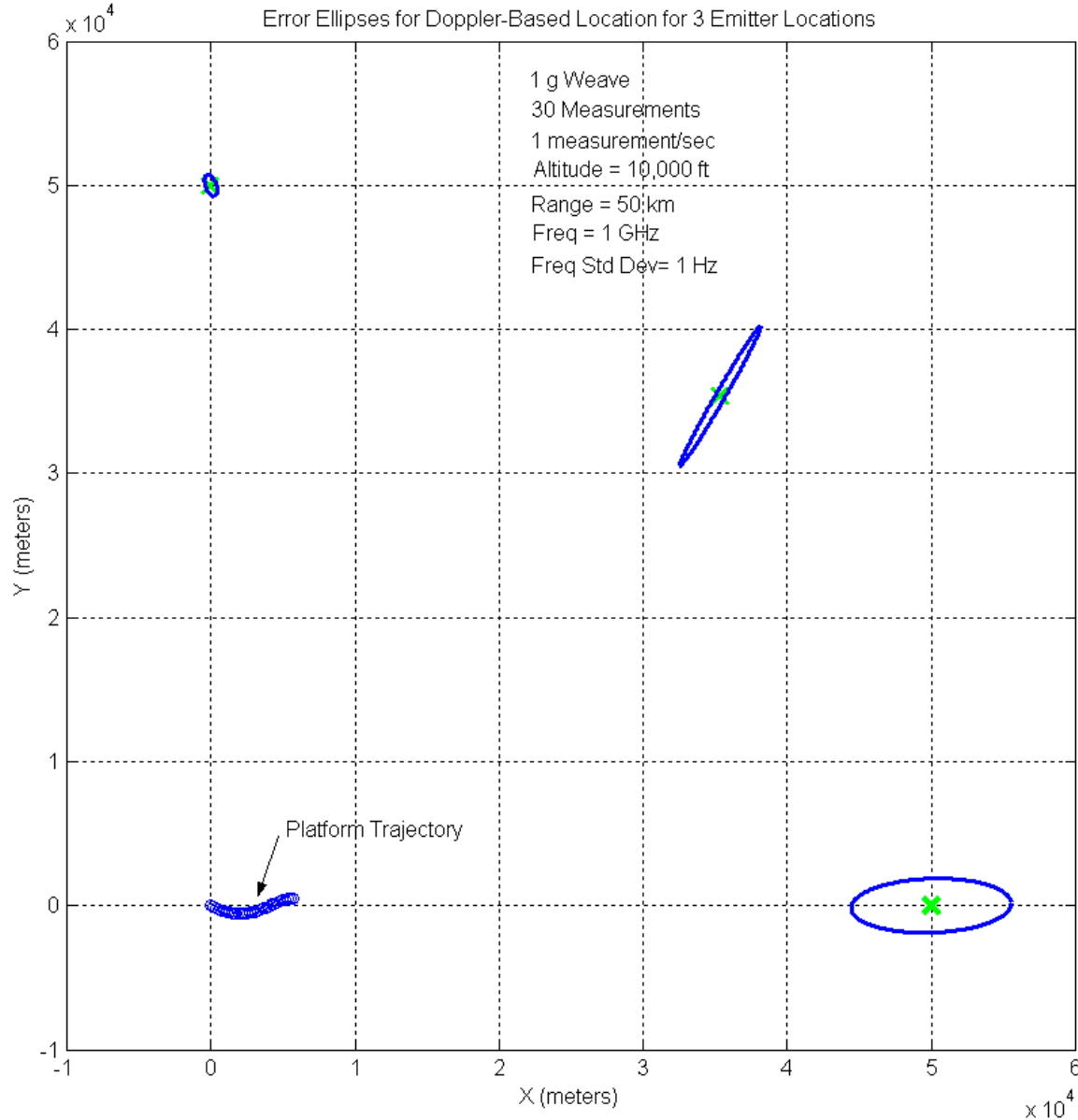
keep only the rows of the variables projecting onto

For this example, frequency-based emitter location: $[X \ Y \ Z \ f_o]^T$

To project this 4-D error ellipsoid onto the X-Y plane:

$$\mathbf{P} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{bmatrix}$$

Projections Applied to Emitter Location



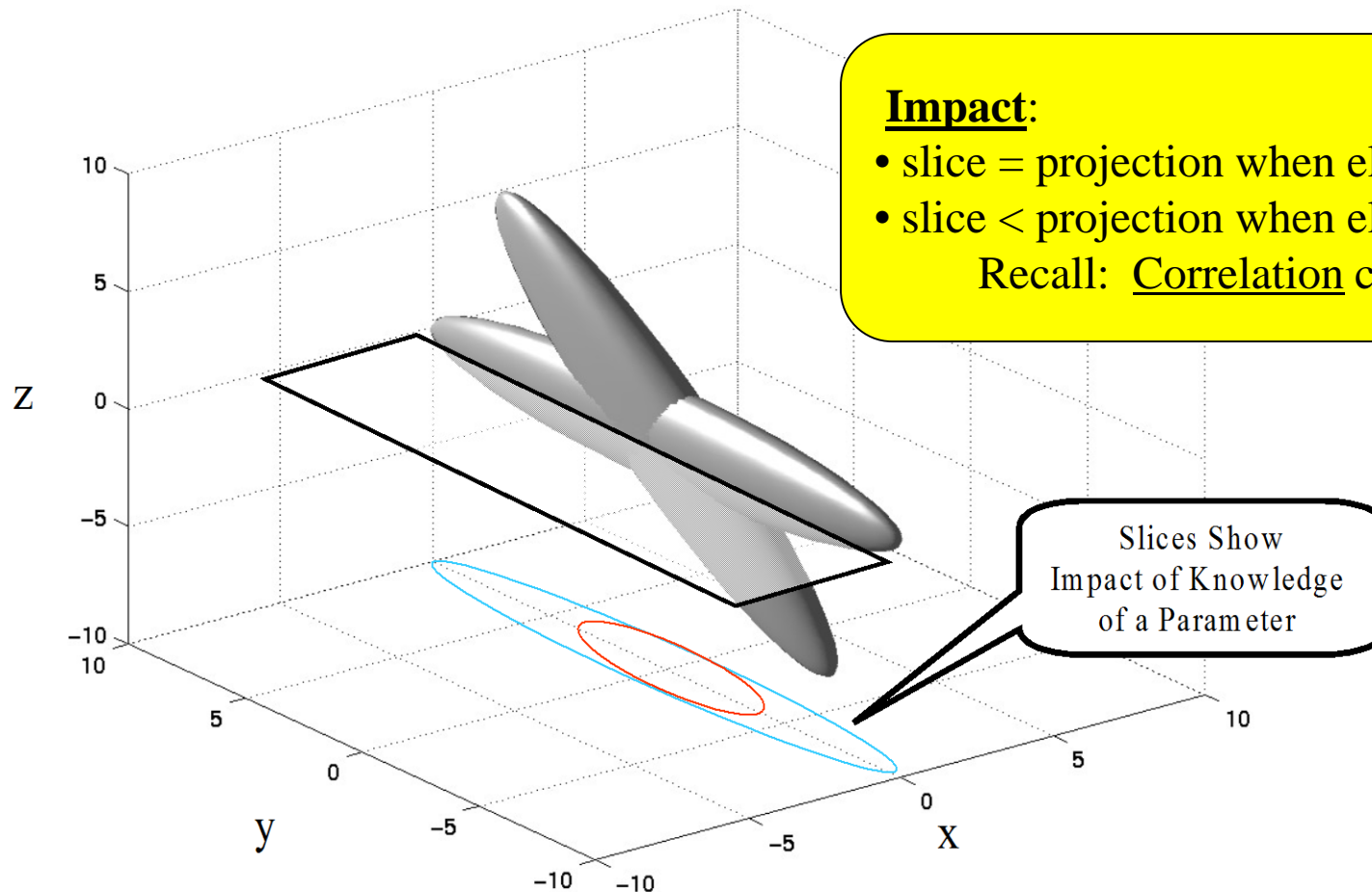
Shows 2-D ellipses that result from projecting 4-D ellipsoids

Slices of Error Ellipsoids

Q: What happens **if one parameter were perfectly known.**

Capture by setting that parameter's error to zero

⇒ **slice through the error ellipsoid.**



Impact:

- slice = projection when ellipsoid not tilted
- slice < projection when ellipsoid is tilted.

Recall: Correlation causes tilt

Slices Show
Impact of Knowledge
of a Parameter