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# Esercitazioni per il corso di STRUMENTAZIONE SPAZIALE

prof. E. Lorenzini

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## *USO DI MATLAB PER LA DETERMINAZIONE D'ASSETTO E L'ANALISI DI COVARIANZA*

- Single-axis attitude determination
- Three-axis attitude determination

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*Materiale didattico preparato da  
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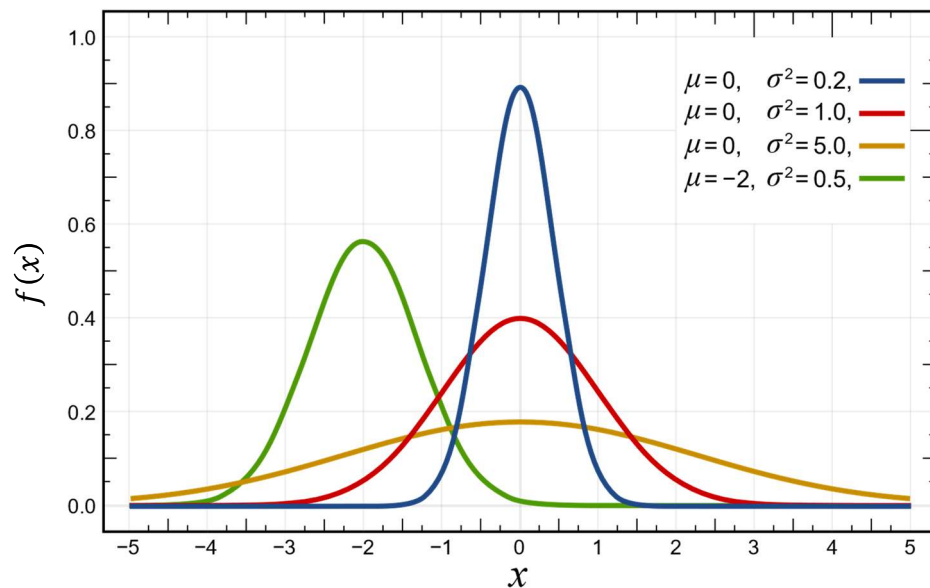
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DI PADOVA

# Univariate Normal or Gaussian distribution $\mathcal{N}(\mu, \sigma)$

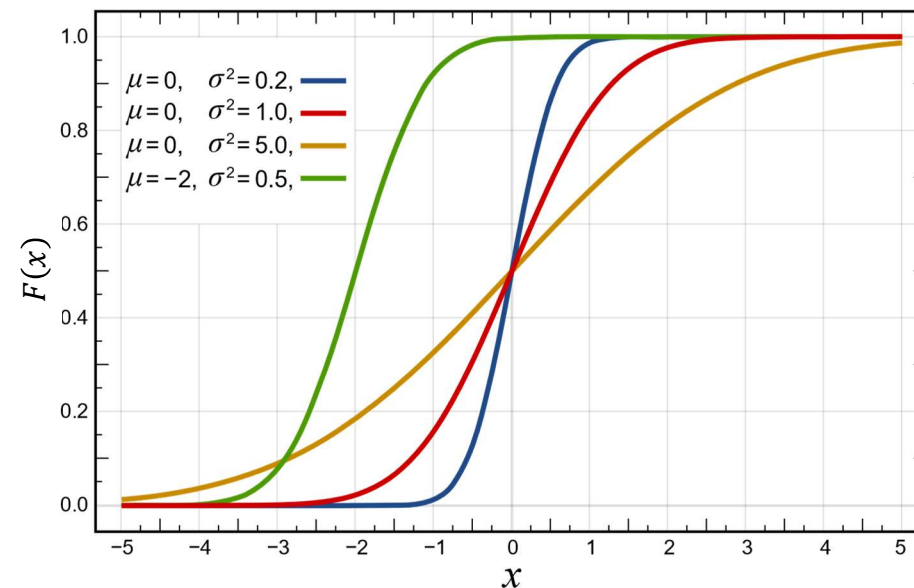
Probability Density Function (PDF)



$$f(x) = \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{1}{2} \left( \frac{x-\mu}{\sigma} \right)^2}$$

$\mu \Rightarrow$  mean value  
 $\sigma \Rightarrow$  standard deviation  
 $\sigma^2 \Rightarrow$  variance

Cumulative Distribution Function (CDF)



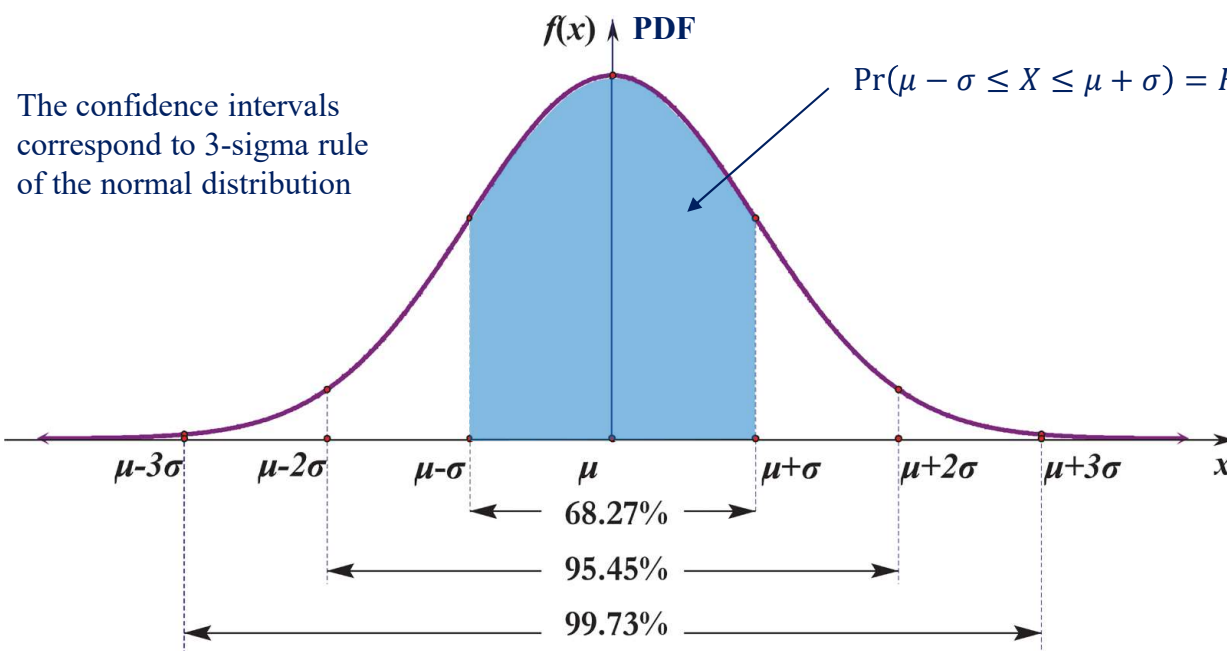
$$F(x) = \int_{-\infty}^x f(t) dt = \frac{1}{2} \left[ 1 + \operatorname{erf} \left( \frac{x-\mu}{\sigma\sqrt{2}} \right) \right],$$

$$\operatorname{erf}(y) = \frac{2}{\sqrt{\pi}} \int_{-\infty}^y e^{-\frac{t^2}{2}} dt$$

$F_X(x) = \Pr(X \leq x) \Rightarrow$  probability that the random variable  $X$  takes on a value less than or equal to  $x$ .

[https://en.wikipedia.org/wiki/Normal\\_distribution](https://en.wikipedia.org/wiki/Normal_distribution)

# The 3-sigma rule



univariate normal distribution		
$r$	$p$	$p$ [%]
1	0.6827	68.27
2	0.9545	95.45
3	0.9973	99.73

Matlab functions:

- erf
- erfc
- erfcinv

See also:

- norminv
- normcdf
- tinv
- tcdf
- unifinv
- unifcdf

$p \Rightarrow$  confidence level (livello di confidenza)

$r \Rightarrow$  magnification or coverage factor  
(fattore di copertura)

*Univariate Normal or Gaussian distribution*

$$p = \operatorname{erf}\left(\frac{r}{\sqrt{2}}\right) \Rightarrow r = \sqrt{2} \operatorname{erf}^{-1}(p)$$

# Multivariate Normal or Gaussian distribution $\mathcal{N}(\mu, \Sigma)$

## Probability Density Function (PDF) $f(\mathbf{x})$

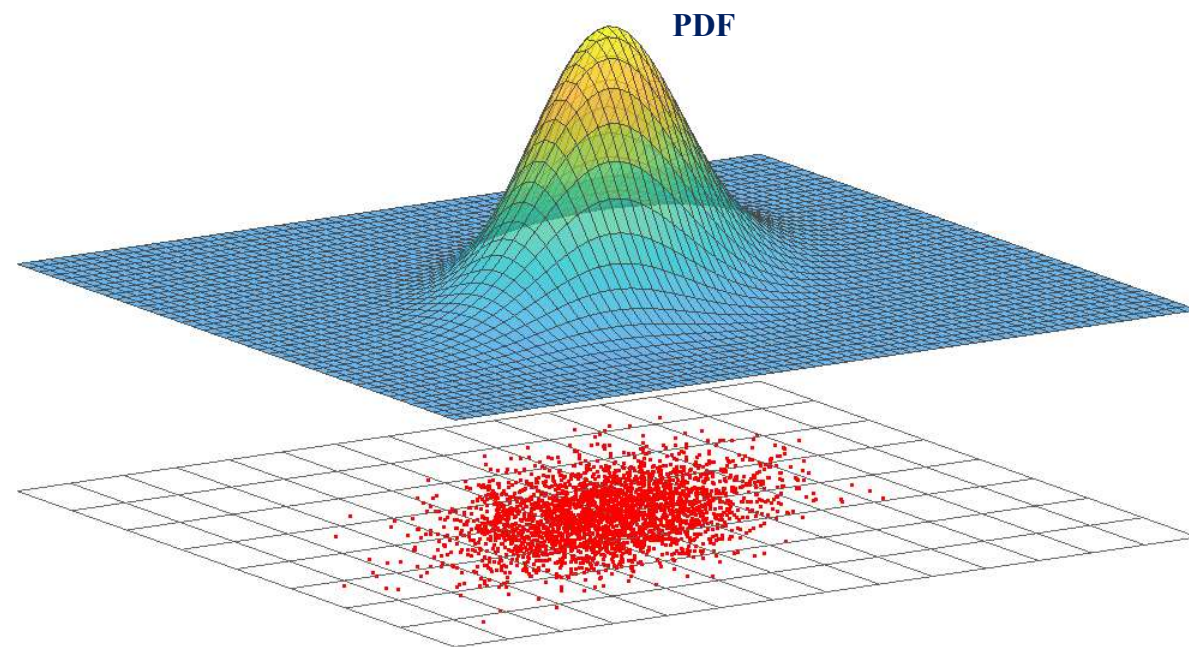
$$f(\mathbf{x}) = \frac{1}{\sqrt{(2\pi)^n |\Sigma|}} e^{-\frac{1}{2}(\mathbf{x}-\mu)^T \Sigma^{-1}(\mathbf{x}-\mu)}$$

$n \Rightarrow$  space dimensionality

$\mu \in \mathbb{R}^n \Rightarrow$  mean value

$\Sigma \in \mathbb{R}^{n \times n} \Rightarrow$  covariance matrix

$$\Sigma_{i,j} = \begin{cases} \text{var}(x_i) & \text{if } i = j \\ \text{cov}(x_i, x_j) & \text{if } i \neq j \end{cases}$$



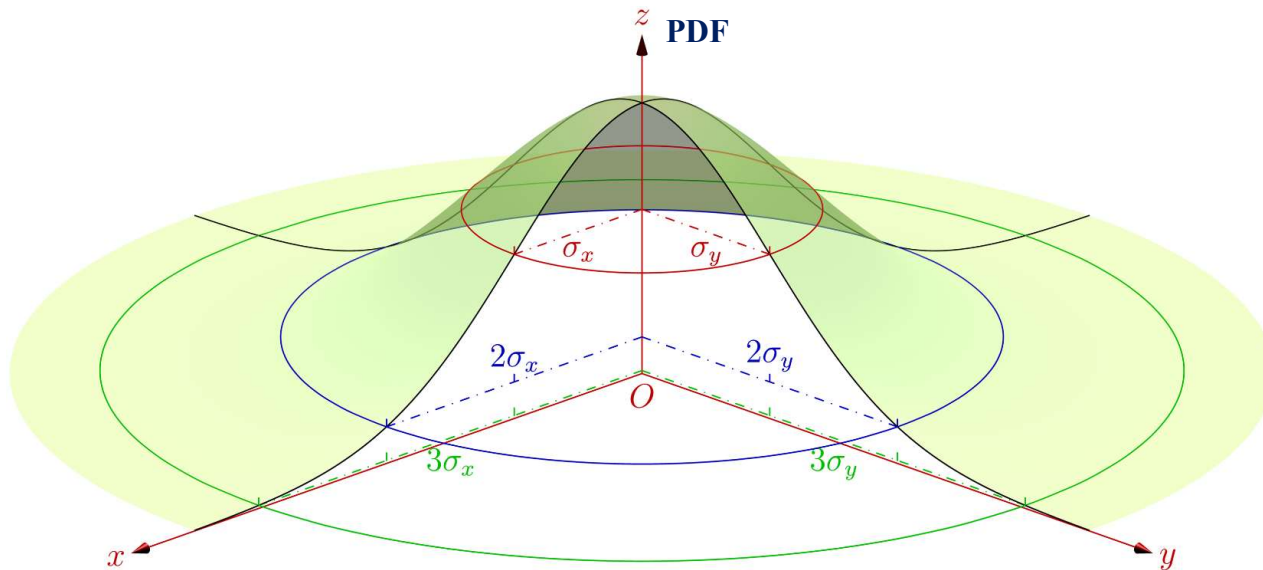
## Cumulative Distribution Function (CDF) $F(\mathbf{x})$

probability that all components of  $\mathbf{X}$  are less than or equal to the corresponding values in the vector  $\mathbf{x}$

$$F_{\mathbf{X}}(\mathbf{x}) = \Pr(\mathbf{X} \leq \mathbf{x})$$

There is no closed form for  $F(\mathbf{x})$ ; however, there are diverse algorithms to estimate it numerically.

# Bivariate Normal or Gaussian distribution $\mathcal{N}(\mu, \Sigma)$



Bivariate normal distribution		
$r$	$p$	$p$ [%]
1	0.3935	39.35
2	0.8647	86.47
3	0.9889	98.89

## *Bivariate Normal or Gaussian distribution*

$p \Rightarrow$  confidence level (livello di confidenza)

$r \Rightarrow$  magnification or coverage factor  
(fattore di copertura)

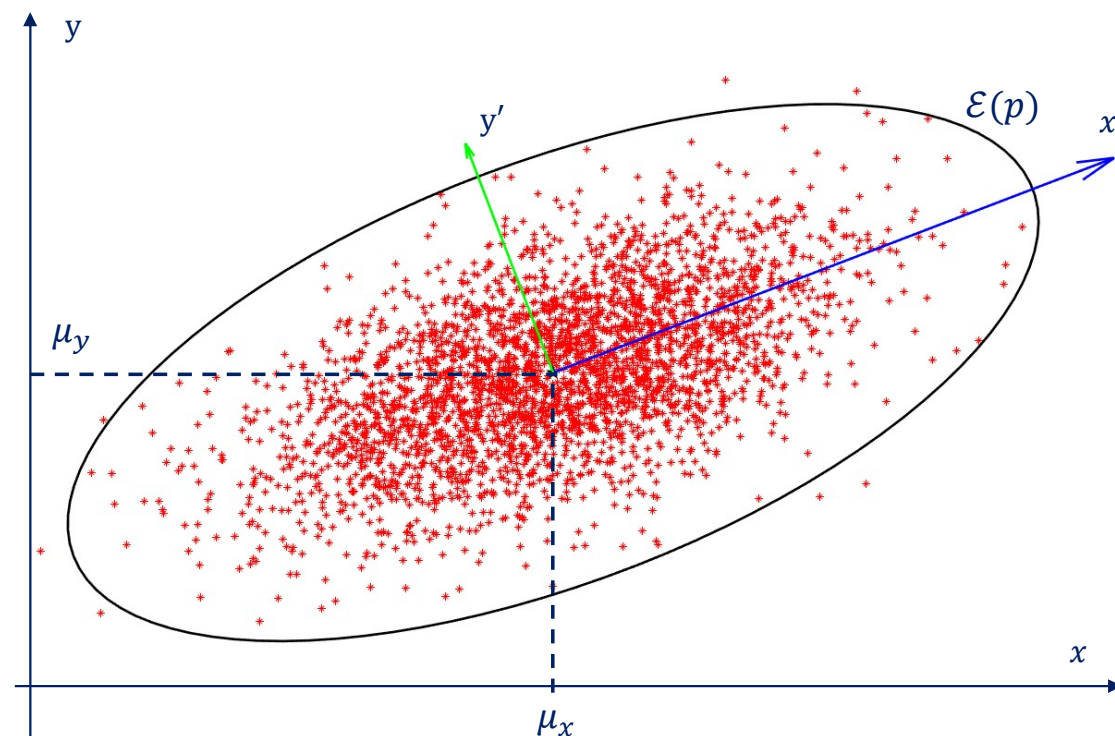
$$p = 1 - e^{-\frac{r^2}{2}} \Rightarrow r = \sqrt{-2 \ln(1 - p)}$$



# Uncertainty ellipse

The covariance matrix  $\Sigma \in \mathbb{R}^{2 \times 2}$  defines the **uncertainty ellipse**  $\mathcal{E}(p)$ :

- ✓ the eigenvectors of  $\Sigma$  defines the principal axes  $x'$  and  $y'$  of the uncertainty ellipse  $\mathcal{E}$
- ✓ the eigenvalues of  $\Sigma$  defines the standard deviations  $\sigma_{x'}$  and  $\sigma_{y'}$
- ✓ the semi-axes of the uncertainty ellipse for a given confidence level  $p$  are:
  - semi-major axis:  $a = r(p) \sigma_{x'}$
  - semi-minor axis:  $b = r(p) \sigma_{y'}$



- The principal axes  $x'$  and  $y'$  of the uncertainty ellipse  $\mathcal{E}$  can be found by computing eigenvectors and eigenvalues of the covariance matrix  $\Sigma$

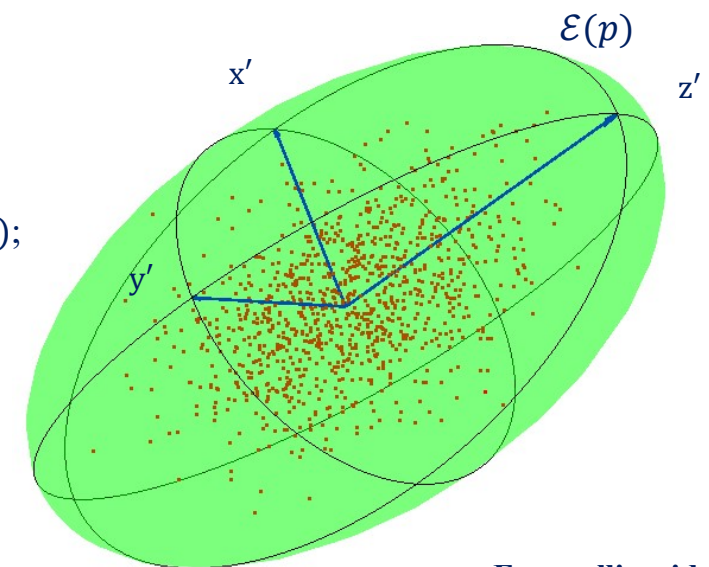
# Trivariate Normal or Gaussian distribution $\mathcal{N}(\mu, \Sigma)$

## Trivariate Normal or Gaussian distribution

- ✓ There is not an explicit relation between the confidence level  $p$  and the magnification factor  $r$ , as for the univariate or bivariate Normal Distribution.
- ✓ There is instead a numeric algorithm, see for instance:  
<https://journals.plos.org/plosone/article?id=10.1371/journal.pone.0118537>
- ✓ Note: use interpolation data  $p$ - $r$  to obtain the magnification factor  $r$  that corresponds to a given confidence level  $p$ .

Trivariate normal distribution		
$r$	$p$	$p$ [%]
1	0.1987	19.87
2	0.7384	73.85
3	0.9707	97.07

- the covariance matrix  $\Sigma \in \mathbb{R}^{3 \times 3}$  defines the error ellipsoid  $\mathcal{E}(p)$ :
  - ✓ the eigenvectors of  $\Sigma$  defines the principal axes  $x'$ ,  $y'$  and  $z'$  of the error ellipsoid  $\mathcal{E}(p)$ ;
  - ✓ the eigenvalues of  $\Sigma$  defines the standard deviations  $\sigma_{x'}$ ,  $\sigma_{y'}$  and  $\sigma_{z'}$  ;
  - ✓ the semi-axes of the error ellipse for a given confidence level  $p$  are:
    - semi axes:  $a_i = r(p) \sigma_{x'_i}$



Error ellipsoid

## Magnification factor $r$ vs. confidence level $p$ vs. dimensionality

Confidence levels of scaled Standard Deviation Hyper-Ellipsoid (SDHE) vary with different magnification factors in spaces with the dimensionality  $\leq 10$ .

Dimensionality	Magnification factor						
	1	2	3	4	5	6	7
1	0.6827	0.9545	0.9973	0.9999	1.0000	1.0000	1.0000
2	0.3935	0.8647	0.9889	0.9997	1.0000	1.0000	1.0000
3	0.1987	0.7385	0.9707	0.9989	1.0000	1.0000	1.0000
4	0.0902	0.5940	0.9389	0.9970	0.9999	1.0000	1.0000
5	0.0374	0.4506	0.8909	0.9932	0.9999	1.0000	1.0000
6	0.0144	0.3233	0.8264	0.9862	0.9997	1.0000	1.0000
7	0.0052	0.2202	0.7473	0.9749	0.9992	1.0000	1.0000
8	0.0018	0.1429	0.6577	0.9576	0.9984	1.0000	1.0000
9	0.0006	0.0886	0.5627	0.9331	0.9970	1.0000	1.0000
10	0.0002	0.0527	0.4679	0.9004	0.9947	0.9999	1.0000

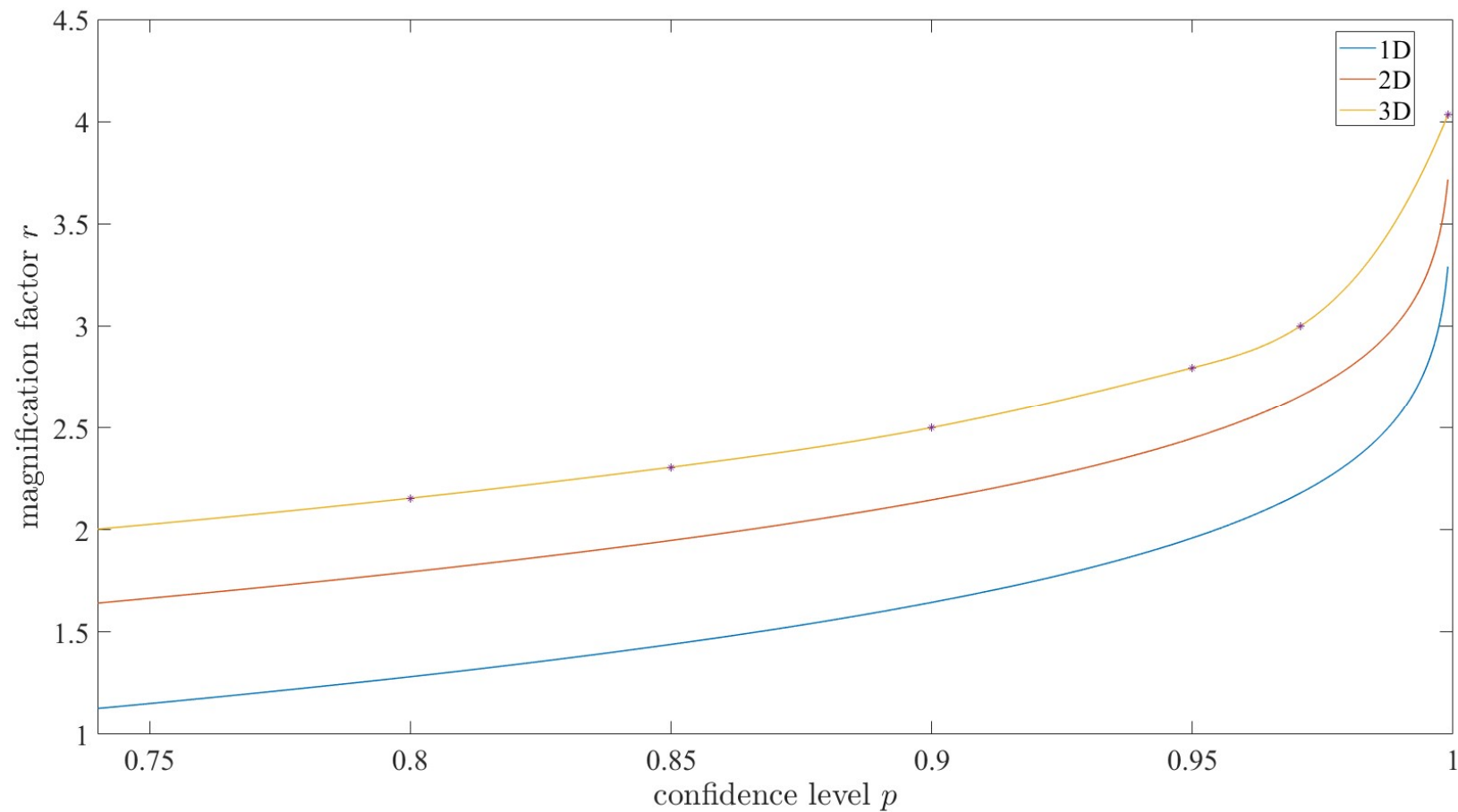
Magnification ratios of scaled SDHE corresponding to different specified confidence levels with space dimensionality  $\leq 10$ .

Dimensionality	Confidence Level (%)					
	80.0	85.0	90.0	95.0	99.0	99.9
1	1.2816	1.4395	1.6449	1.9600	2.5758	3.2905
2	1.7941	1.9479	2.1460	2.4477	3.0349	3.7169
3	2.1544	2.3059	2.5003	2.7955	3.3682	4.0331
4	2.4472	2.5971	2.7892	3.0802	3.6437	4.2973
5	2.6999	2.8487	3.0391	3.3272	3.8841	4.5293
6	2.9254	3.0735	3.2626	3.5485	4.1002	4.7390
7	3.1310	3.2784	3.4666	3.7506	4.2983	4.9317
8	3.3212	3.4680	3.6553	3.9379	4.4822	5.1112
9	3.4989	3.6453	3.8319	4.1133	4.6547	5.2799
10	3.6663	3.8123	3.9984	4.2787	4.8176	5.4395

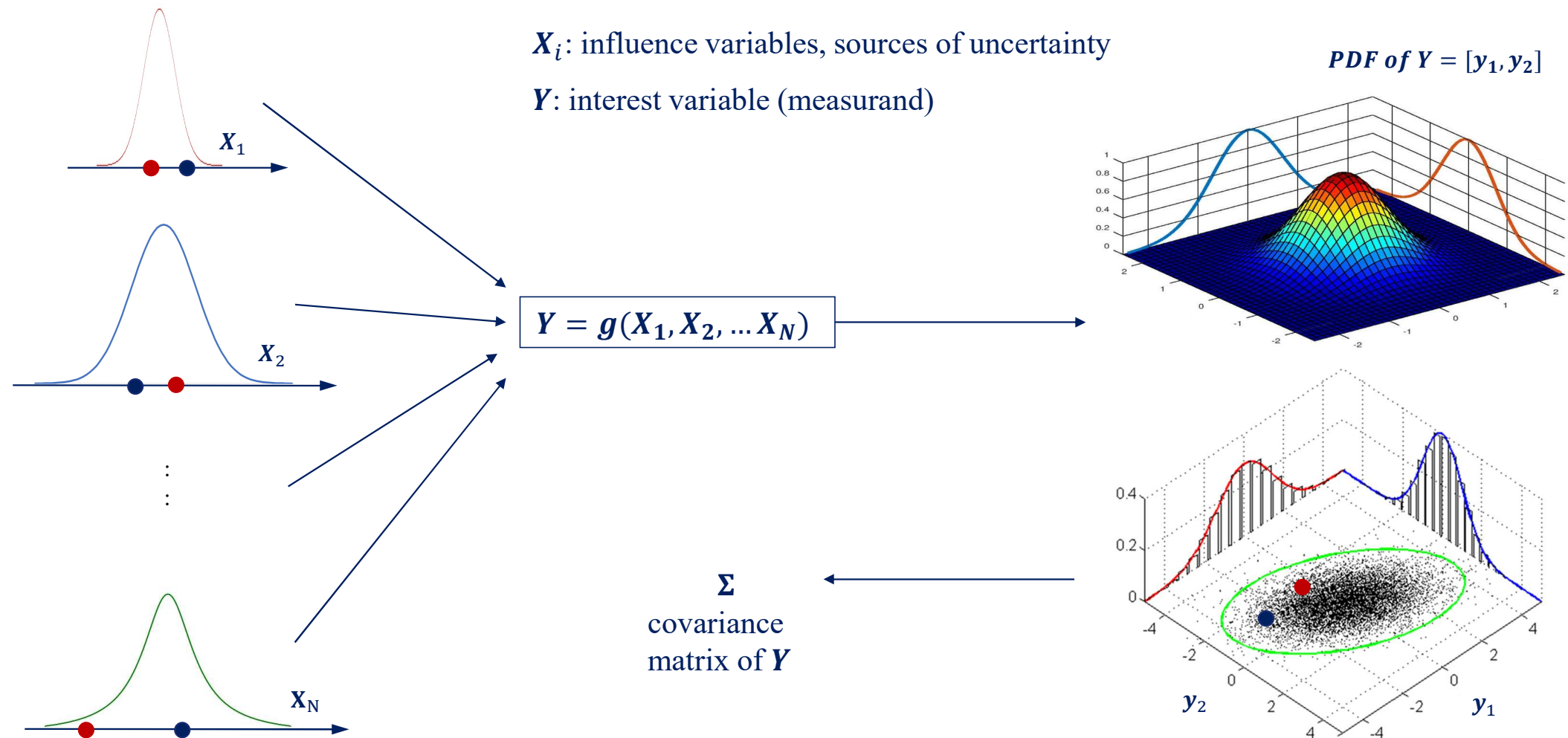
doi:10.1371/journal.pone.0118537.t002



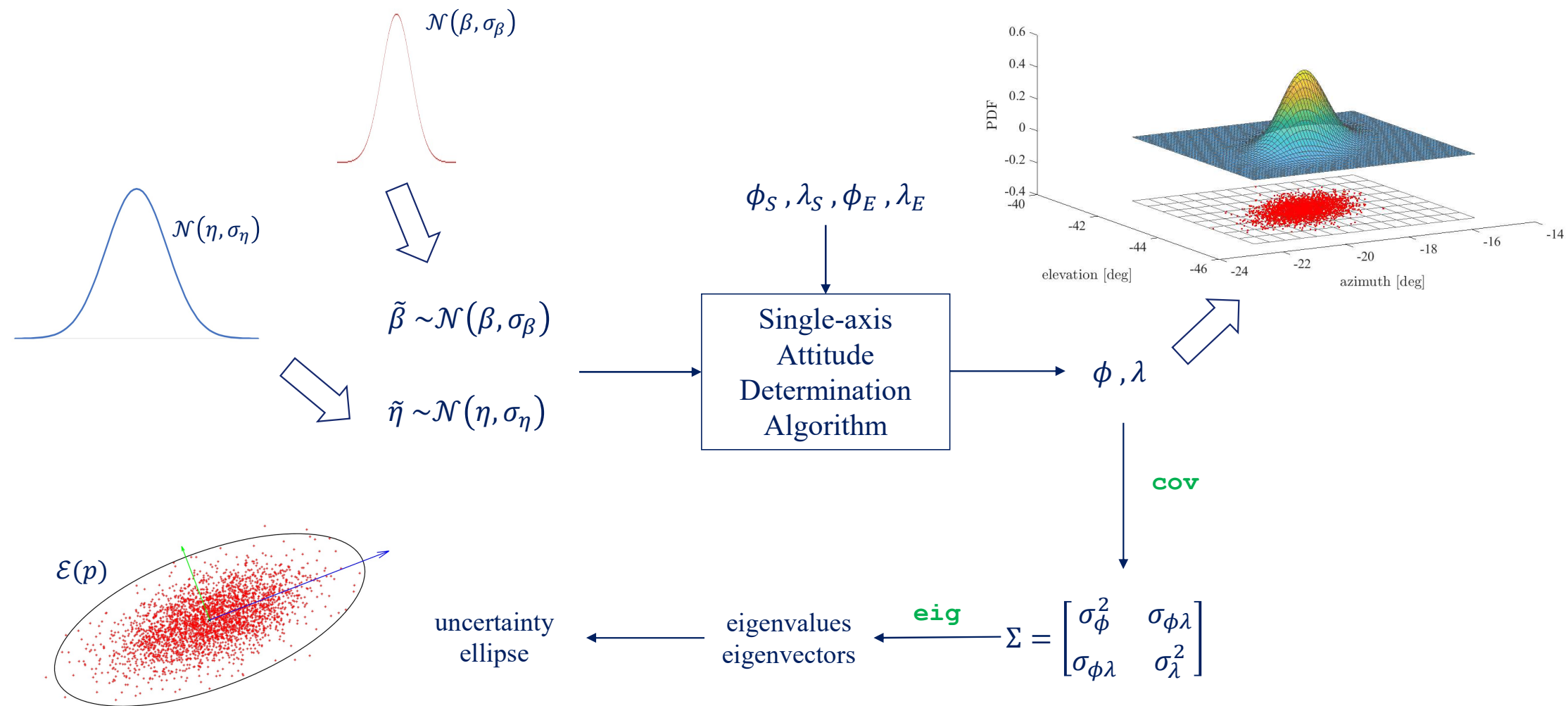
## Magnification factor $r$ vs. confidence level $p$ vs. dimensionality



# Covariance Analysis with Montecarlo Methods



# Covariance Analysis with Montecarlo Methods



## Esercitazione 5/1 di Strumentazione Aerospaziale

### ➤ **Prima esperienza: Single-Axis Attitude Determination**

- ✓ Metodo di Grubin per la determinazione ad asse singolo
- ✓ Analisi di Covarianza con metodo Montecarlo per la stima dell'incertezza d'assetto
- ✓ File Matlab:
  - ✓ main: «Esercitazione\_single\_axis\_determination\_Montecarlo.m»

In questo script di Matlab, dovrete seguire le istruzioni ed implementare la chiamata alla funzione con il metodo di Grubin e la parte di analisi di covarianza. Nello script avrete il testo con la consegna sottoforma di testo, mentre nelle varie celle dovrete implementare il codice richiesto nei commenti.

- ✓ subfunction: «SingleAxisAttDetAlgo.m» che implementa il metodo Grubin
- ✓ funzioni aggiuntive (built-in) di Matlab: deg2rad, rad2deg, eig, cov

## Esercitazione 5/2 di Strumentazione Aerospaziale

### ✓ Seconda esperienza: Three-Axes Attitude Determination

✓ Metodo-q di Davenport => routine QUEST

✓ File Matlab:

✓ main: «Esercitazione\_three\_axes\_determination\_quest.m»

Come nell'esercitazione precedente, sarete guidati nell'implementazione del metodo di Davenport e dell'analisi di covarianza.

✓ subfunctions: «quest.m» (scaricata dal Matlab File Exchange)

Chiamerete questa routine per la generazione delle matrici di rotazione per poi trovare gli angoli di Eulero

✓ altre funzioni: libreria «matGeom» scaricabile da Matlab File Exchange, `trasm`, `trasfg`, `pltassi`, `freccia`,  
`drawEllipsoidV2`

✓ funzioni aggiuntive (built-in) di Matlab: `unifrnd`, `deg2rad`, `rad2deg`, `eul2rotm`, `rotm2eul`, `sph2cart`, `cart2sph`, `randn`,  
`rms`, `eig`, `cov`