

# ***Geometria amostral***

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&  
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SPRINGER BRIEFS IN STATISTICS

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# Model Choice in Nonnested Families

 Springer

**De:** David Cox <[david.cox@nuffield.ox.ac.uk](mailto:david.cox@nuffield.ox.ac.uk)>

**Data:** 12 de maio de 2017 06:56:57 BRT

**Para:** Basilio de Bragança Pereira

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**Assunto:** your book

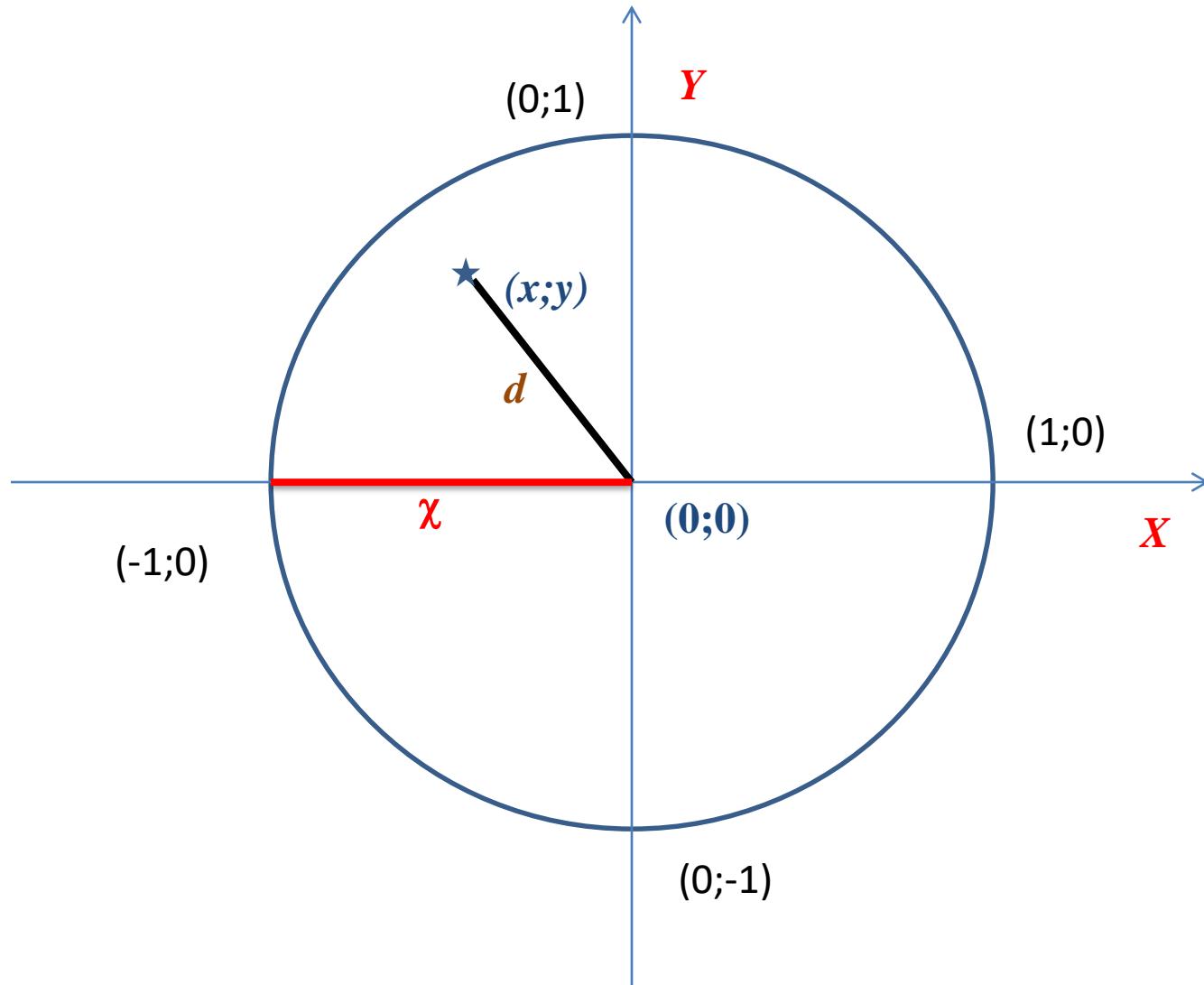
**Dear Basilio, I have just received the copy of  
your book. Very warm congratulations on it.  
It is great to have such a comprehensive account  
of this topic.**

**It is good to have the Bayesian and non-Bayesian  
approaches both so well presented. I deeply  
appreciate your kind remarks and the  
dedication.**

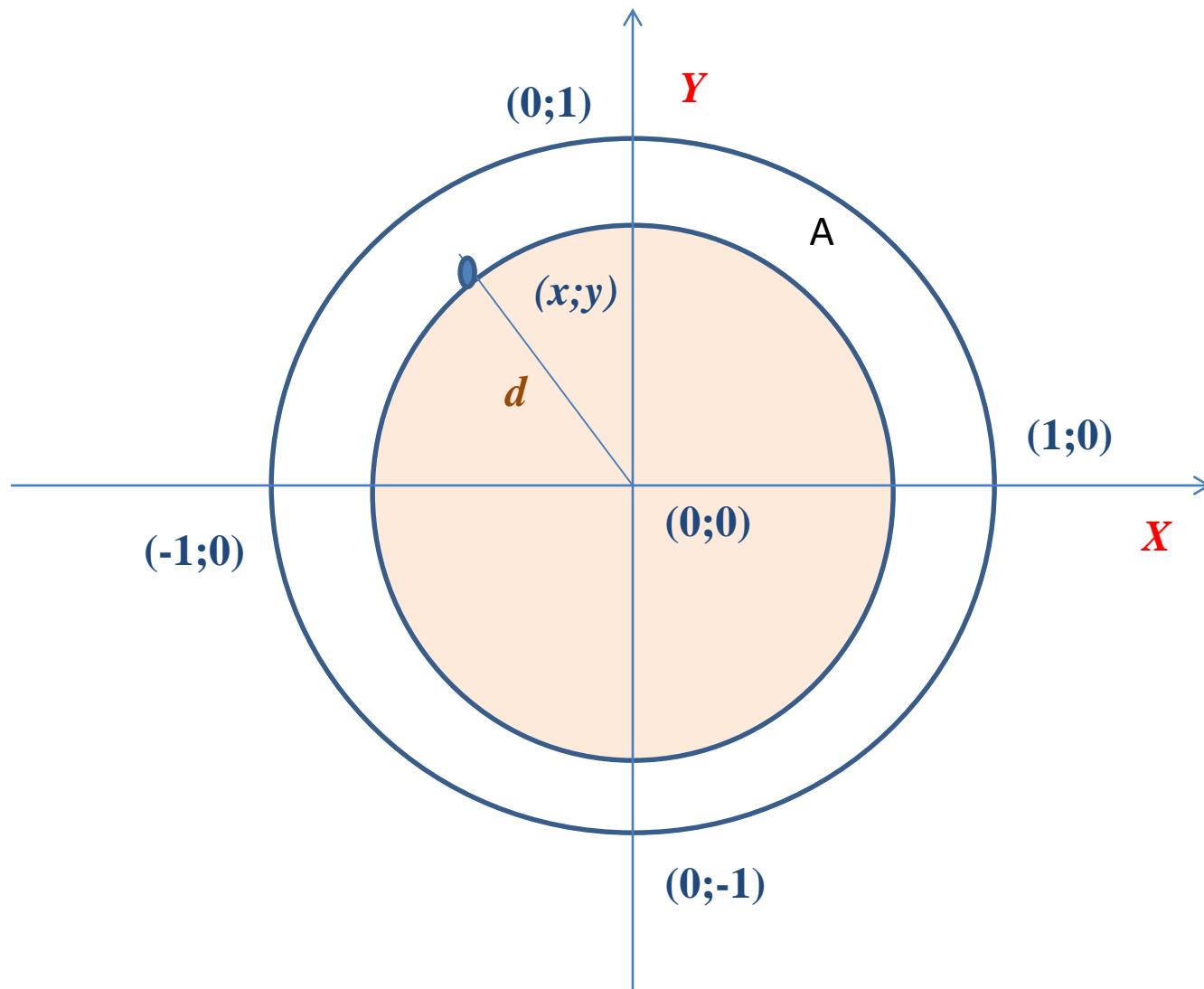
**For me the years at Imperial College  
were wonderful ones, good colleagues,  
outstanding doctoral students from so many  
countries and visitors too**

Best wishes

David



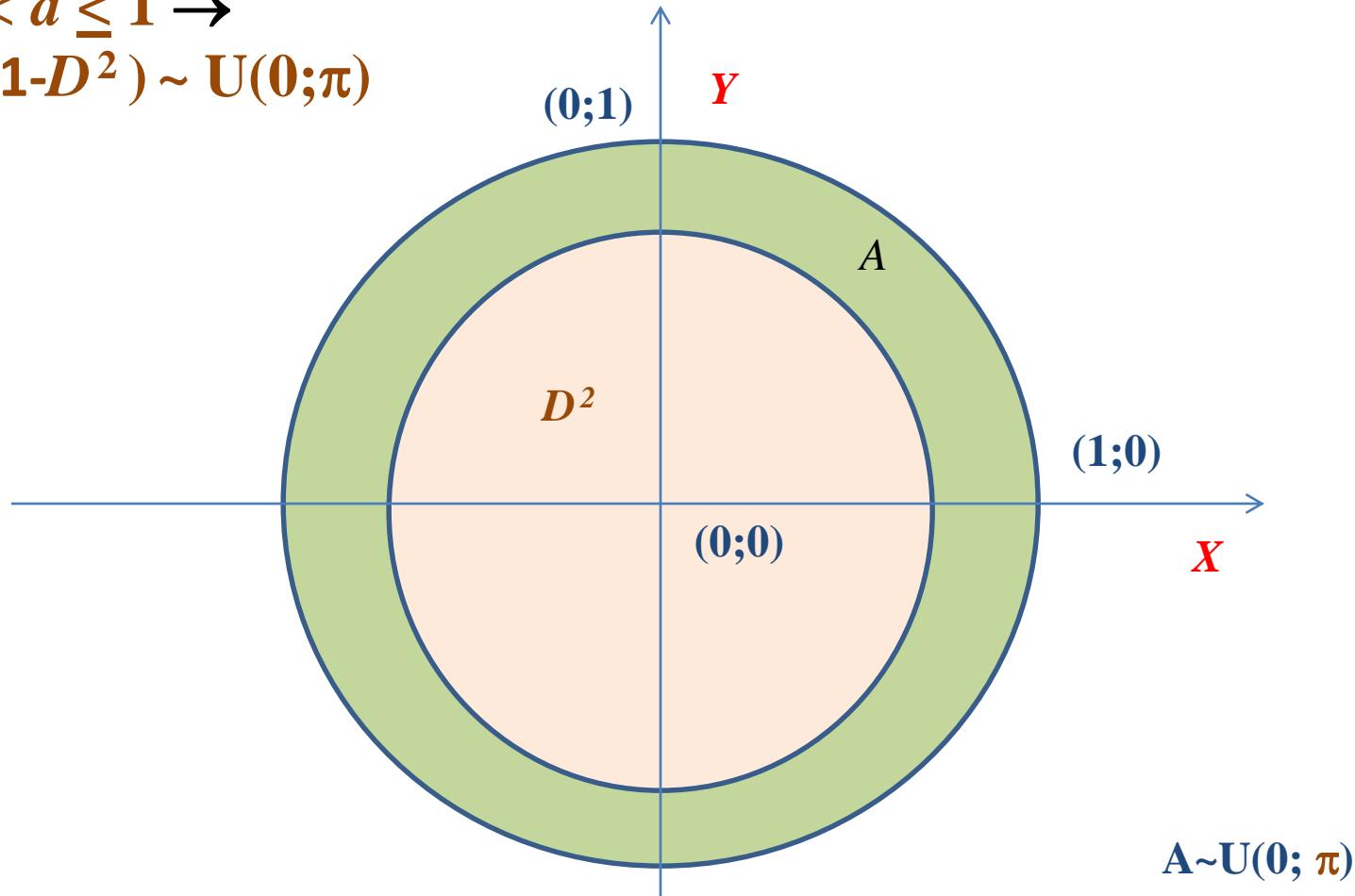
$$\Pr\{D \leq d\} = d^2 \rightarrow f(d) = 2d \text{ for } 0 < d \leq 1$$



$$\Pr\{D \leq d\} = d^2 \rightarrow f(d) = 2d$$

for  $0 < d \leq 1 \rightarrow$

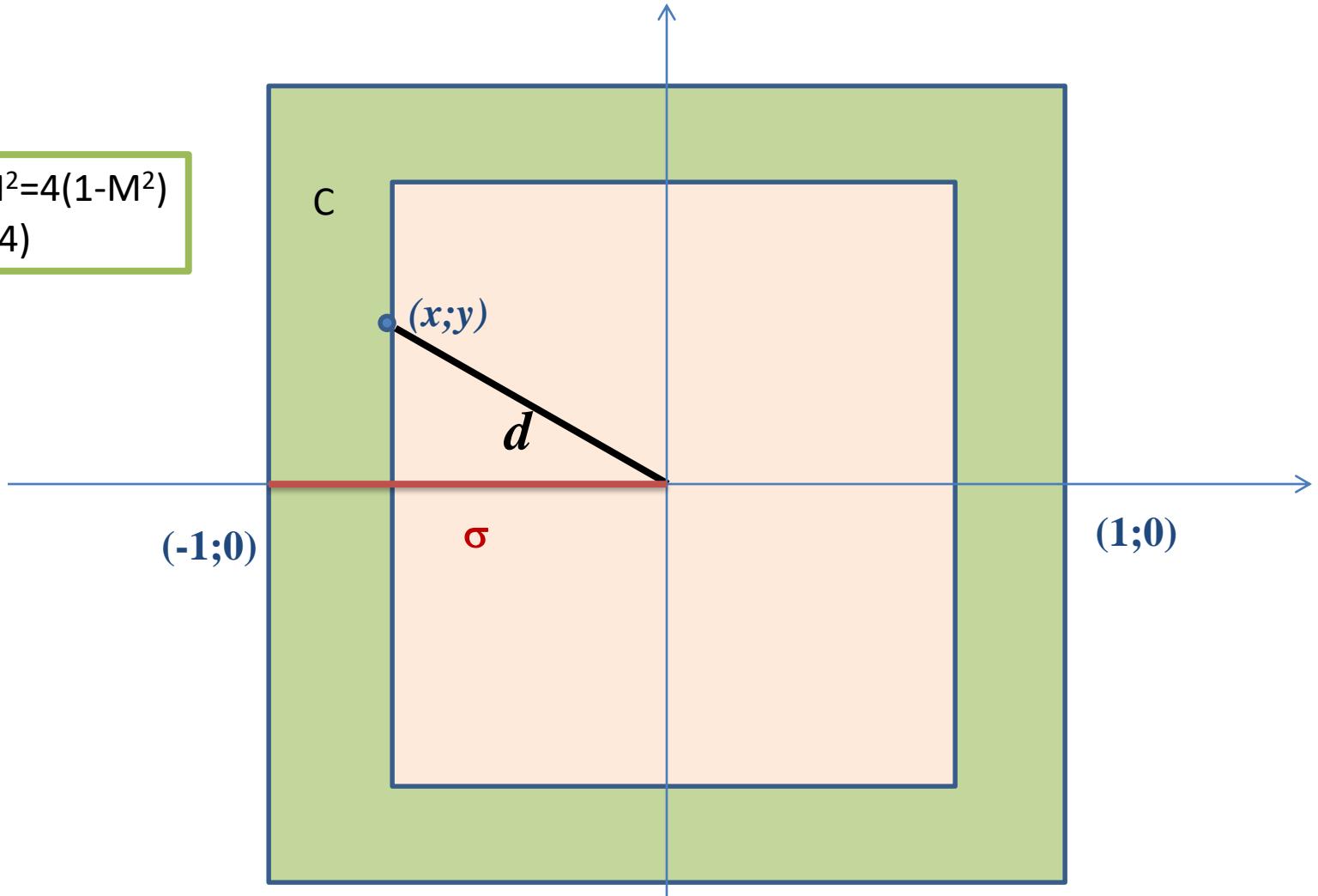
$$A = \pi * (1 - D^2) \sim U(0; \pi)$$



$$A = \pi * (1 - D^2) = 1 - F(D) \sim U(0; \pi)$$

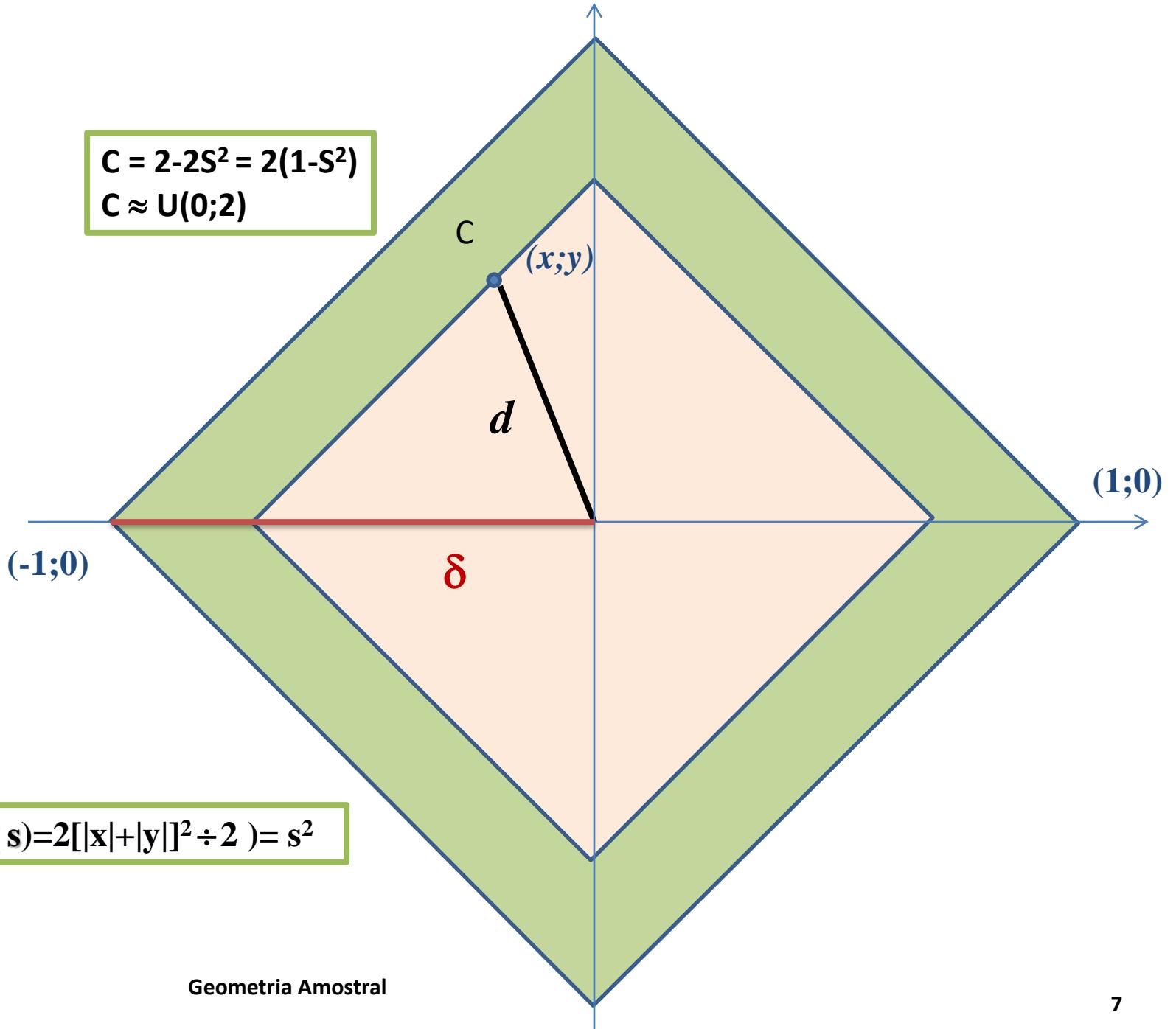
$$C = 4 - 4M^2 = 4(1 - M^2)$$

$$C \approx U(0;4)$$



$$\Pr(M < m) = [2 \operatorname{Max}(|x|; |y|)]^2 \div 4 ) = [\operatorname{Max}(|x|; |y|)]^2 = m^2$$

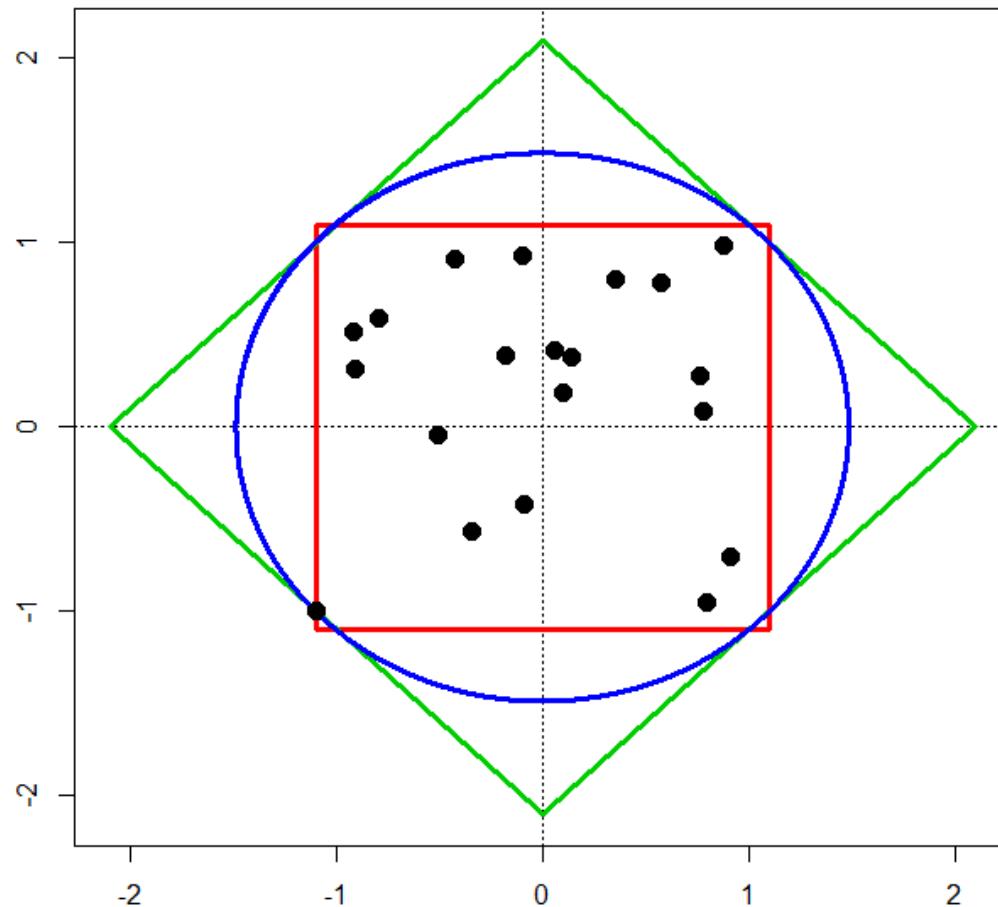
$$C = 2 - 2S^2 = 2(1 - S^2)$$
$$C \approx U(0;2)$$



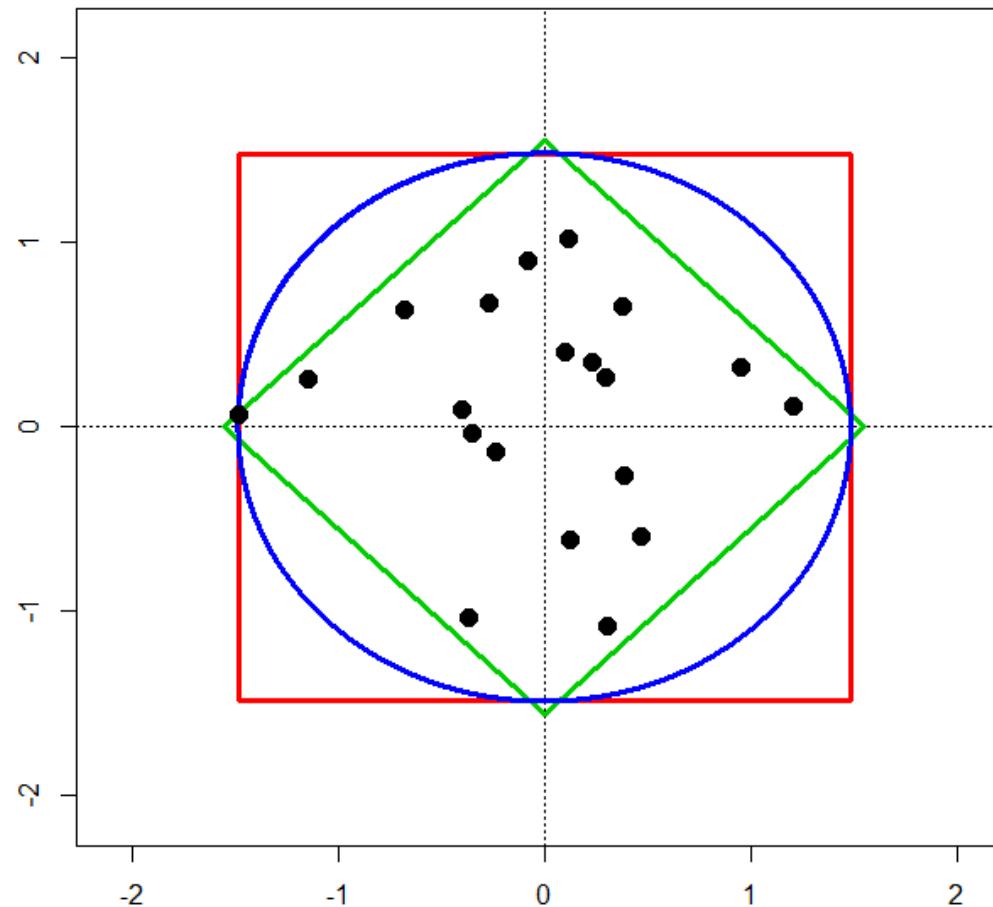
$$\Pr(S \leq s) = 2[|x| + |y|]^2 \div 2 = s^2$$

Sample number	QUADRADO			$\sigma$	$\delta$	$\chi$	DIAMANTE			$\sigma$	$\delta$	$\chi$	CÍRCULO			$\sigma$	$\delta$	$\chi$
	x	y	xy				x	y	xy				x	y	xy			
1	-0.500	0.903	-0.452	0.903	1.403	1.032	0.951	0.326	0.310	0.951	1.277	1.006	0.153	1.044	0.160	1.044	1.197	1.055
2	0.554	0.765	0.424	0.765	1.320	0.945	0.466	-0.597	-0.278	0.597	1.063	0.757	0.914	0.071	0.065	0.914	0.985	0.917
3	-0.244	0.347	-0.085	0.347	0.592	0.425	0.124	-0.614	-0.076	0.614	0.738	0.627	-0.664	0.068	-0.045	0.664	0.733	0.668
4	0.754	0.236	0.178	0.754	0.990	0.790	1.204	0.112	0.135	1.204	1.316	1.209	-1.040	-0.214	0.223	1.040	1.254	1.061
5	0.875	0.988	0.864	0.988	1.862	1.319	0.232	0.355	0.082	0.355	0.587	0.424	0.001	-0.505	0.000	0.505	0.505	0.505
6	-1.009	0.268	-0.271	1.009	1.278	1.044	-0.267	0.676	-0.181	0.676	0.944	0.727	0.792	0.531	0.420	0.792	1.322	0.953
7	0.007	0.381	0.002	0.381	0.387	0.381	-0.403	0.092	-0.037	0.403	0.495	0.413	1.183	0.056	0.066	1.183	1.239	1.184
8	0.774	0.031	0.024	0.774	0.805	0.774	0.294	0.268	0.079	0.294	0.562	0.398	0.399	-0.942	-0.376	0.942	1.341	1.023
9	0.056	0.138	0.008	0.138	0.193	0.148	0.384	-0.260	-0.100	0.384	0.644	0.464	0.180	0.524	0.095	0.524	0.705	0.554
10	-0.144	-0.510	0.073	0.510	0.654	0.530	0.306	-1.079	-0.330	1.079	1.385	1.122	-0.546	-0.588	0.321	0.588	1.133	0.802
11	0.909	-0.812	-0.738	0.909	1.721	1.219	-0.680	0.636	-0.433	0.680	1.317	0.931	0.124	-0.269	-0.033	0.269	0.392	0.296
12	-0.151	0.921	-0.139	0.921	1.072	0.934	-0.077	0.906	-0.070	0.906	0.983	0.909	0.915	0.108	0.099	0.915	1.023	0.921
13	0.321	0.792	0.255	0.792	1.114	0.855	-0.365	-1.032	0.377	1.032	1.397	1.095	0.088	-1.289	-0.114	1.289	1.377	1.292
14	0.100	0.343	0.034	0.343	0.443	0.357	-1.147	0.264	-0.303	1.147	1.411	1.177	-0.068	-1.032	0.071	1.032	1.101	1.034
15	-0.889	0.565	-0.502	0.889	1.454	1.053	0.103	0.411	0.042	0.411	0.514	0.424	0.494	0.317	0.157	0.494	0.811	0.587
16	0.789	-1.072	-0.846	1.072	1.862	1.331	-0.237	-0.131	0.031	0.237	0.367	0.270	-0.786	-0.139	0.109	0.786	0.925	0.798
17	-0.587	-0.110	0.064	0.587	0.697	0.597	-0.354	-0.038	0.013	0.354	0.391	0.356	0.978	-0.113	-0.110	0.978	1.091	0.984
18	-1.017	0.487	-0.495	1.017	1.504	1.127	0.384	0.659	0.253	0.659	1.043	0.763	-0.870	0.190	-0.165	0.870	1.060	0.891
19	-0.415	-0.665	0.276	0.665	1.080	0.784	0.118	1.026	0.121	1.026	1.143	1.032	-1.123	-1.123	1.261	1.123	2.246	1.588
20	-1.094	-1.015	1.110	1.094	2.109	1.492	-1.485	0.071	-0.105	1.485	1.556	1.487	1.458	0.105	0.153	1.458	1.563	1.462
Stat	<b>-0.046</b>	<b>0.149</b>	<b>-0.011</b>	1.094	2.109	1.492	<b>-0.023</b>	<b>0.103</b>	<b>-0.023</b>	1.485	1.556	1.487	<b>0.129</b>	<b>-0.160</b>	<b>0.118</b>	1.458	2.246	1.588
Var	<b>0.434</b>	<b>0.403</b>	<b>0.220</b>				<b>0.383</b>	<b>0.329</b>	<b>0.044</b>				<b>0.569</b>	<b>0.347</b>	<b>0.098</b>			
Cov			<b>-0.004</b>						<b>-0.021</b>						<b>0.138</b>			
Área				<b>4.788</b>	<b>8.895</b>	<b>6.996</b>				<b>8.820</b>	<b>4.840</b>	<b>6.943</b>				<b>8.508</b>	<b>10.090</b>	<b>7.925</b>

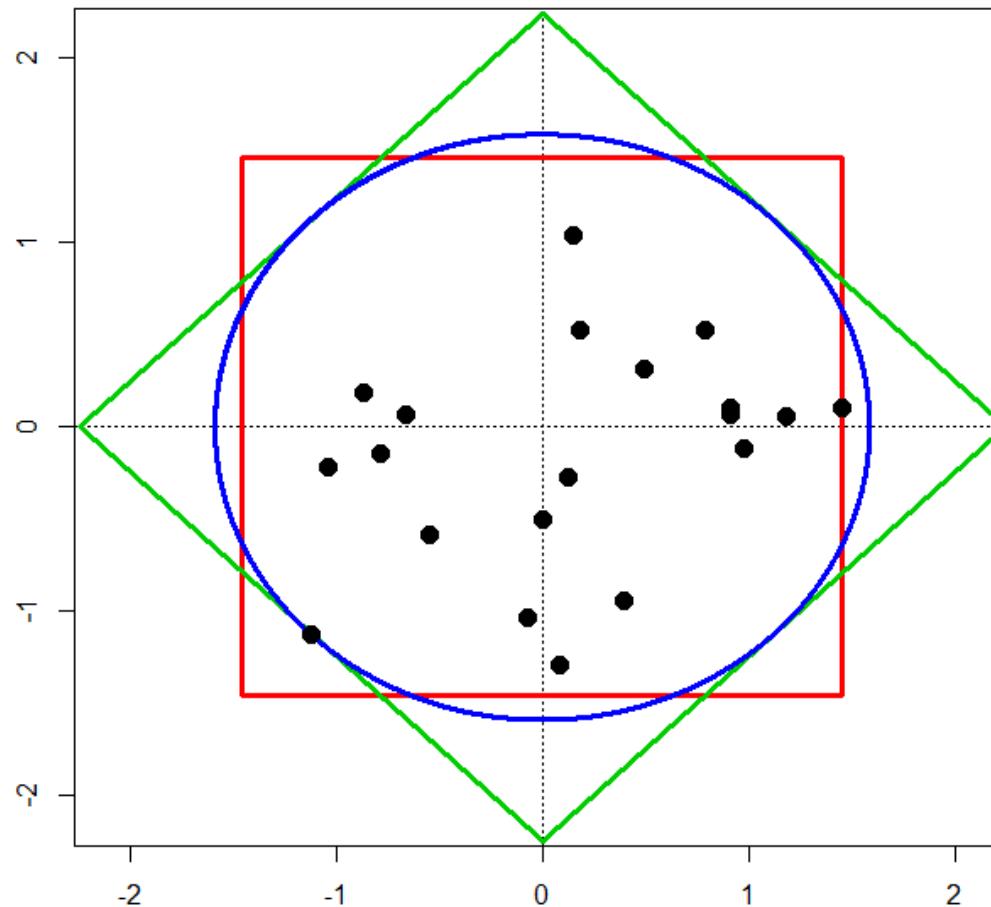
# Square



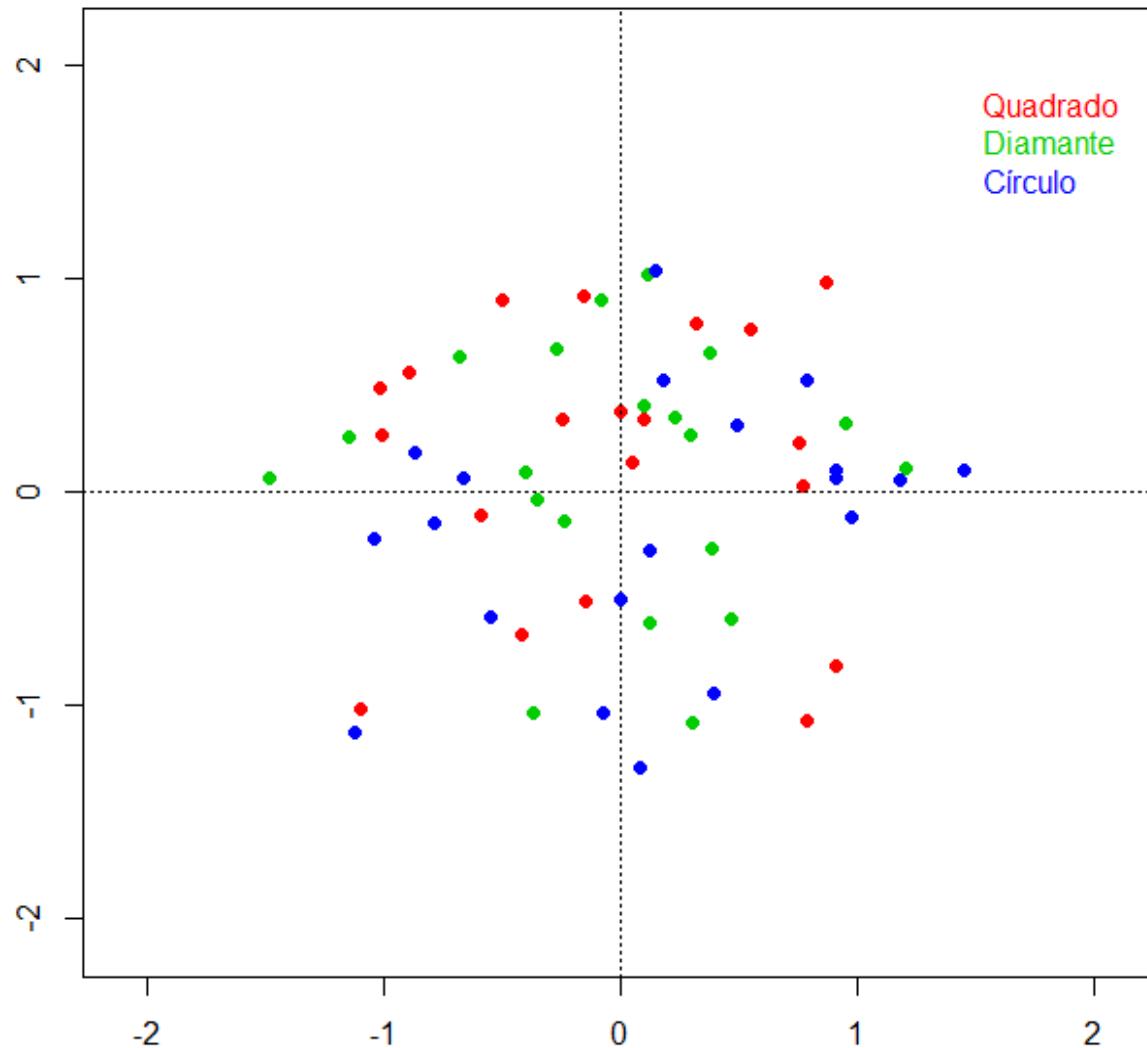
# Diamond



# Circle



# Samples



## Bayesian Operation: Prior to posterior.

### Priors, likelihoods and Posteriors

- **Sufficient Statistics, LIKELIHOODS, and Priors**

- $c = \text{Max} \sqrt{x_i^2 + y_i^2} \Delta d = \text{Max}(|x_i| + |y_i|)$   
 $s = \text{Max}(\text{Max}(|x_i|; |y_i|))$

- **Square:**  $L(\sigma|s) \propto (4\sigma^2)^{-n}$  for  $s \leq \sigma$
- **Diamond:**  $L(\delta|d) \propto (2\delta^2)^{-n}$  for  $d \leq \delta$
- **Circle:**  $L(\chi|c) \propto (\pi\theta^2)^{-n}$  for  $c \leq \theta$
- **Priors:**  $g(\sigma) \propto (\sigma^2)^{-1}$ ;  $g(\delta) \propto (\delta^2)^{-1}$  &  $g(\theta) \propto (\theta^2)^{-1}$
- **Posteriors are Pareto Distributions with parameters:**
  - Shape =  $2n + 1$  and Scales=  $\sigma$ ,  $\delta$ , &  $\theta$ , respectively to square, diamond and Circle.

## Pareto densities

$$f(\sigma|S = s) = \frac{(2n+1)s^{(2n+1)}}{\sigma^{(2n+2)}}; \sigma \geq s$$

$$f(\delta|D = d) = \frac{(2n+1)d^{(2n+1)}}{\delta^{(2n+2)}}; \delta \geq d$$

$$f(\theta|C = c) = \frac{(2n+1)c^{(2n+1)}}{\theta^{(2n+2)}}; \theta \geq c.$$

$$\left( mo_C = c; \quad m_c = \left( 1 + \frac{1}{2n} \right) r; \quad me_c = (\sqrt[2n+1]{2}) c \right);$$

$$\left( mo_S = s; \quad m_s = \left( 1 + \frac{1}{2n} \right) s; \quad me_s = (\sqrt[2n+1]{2}) s \right);$$

$$\left( mo_D = d; \quad m_d = \left( 1 + \frac{1}{2n} \right) d; \quad me_d = (\sqrt[2n+1]{2}) d \right).$$

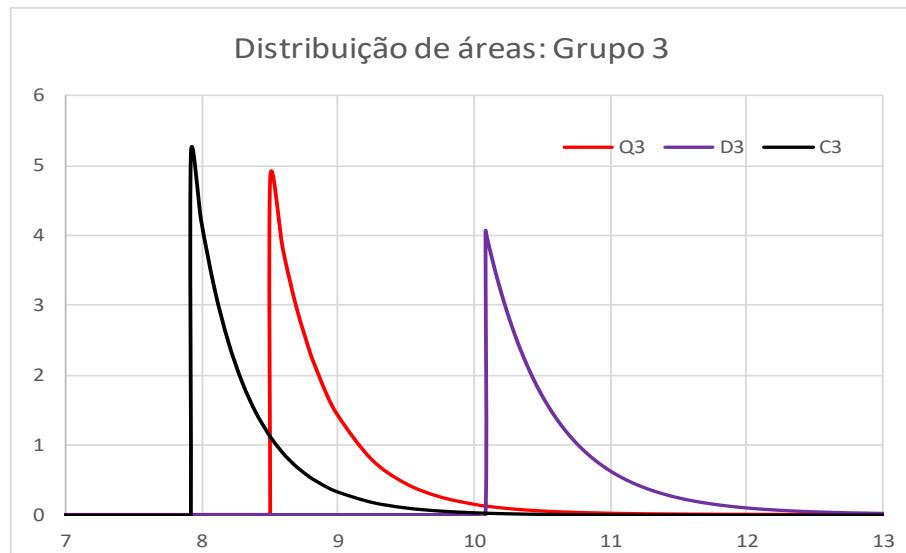
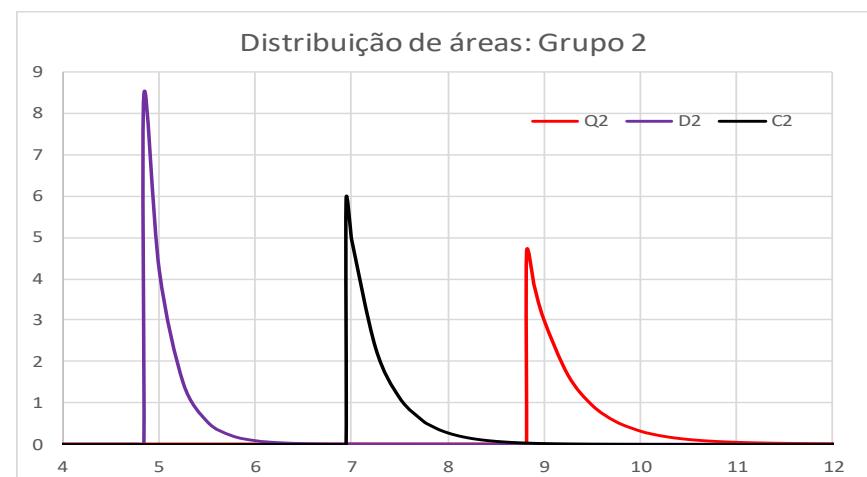
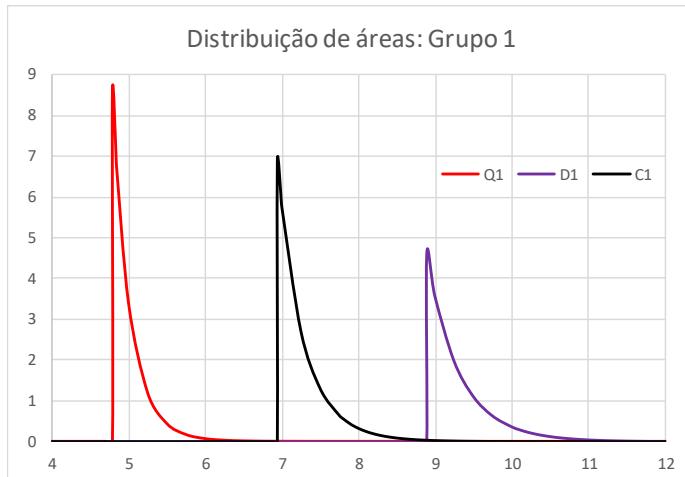
## Area densities

$$f(Q|S = s) = \frac{(2n + 1)4^{(n+5)}s^{(2n+1)}}{Q^{(n+1.5)}} ; Q \geq 4s^2$$

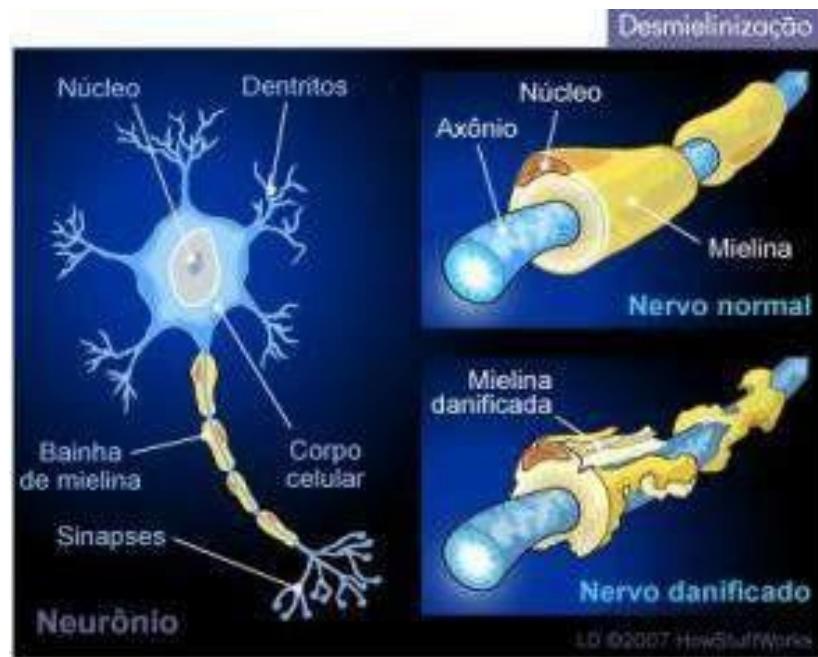
$$f(L|D = d) = \frac{(2n + 1)2^{(n+5)}d^{(2n+1)}}{2L^{(n+1.5)}} ; L \geq 2d^2$$

$$f(A|C = c) = \frac{(2n+1)\pi^{(n+5)}c^{(2n+1)}}{2A^{(n+1)}} ; A \geq \pi c^2.$$

# Gráficos das densidades das áreas



# Esclerose Múltipla 1



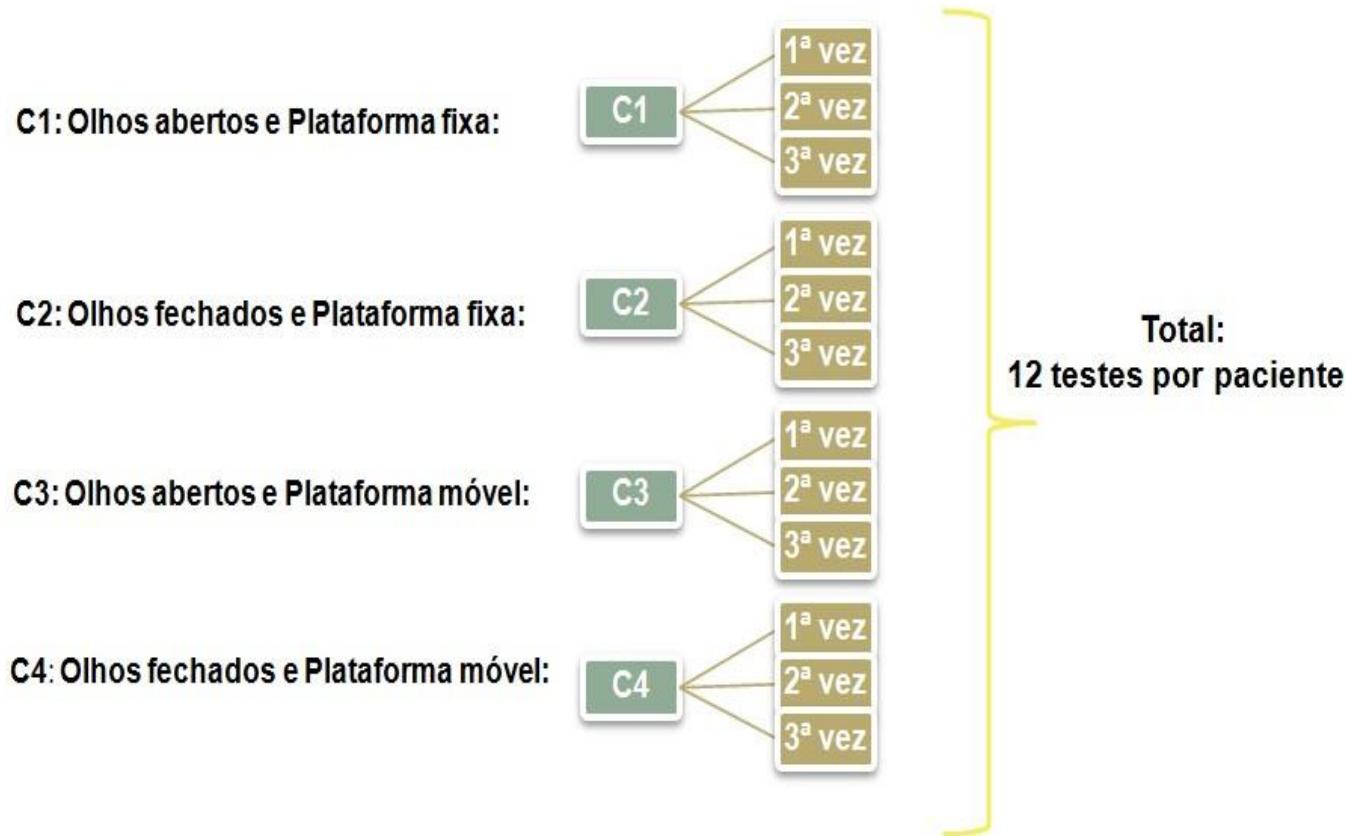
**Figura 1 - Processo de desmielinização**



**Figura 2 – Plataforma utilizada nos testes de PDC**

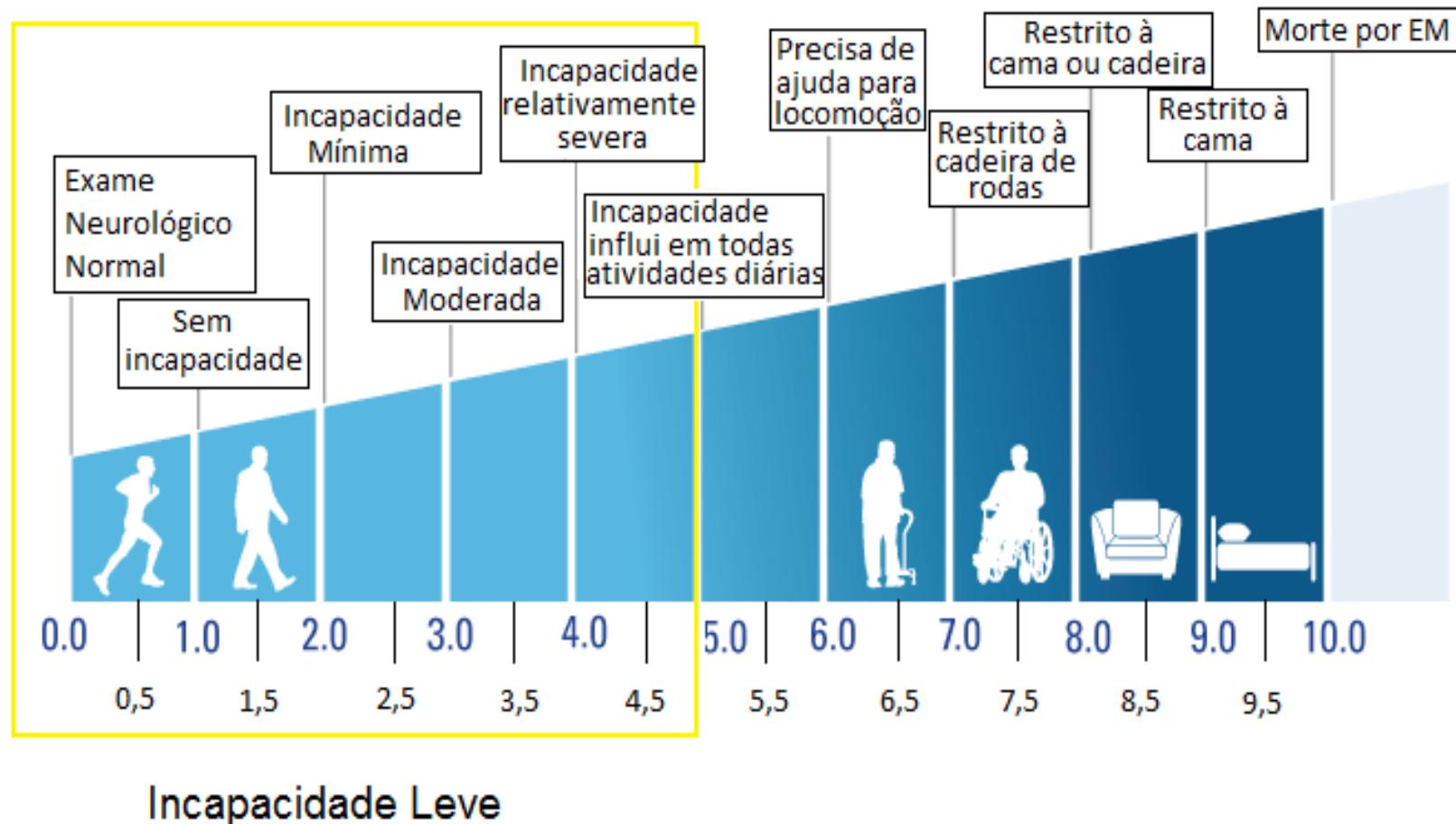
# Esclerose Múltipla 2

**Figura 3 - Condições aplicadas no exame**



# Esclerose Múltipla 3:

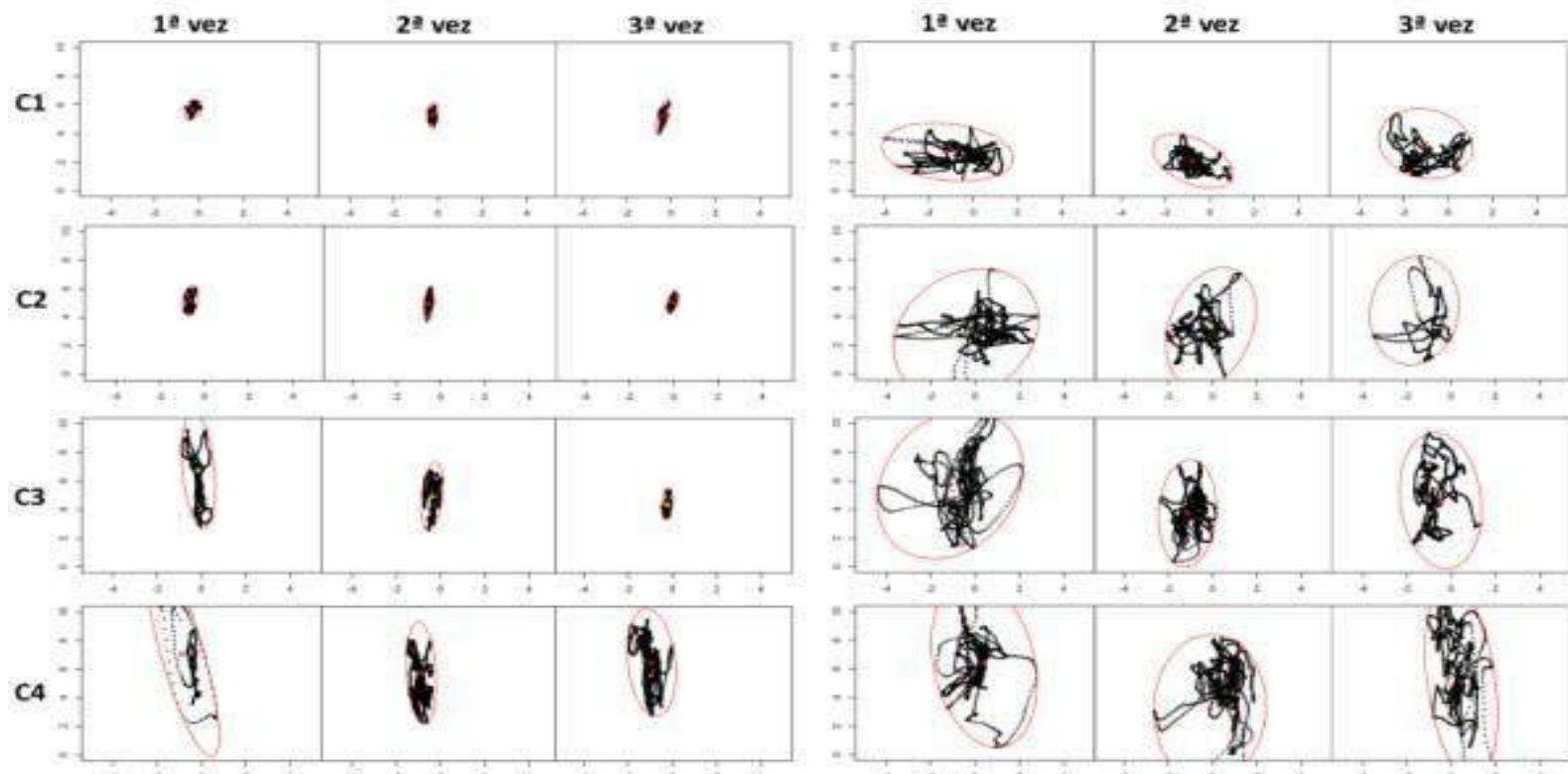
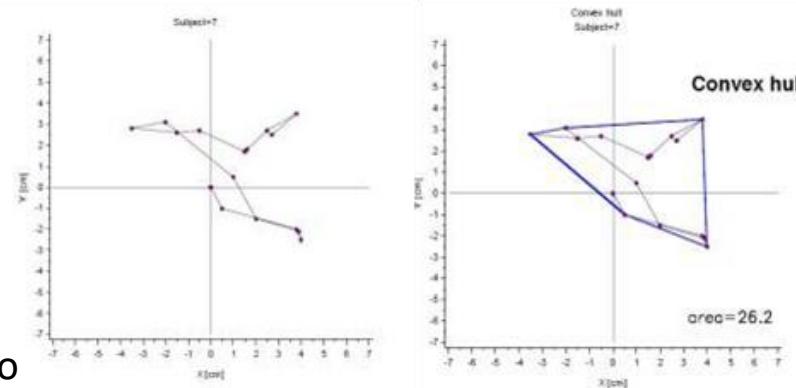
**Figura 4 – Classificação de incapacidade  
dos pacientes através da EDSS**



# Esclerose Múltipla 4

**Figura 5** - Construção do *convex hull* para uma trajetória

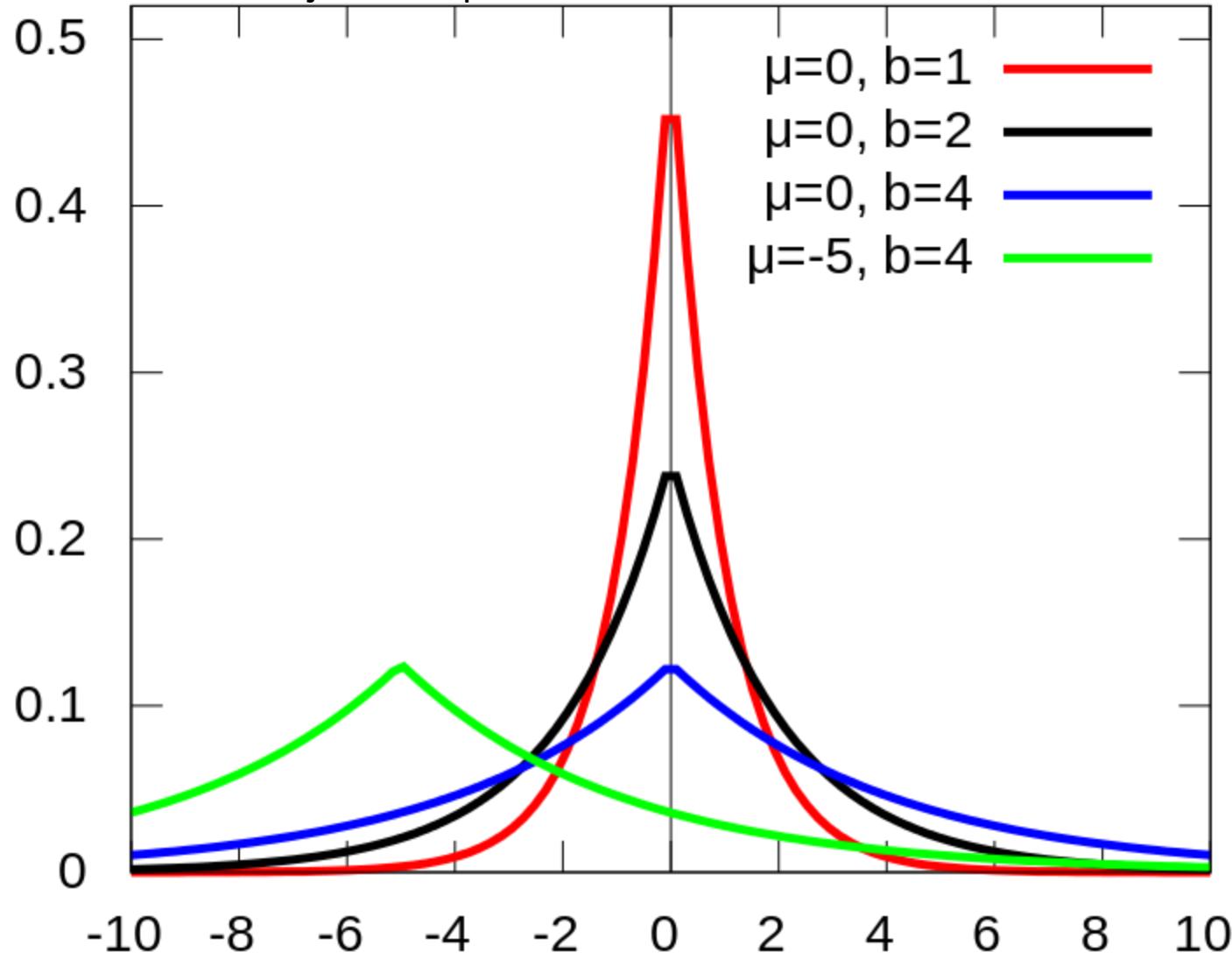
**Figura 7** - Exame posturo-gráfico de um paciente do GSem e um paciente do GQueixa utilizando a elipse.



# Nosso futuro neste problema: Sugestão

- Temos aqui um plano com os pontos de maior peso quando a máquina está trabalhando.
- Para a análise estatística consideramos a área das menores elipses ajustadas. Os interessados aprovaram nossas análises.
- Para um análise mais interessante gostaríamos de procurar pelo melhor modelo que pudesse se aproximar da geração dos dados.
- Mostraremos a seguir como uma composição de distribuições de Laplace poderia nos ajudar a ter uma boa análise estatística

Distribuição de Laplace



Densidade com média zero e  
parâmetro  $\beta$

$$f(x|\beta) = \frac{1}{2\beta} \exp\left(-\frac{|x|}{\beta}\right) = \begin{cases} \frac{1}{2\beta} \exp\left(\frac{x}{\beta}\right); & x \leq 0 \\ \frac{1}{2\beta} \exp\left(-\frac{x}{\beta}\right); & x > 0 \end{cases}$$

Nossa nova proposta é tentar definir para 3 parâmetros

$$f(x|\beta, \rho, \pi) = \begin{cases} \frac{\pi}{\beta} \exp\left(\frac{x}{\beta}\right); & x \leq 0 \\ 0 \leq \pi \leq 1 \\ \frac{1-\pi}{\rho} \exp -\left(\frac{x}{\rho}\right); & x > 0 \end{cases}$$

## Cobrindo o plano

**Consideremos 3 variáveis aleatórias independentes e com distribuição de Laplace Padrão (poderão ser a modificada)**

$$X, Y \text{ e } Z$$

Suponhamos que

$$\begin{aligned} X|\alpha &\approx \text{Lap}(0; \alpha); & Y|\beta &\approx \text{Lap}(0; \beta) ; \\ && \& \\ Z &\approx \text{Lap}(0; \delta) \end{aligned}$$

# Cobrindo o plano

Estas darão origem as variáveis de geração dos modelos

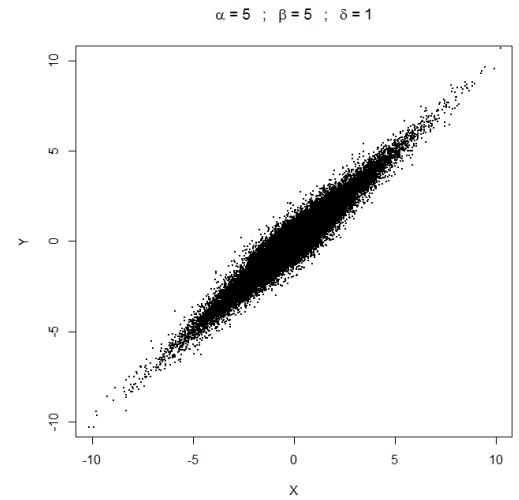
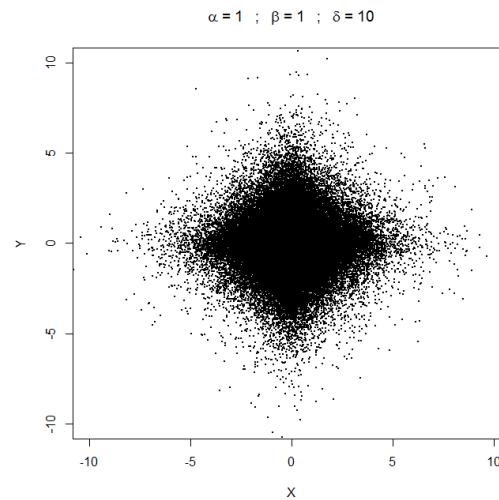
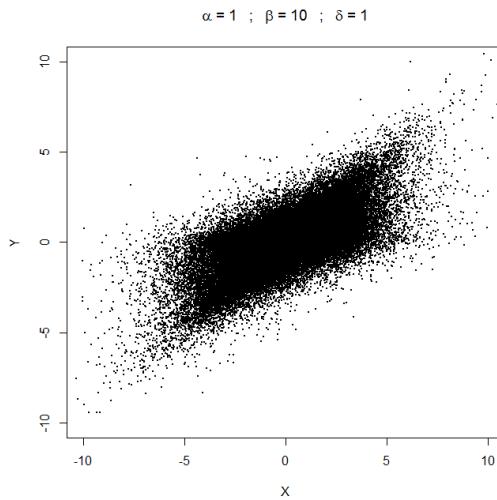
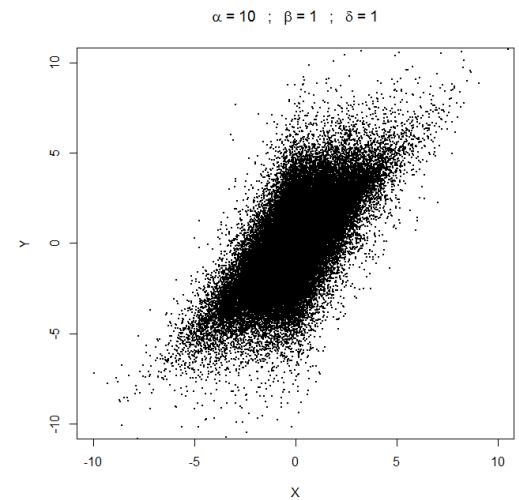
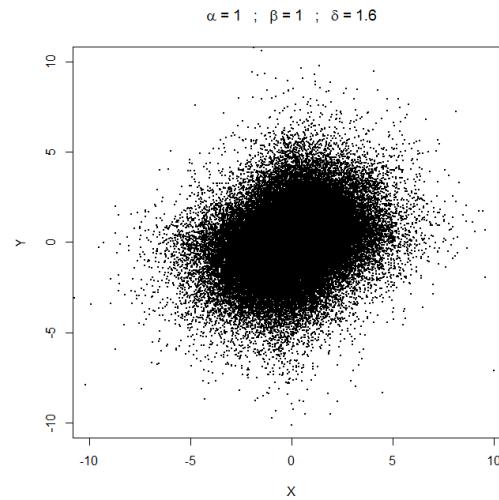
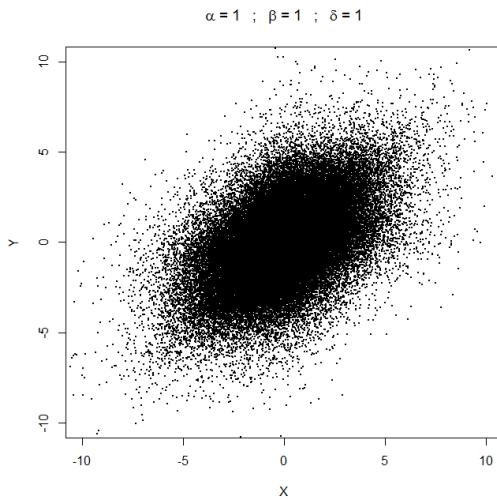
**Modelo 1:**  $U = X + Z$  e  $V = Y + Z$       quando  $\text{Cov}(U, V) > 0$

**Modelo 2:**  $U = \frac{X+Y+2Z}{\sqrt{2}}$  e  $V = \frac{X-Y}{\sqrt{2}}$       quando  $\text{Var}(U) > \text{Var}(V)$

**Modelo 3:**  $U = X + Z$  e  $V = Y - Z$       quando  $\text{Cov}(U, V) < 0$

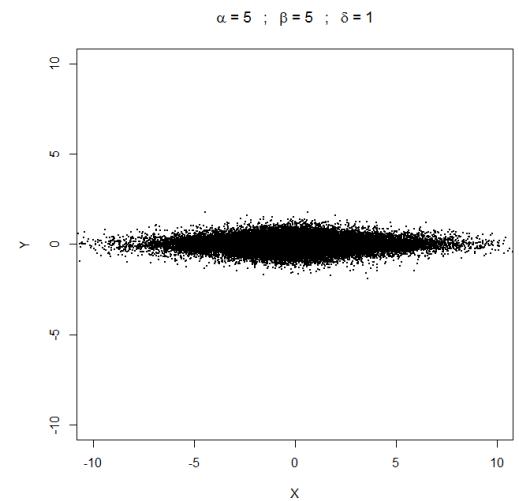
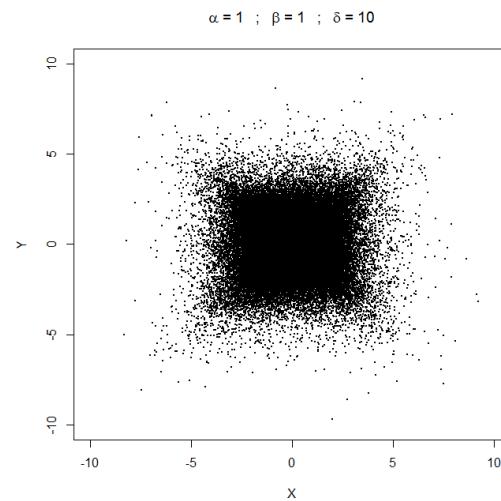
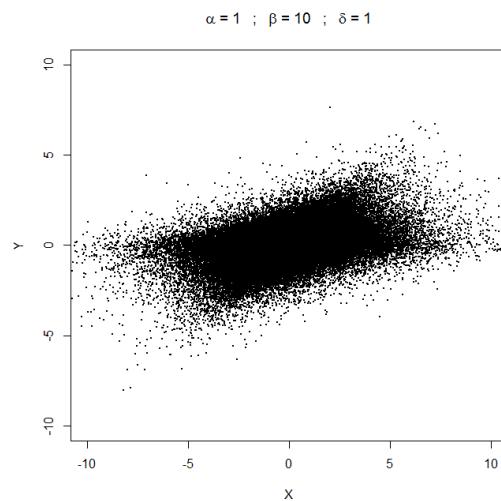
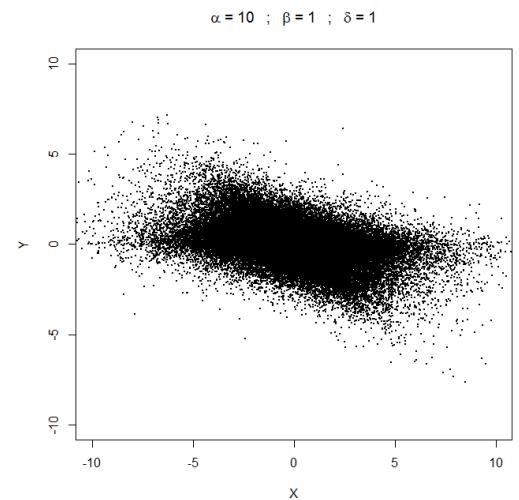
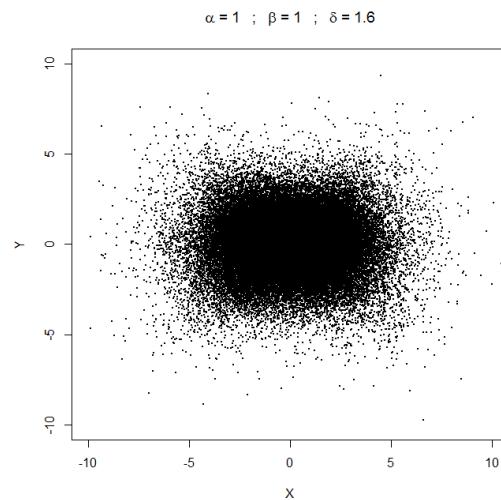
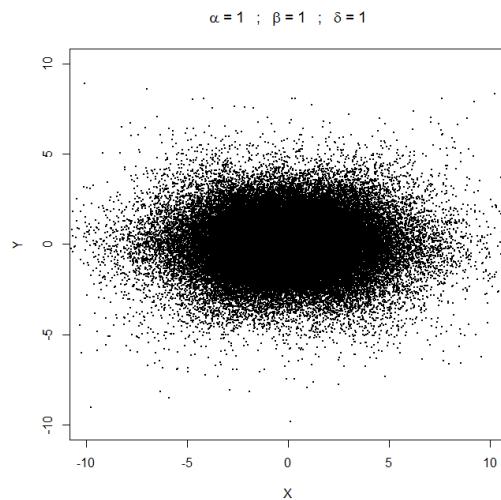
**Modelo 4:**  $U = \frac{X+Y}{\sqrt{2}}$  e  $V = \frac{X-Y+2Z}{\sqrt{2}}$       quando  $\text{Var}(U) < \text{Var}(V)$

# Modelo 1 (Cov positiva)

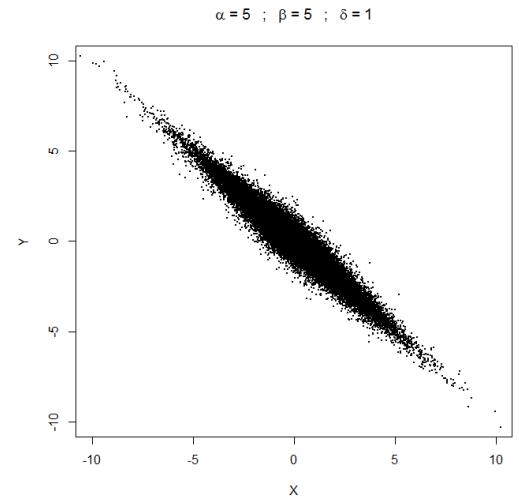
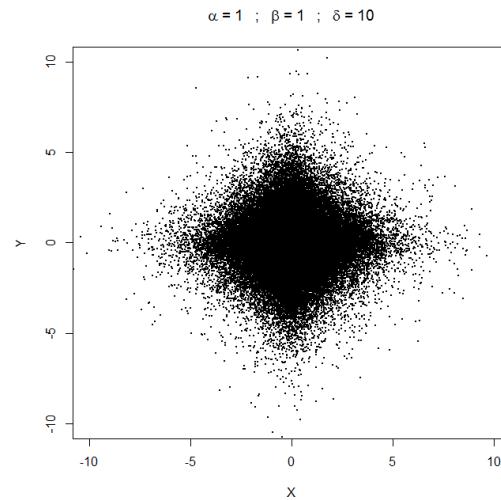
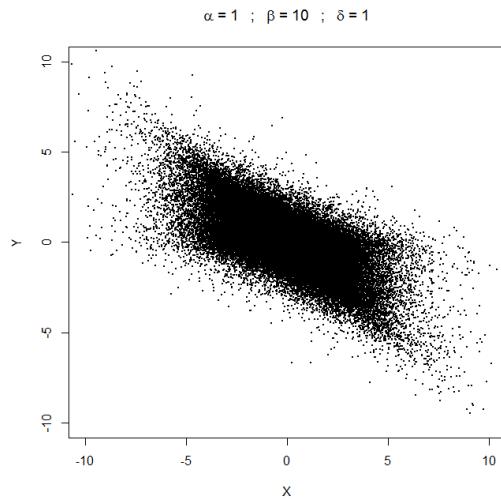
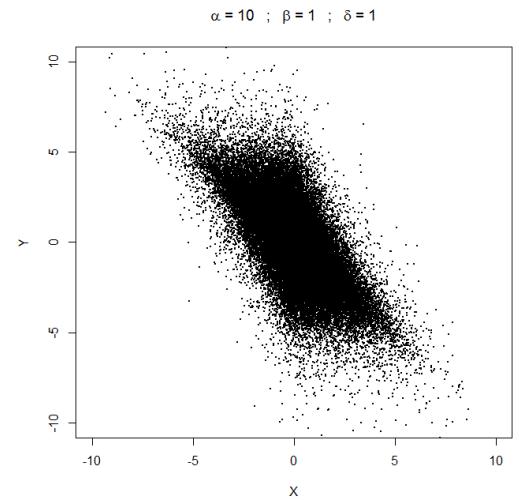
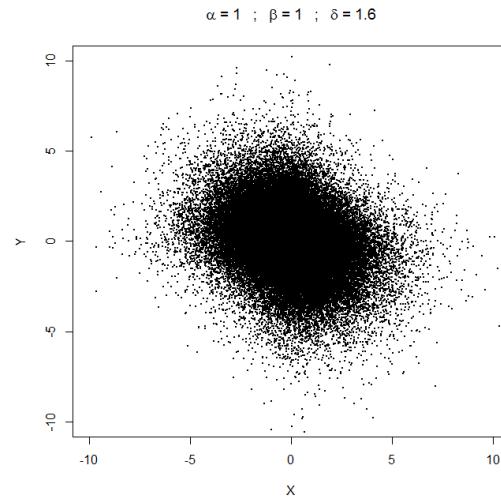
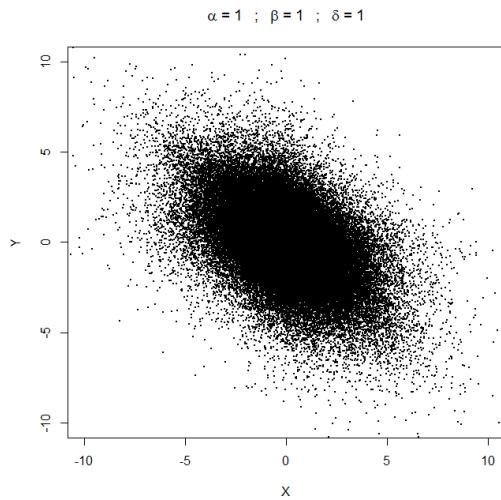


# Modelo 2

## (rotação 45° no Modelo 1)

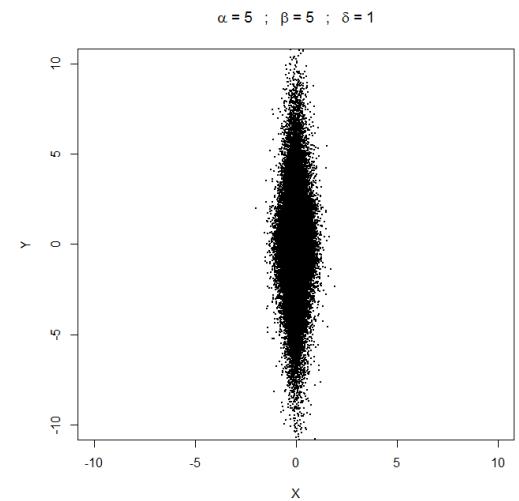
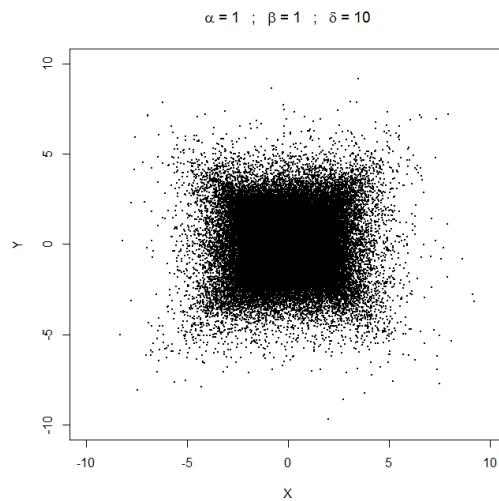
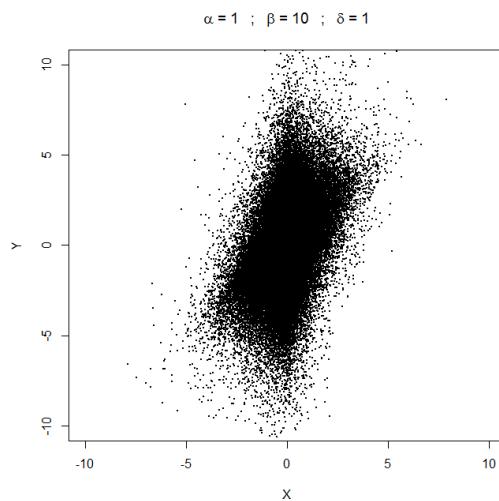
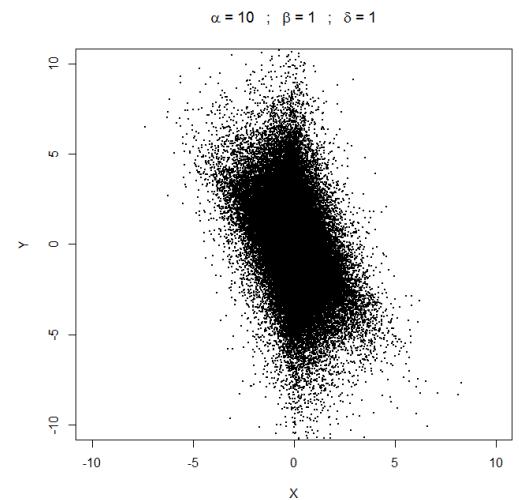
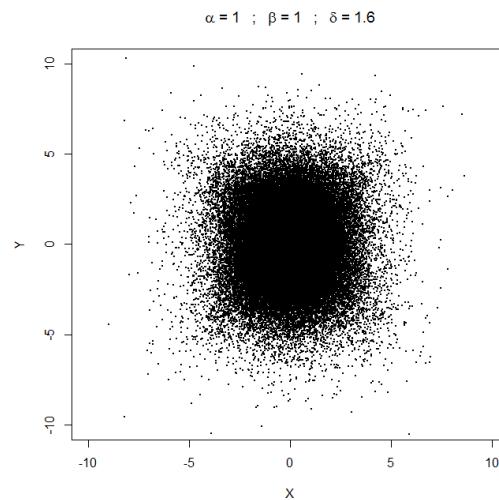
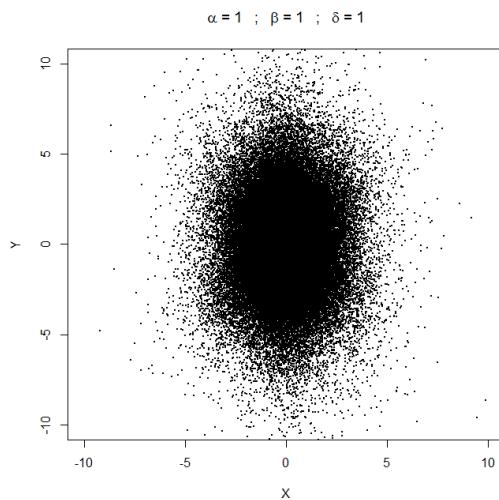


## Modelo 3 (Cov negativa)



# Modelo 4

## (rotação 45° no Modelo 3)



## Estimativas dos parâmetros dos modelos

**Modelo 1:**  $U = x + z$  e  $V = y + z$

$$Var(U) = Var(X) + Var(Z) = 2\alpha^2 + Cov(U, V) \Rightarrow \\ \hat{\alpha} = \sqrt{\frac{Var(U) - Cov(U, V)}{2}}$$

$$Var(U) = Var(Y + Z) = Var(Y) + Var(Z) = 2\beta^2 + Cov(U, V) \Rightarrow \\ \hat{\beta} = \sqrt{\frac{Var(U) - Cov(U, V)}{2}} =$$

$$Cov(U, V) = Cov(X, Y) + Cov(X, Z) + Cov(Y, Z) + Var(Z) = 2\delta^2 \Rightarrow$$

$$\hat{\delta} = \sqrt{\frac{Cov(U, V)}{2}}$$

**Restrição:**  $Cov(U, V) > 0$

## Estimativas dos parâmetros dos modelos

**Modelo 2:**  $U = \frac{X+Y+2Z}{\sqrt{2}}$  e  $V = \frac{X-Y}{\sqrt{2}}$

$$Var(U) = \frac{1}{2}[Var(X + Y + 2Z)] = \frac{Var(X) + Var(Y) + 4Var(Z)}{2} = \alpha^2 + \beta^2 + 4\delta^2$$

$$Var(V) = \frac{1}{2}Var(X - Y) = \frac{Var(X) + Var(Y)}{2} = \alpha^2 + \beta^2$$

$$Cov(U, V) = \frac{Cov(X+Y+2Z; X-Y)}{2} = \frac{Var(X) - Var(Y)}{2} = \alpha^2 - \beta^2$$

$$\hat{\alpha} = \sqrt{\frac{Var(V) + Cov(U, V)}{2}} \quad \hat{\beta} = \sqrt{\frac{Var(V) - Cov(U; V)}{2}} \quad \hat{\delta} = \sqrt{\frac{Var(U) - Var(V)}{2}}$$

**Restrição:**  $Cov(U+V, U-V) > 0 \rightarrow Var(U) > Var(V)$

## Estimativas dos parâmetros dos modelos

**Modelo 3:**  $U = X + Z$  e  $V = Y - Z$

$$Var(U) = Var(X) + Var(Z) = 2\alpha^2 + Cov(U, V) \Rightarrow$$

$$\alpha^2 = \sqrt{\frac{Var(U) + Cov(U, V)}{2}}$$

$$Var(V) = Var(Y - Z) = Var(Y) + Var(Z) = 2\beta^2 + Cov(U, V) \Rightarrow$$

$$\hat{\beta} = \sqrt{\frac{Var(V) + Cov(U, V)}{2}}$$

$$Cov(U, V) = Cov(X, Y) - Cov(X, Z) + Cov(Y, Z) - Var(Z) = -Var(Z) = -2\delta^2$$

$$\hat{\delta} = \sqrt{-\frac{Cov(U, V)}{2}}$$

**Restrição:**  $Cov(U, V) < 0$

## Estimativas dos parâmetros dos modelos

$$\textbf{Modelo 4: } U = \frac{X+Y}{\sqrt{2}} \quad \text{e} \quad V = \frac{X-Y+2Z}{\sqrt{2}}$$

$$Var(U) = \frac{Var(X+Y)}{2} = \frac{Var(X)+Var(Y)}{2} = \alpha^2 + \beta^2$$

$$Var(V) = \frac{Var(X-Y+2Z)}{2} = \frac{Var(X)+Var(Y)+4Var(Z)}{2} = \alpha^2 + \beta^2 + 4\delta^2$$

$$Cov(U, V) = \frac{Cov(X+Y; X-Y+2Z)}{2} = \frac{Var(X)-Var(Y)}{2} = \alpha^2 - \beta^2$$

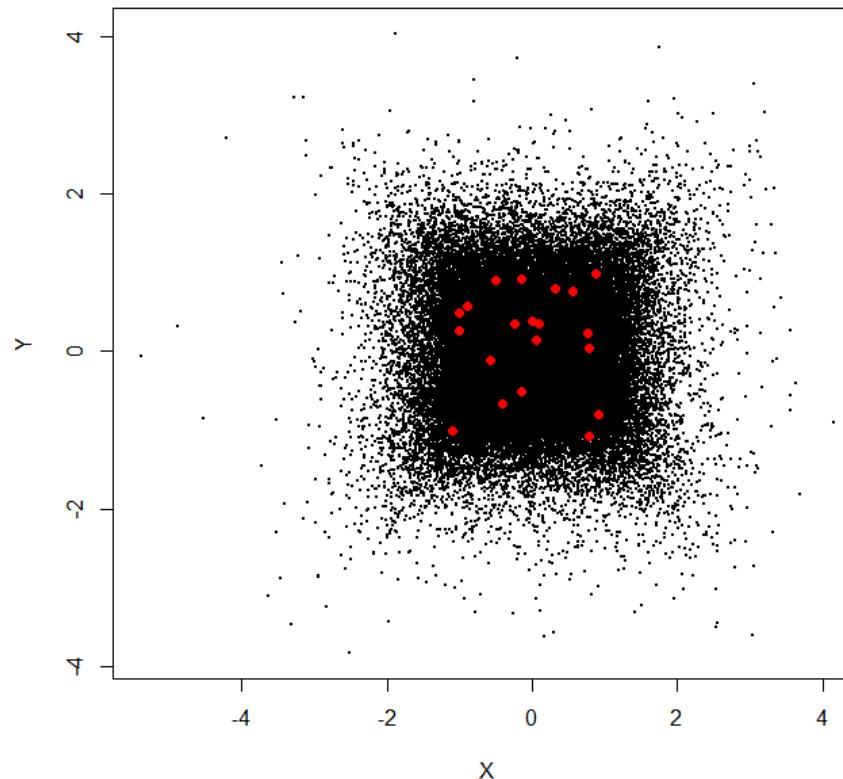
$$\hat{\alpha} = \sqrt{\frac{Var(U) + Cov(U; V)}{2}} \quad \hat{\beta} = \sqrt{\frac{Var(U) - Cov(U; Z)}{2}} \quad \hat{\delta} = \frac{\sqrt{Var(V) - Var(U)}}{2}$$

**Restrição:**  $Cov(U+V, U-V) < 0 \rightarrow Var(U) < Var(V)$

## Exemplo do QUADRADO – Modelo 2 ( var[U]>var[V] )

Sample number	QUADRADO		
	x	y	xy
1	-0,500	0,903	-0,452
2	0,554	0,765	0,424
3	-0,244	0,347	-0,085
4	0,754	0,236	0,178
5	0,875	0,988	0,864
6	-1,009	0,268	-0,271
7	0,007	0,381	0,002
8	0,774	0,031	0,024
9	0,056	0,138	0,008
10	-0,144	-0,510	0,073
11	0,909	-0,812	-0,738
12	-0,151	0,921	-0,139
13	0,321	0,792	0,255
14	0,100	0,343	0,034
15	-0,889	0,565	-0,502
16	0,789	-1,072	-0,846
17	-0,587	-0,110	0,064
18	-1,017	0,487	-0,495
19	-0,415	-0,665	0,276
20	-1,094	-1,015	1,110
Stat	<b>-0,046</b>	<b>0,149</b>	<b>-0,011</b>
Var	<b>0,434</b>	<b>0,403</b>	<b>0,220</b>
Cov			<b>-0,004</b>
Área			

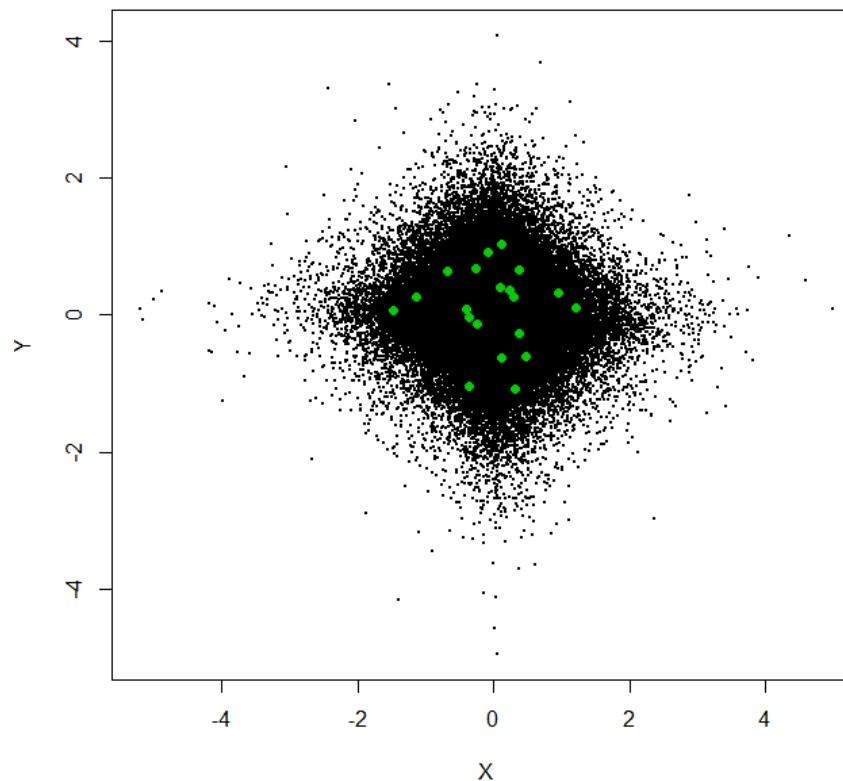
$$\hat{\alpha} = \sqrt{\frac{Var(V) + Cov(U,V)}{2}} = 0,4637 \quad \hat{\beta} = \sqrt{\frac{Var(V) - Cov(U,V)}{2}} = 0,4511 \quad \hat{\delta} = \sqrt{\frac{Var(U) - Var(V)}{2}} = 0,0880$$



## Exemplo do DIAMANTE – Modelo 3 ( Cov[U,V]<0 )

Sample number	DIAMANTE		
	x	y	xy
1	0,951	0,326	0,310
2	0,466	-0,597	-0,278
3	0,124	-0,614	-0,076
4	1,204	0,112	0,135
5	0,232	0,355	0,082
6	-0,267	0,676	-0,181
7	-0,403	0,092	-0,037
8	0,294	0,268	0,079
9	0,384	-0,260	-0,100
10	0,306	-1,079	-0,330
11	-0,680	0,636	-0,433
12	-0,077	0,906	-0,070
13	-0,365	-1,032	0,377
14	-1,147	0,264	-0,303
15	0,103	0,411	0,042
16	-0,237	-0,131	0,031
17	-0,354	-0,038	0,013
18	0,384	0,659	0,253
19	0,118	1,026	0,121
20	-1,485	0,071	-0,105
Stat	<b>-0,023</b>	<b>0,103</b>	<b>-0,023</b>
Var	<b>0,383</b>	<b>0,329</b>	<b>0,044</b>
Cov			<b>-0,021</b>
Área			

$$\hat{\alpha} = \sqrt{\frac{Var(U) + Cov(U,V)}{2}} = 0,4254 \quad \hat{\beta} = \sqrt{\frac{Var(V) + Cov(U,V)}{2}} = 0,3924 \quad \hat{\delta} = \sqrt{-\frac{Cov(U,V)}{2}} = 0,1025$$



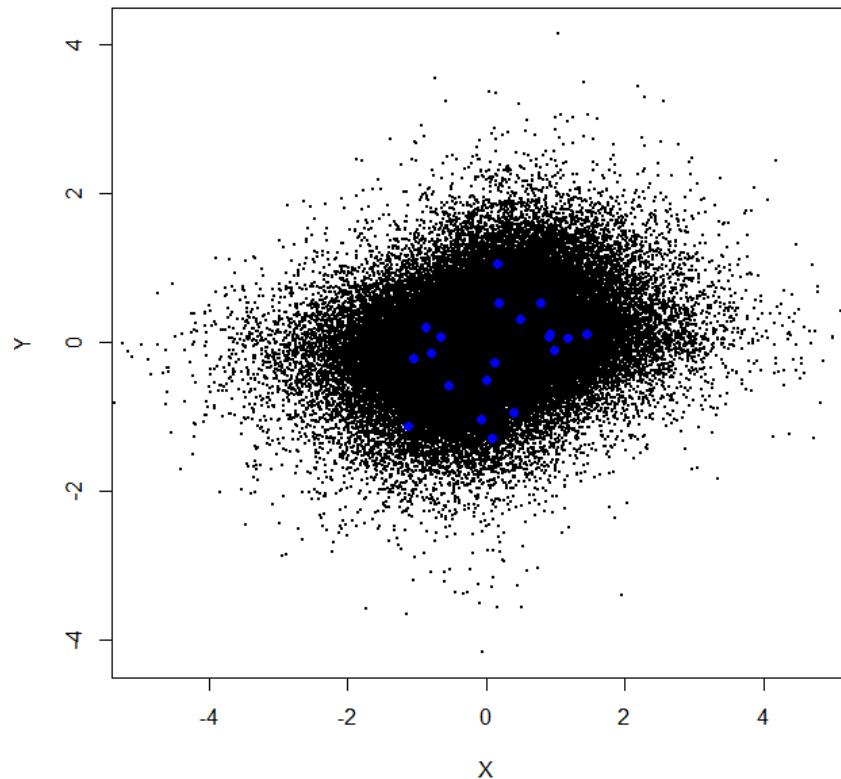
## Exemplo do CÍRCULO – Modelo 1 ( Cov[U,V]>0 )

Sample number	CÍRCULO		
	x	y	xy
1	0,153	1,044	0,160
2	0,914	0,071	0,065
3	-0,664	0,068	-0,045
4	-1,040	-0,214	0,223
5	0,001	-0,505	0,000
6	0,792	0,531	0,420
7	1,183	0,056	0,066
8	0,399	-0,942	-0,376
9	0,180	0,524	0,095
10	-0,546	-0,588	0,321
11	0,124	-0,269	-0,033
12	0,915	0,108	0,099
13	0,088	-1,289	-0,114
14	-0,068	-1,032	0,071
15	0,494	0,317	0,157
16	-0,786	-0,139	0,109
17	0,978	-0,113	-0,110
18	-0,870	0,190	-0,165
19	-1,123	-1,123	1,261
20	1,458	0,105	0,153
Stat	0,129	-0,160	0,118
Var	0,569	0,347	0,098
Cov			0,138
Área			

$$\hat{\alpha} = \sqrt{\frac{Var(U) - Cov(U,V)}{2}} = 0,4642$$

$$\hat{\beta} = \sqrt{\frac{Var(V) - Cov(U,V)}{2}} = 0,3233$$

$$\hat{\delta} = \sqrt{\frac{Cov(X,Y)}{2}} = 0,2627$$



# Ajuste do modelo em dados de um paciente

