

Article

Numerical simulation and FRAP experiments show that the plasma membrane binding protein PH-EFA6 does not exhibit anomalous sub-diffusion in cells.

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Abstract: FRAP technique have been used for decades to measure movements of molecules in 2D. Data obtained by FRAP experiments in cell plasma membranes are assumed to be described through means of two parameters, a diffusion coefficient D (as defined in a pure Brownian model) and a mobile fraction M . Nevertheless, it has also been shown that recoveries can be nicely fit using anomalous sub-diffusion. FRAP at variable radii has been developed using the Brownian diffusion model to access geometrical characteristics of the surrounding landscape of the molecule. Here we performed numerical simulations of continuous time random walk (CTRW) anomalous subdiffusion and interpreted them in the context of variable radii FRAP. These simulations were compared to experimental data obtained at variable radii on living cells using the PH domain of the membrane binding protein EFA6 (exchange factor for ARF6, a small G protein). This protein domain is an excellent candidate to explore the structure of the interface between cytosol and plasma membrane in cells. By direct comparison of our numerical simulations to the experiments, we show that this protein does not exhibit anomalous diffusion in BHK cells. The non Brownian PH-EFA6 dynamics observed here is more related to spatial heterogeneities such as cytoskeleton fences effects.

Keywords: Anomalous diffusion; FRAP; Numerical Simulations; PH - Domain, Membrane Binding

16 1. Introduction

17 Early models of the plasma membrane, notably the fluid mosaic model [1] postulated that
18 transmembrane proteins were freely diffusing in a sea of lipids. During these two last decades, it
19 has become apparent that cell surface membranes are far from being homogeneous mixture of their
20 lipid and protein components. They are compartmented into domains whose composition, physical
21 properties and function are different. Numerous studies on transmembrane proteins and plasma
22 membrane lipids by means of single particle tracking (SPT), Fluorescence Correlation Spectroscopy
23 (FCS) or fluorescence recovery after photobleaching (FRAP) has shown the existence of micrometer and
24 nanometer size domains on both model membrane [2,3] and living cells[4–6]. In the plasma membrane
25 of living cells, these domains can come from different origins, but are generally classified into two
26 main groups :

- 27 • "rafts" model where lipid/lipid phase separation drives the lateral partitioning of transmembrane
28 proteins. [7]
- 29 • "cytoskeleton fence" model in which transmembrane proteins are coralled by a fence of
30 cytoskeleton just beneath the plasma membrane. [8,9]

31 Variable radii FRAP first [5,6], then spot variation FCS [10–12] helped in discriminate amongst
32 these two models the nature of the deviation to pure Brownian diffusion of membrane components in
33 living cells.

34 FRAP experiments have been used for determination of long-range molecular diffusion of proteins
35 and lipids on both model system and cells for more than 30 years [13,14]. Briefly, fluorescently labelled
36 molecules localized within a predefined area are irreversibly photo-destroyed by a short and intense
37 laser pulse. The recovery of the fluorescence in this area is then measured against time. Since
38 no reversible photoreaction occurs, recovery of the fluorescence in the photobleached area is due
39 to diffusion. FRAP data are generally interpreted by assuming classical Brownian diffusion. Two
40 parameters can then be obtained : D , the lateral diffusion coefficient and M , the mobile fraction of the
41 diffusing molecule. When the radius of the photobleached area is small compared to the diffusion area,
42 M must be equal to 1 for freely diffusing species. In fact, most of the data reported so far in biological
43 membranes for transmembrane proteins shows a value of $M < 1$. This lack in total fluorescence
44 recovery can be interpreted as a restriction to free-diffusion behaviour. Parameters obtained have then
45 to be re-evaluated to recognize the effect of time-dependent interactions in a field of random energy
46 barriers.

47 An experimental approach to that question has been proposed by Feder *et al.* [15] by introducing
48 anomalous (sub)diffusion in the motion of transmembrane proteins. Many sources of motion restriction
49 can lead to anomalous diffusion (for review see [16] and [17]). Saxton has performed extensive
50 numerical simulations to help in identifying the sources of anomalous diffusion (obstacles, binding...)
51 using SPT measurements [18,19] and he declined this more recently to FRAP experiments [20] using
52 fractional Brownian motion (fBm) or continuous time random walk (CTRW) models as sources of
53 anomalous diffusion.

54 Membrane bound proteins should also be submitted to several interactions with their surrounding
55 environment that should account for an anomalous subdiffusion behaviour. Sources of deviation from
56 Brownian motion in their lateral diffusion may include lipid domains trapping, binding to immobile
57 proteins and/or obstruction by cytoskeletal elements. This different possible interactions can exhibit
58 different characteristic times, or different distributions of characteristic times. Here, diffusion of an
59 intracellular membrane-bound protein domain (pleckstrin homology domain of EFA6, the ARF6
60 exchange factor) has been analysed inside living cells by FRAP experiments. Previous studies have
61 shown that these proteins are linked to the polar head of $PI(4,5)P_2$ lipids by means of electrostatic
62 interactions[21]. Furthermore, the protein used in this study appears to have a functional requirement
63 to be associated to the plasma membrane within cells [22,23]

64 In this paper, numerical simulation of the CTRW model of anomalous sub-diffusion were first
65 performed at a single spot size. Based on the quality of the fit using different analytical expression,
66 we tested the ability to retrieve this anomalous diffusion in the simulated recovery curves first and in
67 the experimental one afterwards. We showed that performing FRAP experiments at a single spot size
68 did not allow to discriminate between the CTRW induced anomalous diffusion case and the empirical
69 classical approach using mobile and immobile fraction.

70 We then computed and performed experimental FRAP at variable radii. By plotting changes in
71 the anomaly of the diffusion or in the mobile fraction as a function of the inverse of the bleached
72 radius, as in Salomé *et al.* [5] we showed that it was possible to discriminate between the two models.
73 Interestingly we observed that the restriction to the mobility of the PH-EFA6 domain is not due to
74 CTRW anomalous sub diffusion, but more certainly to the sub cortical actin fences.

75 2. Results

76 2.1. Anomalous sub-diffusion Modeling

A way to describe the continuous time random walk sub-diffusion is to start from a two dimensional random diffusion process. A particle walks from trap to trap and spend a certain (random) time in each trap. It is characterized by the following operation :

$$\mathbf{r} \rightarrow \mathbf{r} + \Delta; t \rightarrow t + \tau \quad (1)$$

77 \mathbf{r} and t are respectively the two dimensional position and the age of the particle, where Δ is a two
78 dimensional random (Gaussian) variable with variance $v = 2D$, and τ is the (random) time the particle
79 spend into the trap.

80 In our model, the particle is supposed to diffuse very rapidly between two traps. This travel time
81 is therefore neglected (this, because it was not experimentally accessible). The time τ the particle stays
82 in a trap is supposed to have very strong fluctuations, this give rise to anomalous diffusion pattern.

As an example a generic distribution is used which leads, after a while, to a standard Levy law in time :

$$P_0(\tau) = \frac{\alpha}{(1 + \tau)^{\alpha+1}} \quad (2)$$

83 This distribution have been used in the same type of context by Naggle [24].

The Levy exponent α is the characteristic exponent of subdiffusive behaviour. For long times we have :

$$\langle r^2(t) \rangle \propto t^\alpha \quad (3)$$

84 When $\alpha < 1$ a spatio-temporal Fourier (Laplace) analysis leads to the following asymptotic (ω, k
85 $\rightarrow 0$) Green function :

$$\tilde{g}(\mathbf{k}, \omega) = \frac{1}{\omega(D_\alpha k^2 \omega^{-\alpha} + 1)} ; D_\alpha = D/\Gamma(1 - \alpha) \quad (4)$$

86 where ω and \mathbf{k} are respectively the conjugate variables of position \mathbf{r} and time t , where $k = |\mathbf{k}|$.
87 Notice that the solution of the inverse Laplace transform is a function of the variable $k^2 t^\alpha$. It follows
88 that the Green function is a function of the variable $x = r^2/t^\alpha$. When x is high enough one can perform
89 an approximate inverse transformation via a saddle point method :

$$g(\mathbf{r}, t) \propto \exp(-cst x^\nu) ; \nu = \frac{1}{2 - \alpha}, cst : \text{a known constant} \quad (5)$$

Notice that the exponent ν interpolate nicely between the gaussian case ($\alpha = 1$) and the exponential case. The general solution of this type of anomalous diffusion process is then :

$$\rho(\mathbf{r}, t) = \int \rho_0(\mathbf{r}' - \mathbf{r}) g(x(\mathbf{r}', t)) d^2 \mathbf{r}' \quad (6)$$

90 where ρ is the probability density to find the particle at the point \mathbf{r} at instant t and ρ_0 is the initial
91 state.

92 As the Green function is a bell-shaped fast decreasing function, one approximate it by a Gaussian
93 shape with the exact dispersion, $D_\alpha = D \sin(\pi\alpha)/(\pi\alpha)$, which can be calculated from eq.4. This
94 permits to construct an analytical expression of the fluorescence recovery using standard properties of
95 Gaussian functions.

Starting from Axelrod [13] initial density as it is immediately after a Gaussian laser beam profile extinction indeed :

$$\rho_0(\mathbf{r}) = \exp(-K \exp(-2 \frac{\mathbf{r}^2}{R^2})) \quad (7)$$

96 (K=photobleaching constant, depending on experimental conditions [13]) and using the standard
97 properties of the Gaussian shape in the convolution operation, one can obtain the time evolved result
98 as a series.

99 Once integrated upon a disk of radius R , and, after normalization to the surface of the disk, one
100 obtain the FRAP signal :

$$I_R(t) = 1 + \sum_1^{\infty} \frac{(-K)^n}{n!} \frac{1}{2n} \left(1 - \exp \left(-\frac{2nR^2}{R^2 + 4nD_\alpha t^\alpha} \right) \right) \quad (8)$$

101 This function will be used to fit experimental data.

102

103 Systematic corrections of this procedure are determined using numerical Monte-Carlo simulations
104 of the fluorescence recoveries, using known α and $D = D_\alpha \frac{(\pi\alpha)}{\sin(\pi\alpha)}$

105 In order to keep in our calculation the finite size effects, the simulation were made in a ring of a
106 radius of 30 arbitrary unit (a.u.) length explored by 10^7 particles for each recovery. Radii varying from
107 0.5 to 3 a.u. were photo-destroyed during the simulation. Reflective type of boundary conditions
108 were used. This means that when a particle gets out of the 30 a.u. radius it is re-introduced in the
109 same direction at a small distance of the boundary. See Appendix A for examples of recovery curves
110 generated numerically by this approach.

111 2.2. Validating numerical simulation and analytical models

112 In order to verify the validity of our analytical model, a set of numerically simulated recovery
113 curves using anomalous diffusion as a model has been fitted with equation 8. Each parameter (α and
114 D) were tested. Figure 1a shows the value of D obtained after fit of the numerical (D_{output}) simulation
115 using given D (D_{input}) for the three different α tested above. Figure 1b illustrates the variation of fitted
116 α as a function of D used in the simulation, for $\alpha = 0.6$ (red); $\alpha = 0.7$ (green); $\alpha = 0.8$ (blue). This clearly
117 shows that both parameters (α, D) are always underestimated when fitting with an analytical model
118 the numerically simulated fluorescence recoveries.

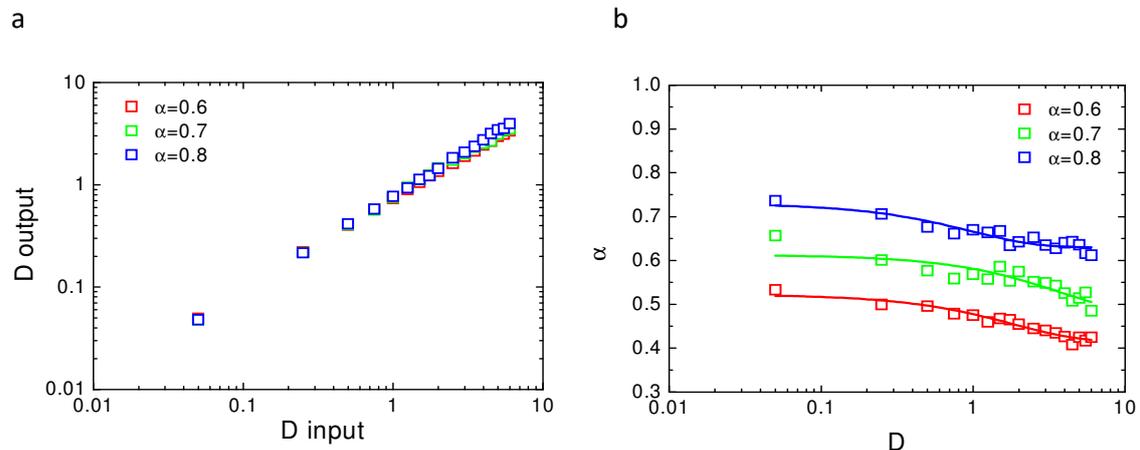


Figure 1. Values of the parameters (D, α) obtained from the fit of the simulated recoveries a- D values obtained after fit of the fluorescence recovery with eq.8 (D_{output}) as a function of D values used in the numerical simulation (D_{input}) for different α . Note that the slope is always less than 1 b- Values of α obtained after fit of the simulated curves with eq. 8 for different α used in the simulation and as a function of the D values used in the simulation. Note that the original α value used in the simulation is never reached by fit of the simulated recoveries.

119 This is mainly due to the finite size and finite time effect of our numerical simulations and
120 paradoxically is also nicely mimicking what could occur experimentally in a finite size cell reservoir.

121 2.3. Challenging analytical models to identify numerically simulated anomalous diffusion fluorescence recoveries

122 Fluorescence recoveries have been numerically simulated using CTRW anomalous diffusion as
123 the model of molecular motion. These curves were then fitted with three different analytical expression
124 of FRAP recoveries, each being specific of a diffusion model :

- 125 • Anomalous diffusion motion (aDm): see eq. 8 in section 2.1
- 126 • Free Brownian motion (Bm) :

$$I_R(t) = \sum_1^{\infty} \frac{(-K)^n}{n!} \frac{1}{1+n+\frac{8nDt}{R^2}} \quad (9)$$

- Restricted Brownian motion (rBm) :

$$I_R(t) = (1-M) \frac{1-e^{-K}}{K} + M \sum_1^{\infty} \frac{(-K)^n}{n!} \frac{1}{1+n+\frac{8nDt}{R^2}} \quad (10)$$

127 where M accounts for the mobile fraction.

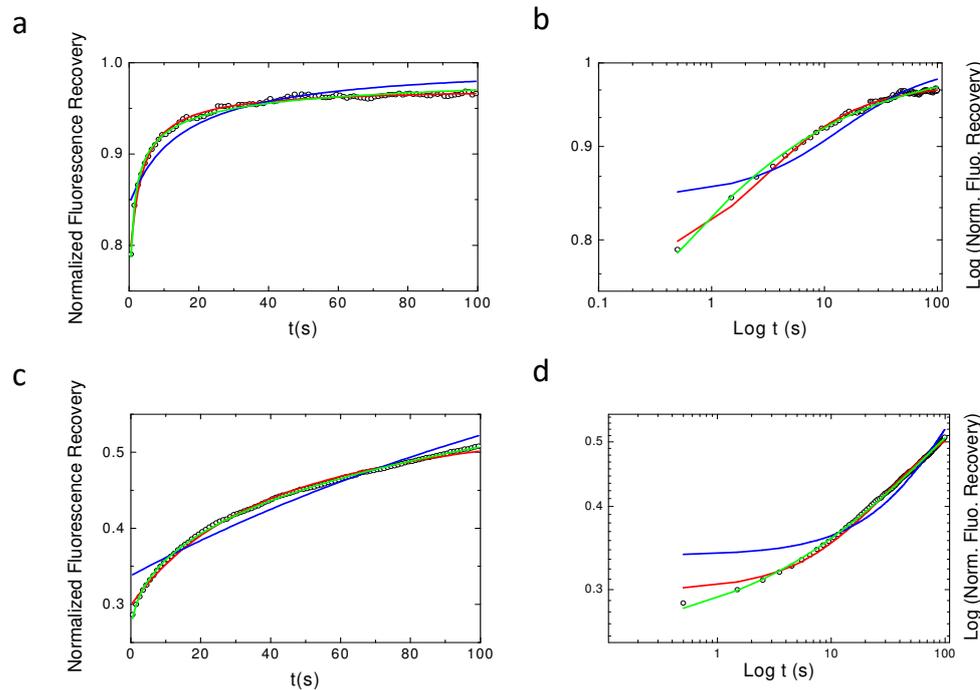


Figure 2. Best fits using the different models of Normal and Log-Log plots of simulated CTRW anomalous sub-diffusion recoveries. $\alpha = 0.6$ is the value used for the simulation in the four plots. In blue, fBm model (eq. 9), in red rBm model (eq.10) and in green aDm (eq.8) fits. In **a & b** $D_\alpha=2$ and in **c & d** $D_\alpha=0.1$. **a & c** are the normal plots while **b & d** are the log-log plots. From these graph it can be seen that one can hardly distinguish between the rBm model (red) and the aDm model (green) fits of the simulated recovery.

128 Figure 2 shows the obtained results for the three tested models (aDm in green, fBm in blue, rBm
 129 in red) with two different values of D_α and with $\alpha=0.6$. It can easily be seen that the fBm does not
 130 fit to the curve, as expected, but surprisingly it can also be seen that aDm and rBm models fit quasi
 131 equivalently the numerical simulation. Even log-log plot (Fig. 2 b and d) hardly allows to directly
 132 separate the two-models. Still log-log plots show that these models can be discriminated at short times
 133 ($t \ll \tau_c$, τ_c being the characteristic half-time of the recovery) and at very long times ($t \gg \tau_c$).

134 2.4. Single spot FRAP does not allow to identify the nature of PH-EFA6 diffusion in cells.

135 FRAP experiments have been performed on 15 different BHK cells (3 recoveries per cells on
 136 average) expressing the PH domain of EFA6 linked to the GFP. These data have been acquired at a
 137 given radius using the 63x objective (see experimental section for explanation). EFA 6, an exchange
 138 factor for ARF 6 (a small G protein) has recently been described as being located on the internal part of
 139 the plasma membrane, with its PH domain responsible for the interaction with lipids [23].

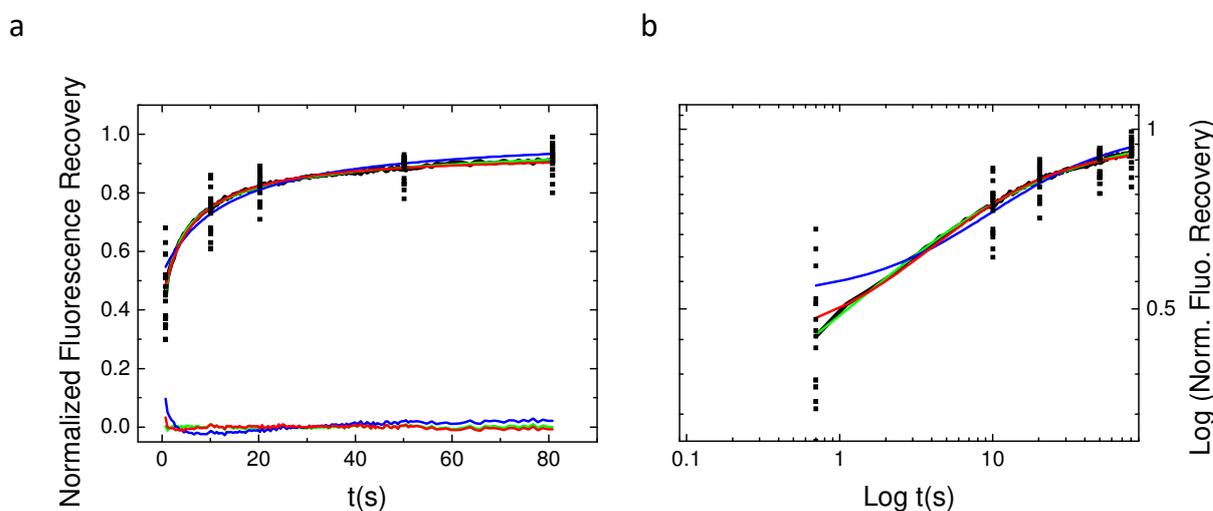


Figure 3. Average experimental recovery curve fitted by the three diffusion models. The three models (bM, rBm and aDm) were used to fit the average recovery curve obtained from 45 different experiments. The normal (part a) plot shows the residual from the fit of the different models. Note that only the bM model fit is inaccurate. The Log-Log plot (part b) illustrate again the difficulty to discriminate between the aDm and the rBm model in the goodness of the fit.

140 Figure3 shows the average fluorescence recovery (black line, mean of the 45 recoveries) as well as
 141 some points extracted from of the 45 different recoveries in order to illustrate the discrepancy observed
 142 when working with cells. This mean fluorescence recovery has been fitted by the three different models
 143 used in the previous section. On the bottom of Fig. 3a is depicted the differences between the fit and the
 144 observed fluorescence ($F_f - F_o$) in order to illustrate the quality of the fit. From figure 3a (normal plot)
 145 and figure 3b (log-log plot) it can easily be seen that except for the free Brownian motion model(Bm),
 146 the nature of the diffusion of PH-EFA6 cannot be discriminated between restricted Brownian motion
 147 (rBm) and anomalous diffusion (aDm). This is confirmed by a χ^2 statistical test to probe the quality
 148 of the fit as shown in table 1. Table 1 also summarized the average values of the set of parameters
 149 (D, M, D_α, α) that can be extracted from the different fits.

Table 1. Parameters values obtained by the fit of the average experimental recovery with the different analytical models

Model	$D(\mu\text{m.s}^{-1})$	M	α	$D_\alpha(\mu\text{m.s}^{-\alpha})$	χ^2
Bm	0.12 ± 0.06	1	-	-	5.7 ± 0.5
rBm	0.22 ± 0.01	0.92 ± 0.01	-	-	3.8 ± 1.6
aDm	-	-	0.65 ± 0.02	1.48 ± 0.05	2.9 ± 1.6

150 2.5. Variable radii FRAP allows correct estimation of the anomalous sub-diffusion exponent α

151 Previous results using direct analysis on both numerically simulated recoveries and experimental
 152 recoveries clearly showed that : i) Parameters α and D_α) of the aDm model were always underestimated
 153 ii) aDm and rBm models could only be discriminated at short and long times. Nevertheless, since time
 154 and space are correlated in diffusion and since experimental time-scale is finite, variable radii FRAP
 155 experiments were firstly numerically simulated and performed after on cells (see experimental section

156 for explanation). Each parameter (α and D_α) were estimated again, by fitting simulated recoveries at
 157 different radius (see Monte-Carlo simulation section for explanation) with our analytical model (eq. 8).
 158 Figure 4 shows the behaviour of fitted α as a function of $1/R$. Theory of anomalous diffusion processes
 159 predicts that α is space-invariant in a “homogeneously heterogeneous” environment. It can be directly
 160 seen on the plot that this is not what our data suggest, but on the opposite they showed that a linear
 161 dependence of α as a function of $1/R$ (at least for $R > 1$ a.u.) with a negative slope is observed for the

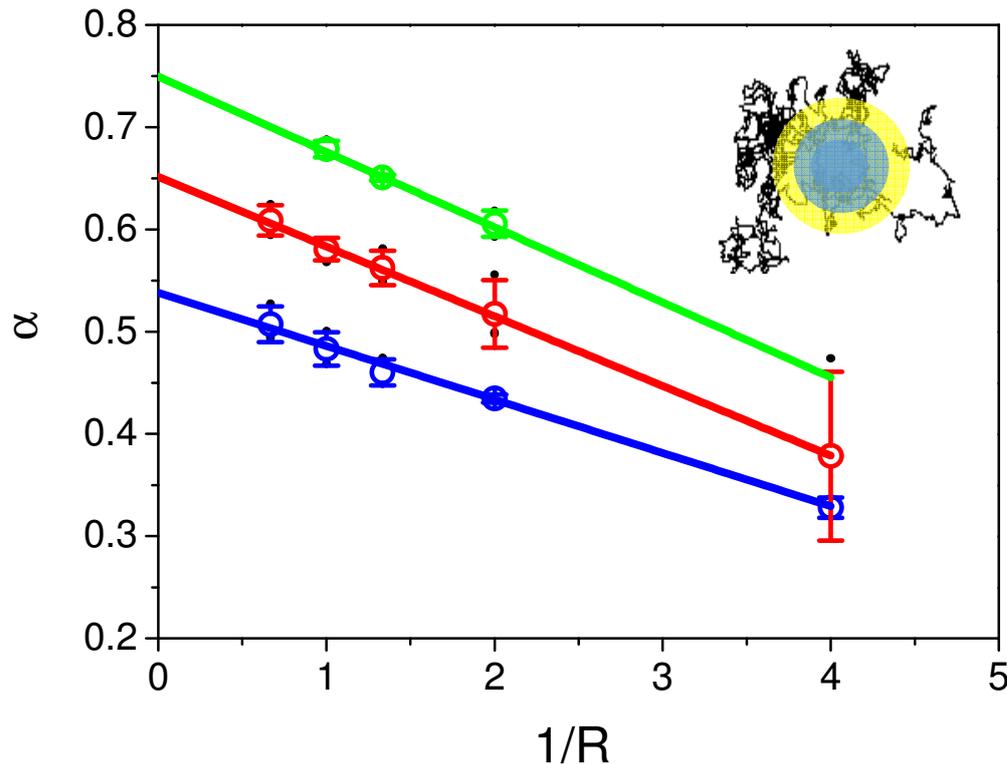


Figure 4. Values of α obtained from the fit of CTRW simulated recoveries as a function of $1/R$. Values of α introduced in the fit were respectively 0.6 (in blue), 0.7 (in red), 0.8 (in green). Dots represent the mean \pm s.d. values of α obtained with the fit using eq. 8 of simulated recoveries for a set of D values (0.01, 0.05, 0.1, 0.5, 1, 2, 3, black dots in the graph). Extrapolation at $1/R = 0$ of the linear fit of the different α obtained from the fits of recoveries at different radius gives α values close to the one used for the simulations.

164 Similar results were obtained when plotting D fitted as a function of $1/R$. Therefore, in order to
 165 test the hypothesis that α and D could be correctly determined by performing linear regressions of
 166 fitted values of both parameters as a function of $1/R$, a set of numerical simulation were performed
 167 using different D , α and R . Values of α and D at $1/R = 0$ intercepts are resumed in Appendix fig. A2.
 168 Fig.A2a and A2b show that, except for few values, when using this approach, α and D can be estimated
 169 with an error of less than 5 % of their real values. With regards to the discrepancy of the experiments
 170 on cells, this uncertainty seems enough accurate for correct determination of both parameters in case
 171 of anomalous diffusion processes using variable radii.

172 2.6. CTRW anomalous subdiffusion does not explain PH-EFA6 motions at the plasma membrane of BHK cells.

173 Variable radii FRAP has already been proposed by Salomé et al. in order to characterize membrane
 174 domains in cells [5]. They found both by numerical simulation and by experimental approaches that
 175 fitting recovery curves using the rBm model lead to a linear regression of the mobile fraction (M) as a
 176 function of $1/R$ with a positive slope. In this study our results show that the same approach is valid
 177 with aDm model and that plotting of α as a function of $1/R$ led to a linear regression with a negative
 178 slope. Therefore several experiments ($n < 30$) has been performed on cells expressing the PH domain
 179 of EFA6 at variable radii of photodestruction. Experimental recoveries were fitted with both models
 180 (aDm and fBm).

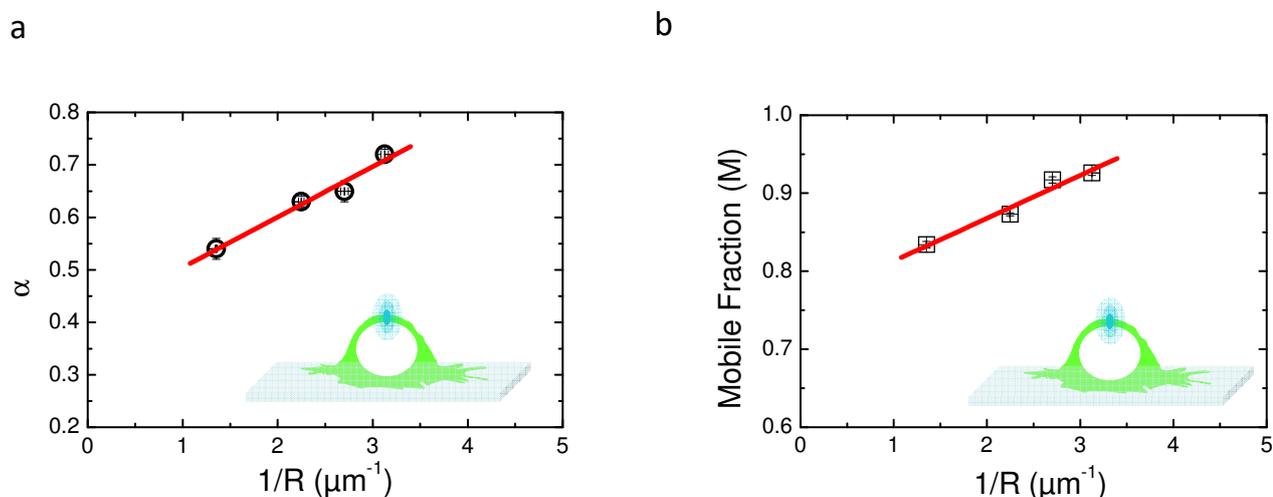


Figure 5. Comparison of the two models aDm and rBm using vrFRAP in the case of PH-EFA6 diffusion at the plasma membrane of BHK cells a: Plot of α values obtained by fitting the experimental recoveries with eq. 8 as a function of $1/R$. The plot exhibit a positive slope in opposite to the one observed in Fig. 4, suggesting an absence of a CTRW anomalous sub-diffusion in the motion of PH-EFA6. **b:** Plot of M value obtained by fitting the experimental recoveries with eq. 10 as a function of $1/R$. The plot exhibit a positive slope as observed in [25] and [5], suggesting a diffusion with trapping in spatial domains.

181 Figure 5 shows the behaviour of the characteristic parameters of each model (α for aDM and M for
 182 rBm) as a function of $1/R$. In Fig.5a is depicted the linear dependence of experimental α as a function
 183 of $1/R$. This results clearly show a positive slope for the regression, suggesting that the aDm model is
 184 not the correct model for the analysis of diffusion in the case of PH-EFA6 in this experimental time and
 185 length scale. On the opposite, Fig.5b, where M is plotted as a function of $1/R$ clearly shows a positive
 186 slope as already observed in Salomé's work [5]. This shows that, in the case of PH-EFA6 diffusion at
 187 the plasma membrane of BHK cells, the rBm is the more appropriate model to describe the restriction
 188 of diffusion observed in FRAP experiments.

Schram *et al.* empirically determined a relationship between the size of the trapping domains (L) and the variation of the mobile fraction (M) [25]:

$$M = M_p + \frac{0.63 \cdot L}{R}; L < R \quad (11)$$

189 Using eq.11, we could determine that 75% of PH-EFA6 molecules exhibit free diffusion while the
190 25 % left are confined in domains of approximately 90 nm radius.

191 3. Discussion

192 This work has been initiated to characterize the nature of the diffusion of molecules binding
193 the inner leaflet of the cell plasma membranes by means of FRAP experiments. In a first attempt,
194 we decided to compare experimental data obtained with the PH domain of EFA6 expressed in BHK
195 cells to FRAP curves generated from anomalous sub-diffusive particles numerically simulated. Then,
196 we analyzed the recoveries with three different diffusion models, namely the pure Brownian motion
197 (Bm), the restricted Brownian motion (rBm) and the CTRW anomalous subdiffusion (aDm). Four
198 parameters can be extracted from these diffusion models. The Brownian diffusion coefficient D and
199 the mobile fraction M ($M=1$ in the case of Bm) on one side, and the anomalous subdiffusion exponent
200 α and its related anomalous diffusion coefficient D_α on the other side. The aDm model has been
201 extensively studied by numerical simulations. Direct analysis of numerically simulated curves lead to
202 an underestimation of D_α and α . This was already observed by Feder *et al.* whom proposed, in order to
203 circumvent this underestimation, to add a mobile fraction (M) as a new parameter [15]. On a physical
204 point of view this is incorrect since the phenomenological parameter “mobile fraction” is indeed a part
205 of α as discussed by Nagle *et al.* [24]. This underestimation of D_α and α is mainly due to finite size effect
206 (space and time) that cannot be easily overcome neither in simulations nor in experiments. We directly
207 tested for anomalous sub-diffusion in the simulated and experimental curves by fitting the recovery
208 curves with normal and anomalous equations and look for systematic deviations of the fit, both in
209 linear plots to see the fit at large times and log-log plots to see the fit at short times. From this approach
210 we could see that the Bm can be immediately discarded. The difference between the aDm and the rBm
211 could only be observed at very short times (log-log plots) and very long times. Unfortunately, these
212 two extrema times are hardly easy to analyse in experiments. Indeed, at short times the curve may be
213 distorted by diffusion during the bleach pulse [26] and limits in the frequency of data collection. At
214 long times, motion of the membrane or photobleaching of the fluorescent probe might appear. This is
215 illustrated here in our experimental data. Fits of single spot fluorescence recoveries did not allow to
216 determine without uncertainties which of the aDm or the rBm model reflect the nature of PH-EFA6
217 diffusion at the plasma membrane of BHK cells. Although underestimated, the α value we found here,
218 when fitting with the aDm model, reflect a strong deviation from the Brownian motion and suggest
219 that PH-EFA6 explores a strongly compartmentalized landscape while travelling at the inner leaflet of
220 the plasma membrane. Nevertheless, this α value, as well as the M value in the case of the rBm model
221 are higher than the one found for IgE receptor transmembrane protein in RBL cells ($\alpha = 0.46 \pm 0.22$)
222 [15]. Using single particle tracking experiments, other transmembrane proteins such as MHC class I in
223 HeLa cells has also been shown to exhibit anomalous subdiffusion with an α value close to 0.5 [27]. On
224 the opposite, other transmembrane proteins exhibit high α values ($\alpha=0.8$) (Kv2.1 potassium channel
225 in HEK293T cells [28]) or pure Brownian motion (MHC class II in CHO cells [29] or aquaporin-1 in
226 MDCK cells [30]).

227 The inability of FRAP to cover several time decades as SPT or FCS techniques can be overcome
228 by probing the environment at different space scale using variable radii FRAP [5,6]. Here we have
229 simulated recoveries in the case of CTRW anomalous subdiffusion at different space scales and fit them
230 with the aDm model in order to extract the set of parameters (α , D). By monitoring the change of fitted
231 (α , D) parameters as a function of space ($1/R$) we observed that the fitted values of α decreased with
232 decreasing radius of observation. We showed that the correct values of α and D could be determined

233 at $1/R = 0$, i.e. when $R \rightarrow \infty$. This is one way to overcome the finite size effect of the measurements.
234 Then, we apply this approach to the experimental recoveries obtained at different radii. Surprisingly
235 we observed the opposite tendency to the one observed in our simulation, suggesting that the CTRW
236 anomalous subdiffusion is not the correct model to describe the motion of PH-EFA6 in BHK cells. On
237 the opposite, when monitoring the change of the mobile fraction obtained by fitting the experimental
238 recoveries with the rBm model, we observed the same tendency as the one described in [5] and [31], i.e.,
239 an increase of the mobile fraction with a decreasing radius. Using this approach, we could determine
240 that 25% of PH-EFA6 molecules are confined in domains of 90 nm radius.

241 CTRW is not the only source of anomalous sub-diffusion. The experimental increase of α with
242 decreasing radius can also be an apparent consequence of a cross over regime with two different
243 diffusion coefficient as it is described by the rBm model in this study. Using FCS experiments and
244 simulations at different radii of a two phases two component lipid mixtures at different temperature,
245 Favard *et al.* showed that changes in anomalous sub-diffusion exponent α could nicely predict the
246 phase transitions temperatures but failed in determining the average size of domains coexisting in
247 the two-phases [2]. On the opposite by monitoring the change in diffusion regimes, they could nicely
248 determine the mean size of the gel-phase domains. If we extend this approach to our *alpha* plot as
249 a function of the probes radius, we see that the transition from anomalous sub-diffusion ($\alpha < 1$) to
250 normal diffusion ($\alpha = 1$) occurs at radius of 160 nm, i.e., not far from the values obtained with the rBm
251 model.

252 The range of domain sizes observed here (90 to 160 nm radius), independently of the model used
253 to describe the dynamics of PH-EFA6, is likely to be due to subcortical actin cytoskeleton. Equivalent
254 sizes has been observed in NRK cells [32] using electron microscopy, and recently in several cell lines,
255 by monitoring membrane lipids dynamics using STED-FCS [33]. Interestingly, Krapf *et al.* described
256 that this meshwork has a fractal dimension and could therefore lead to anomalous sub-diffusion [34].
257 Therefore, further investigations and numerical simulations using a meshwork with fractal dimension
258 as the origin of the anomalous sub-diffusion are likely to be conducted in order to understand the
259 origin of our $\alpha = f(1/R)$ behaviour in our vrFRAP experiments.

260 In conclusion, by performing FRAP at variable radii, we found a way to discriminate CTRW
261 induces anomalous sub-diffusion from other restricted motion at the plasma membrane of living
262 cells. We also show that, while travelling at the inner leaflet of the plasma membrane, PH-EFA6 is not
263 stopped in various traps with different residence times but on the opposite is mainly freely diffusing
264 with on average 25% of the molecules confined in 90-160 nm radius open domains most probably due
265 to the actin cortical cytoskeleton.

266 4. Materials and Methods

267 4.1. Monte Carlo Simulation

268 In order to keep in our calculation the finite size effects, the simulation were made in a ring of
269 a radius of 30 arbitrary unit (a.u.) length explored by 106 particles for each run. Radii varying from
270 0.5 to 3 a.u. were photo-destructed. Reflective type of boundary conditions were used. This means
271 that when a particle gets out of the 30 a.u. radius it is re-introduced in the same direction at a small
272 distance of the boundary.

273 4.2. Cell culture and transfection

274 Baby hamster kidney cells (BHK) were grown on a coverslip in BHK-21 medium (Gibco-BRL),
275 containing 5% FCS, 10% Tryptose phosphate broth, 100U/ml penicillin, 100 μ g/ml streptomycin and
276 2mM L-Glutamine. Cells were transfected using Fugene 36 hours before the FRAP experiments with a
277 pC1EGFPHEFA6 plasmid. Fugene containing medium was replaced 12 h before the experiments by
278 fresh medium. pC1EGFPHEFA6 contains the sequence for both PH-EFA6 domain and EGFP as a

279 fluorescent label, linked to the N-terminus of the PH-EFA6 domain in order to avoid any perturbation
280 to the membrane linkage.

281 4.3. FRAP experiments

282 FRAP measurements were made with a commercially available confocal microscope, Leica
283 TCS-SP1 (Leica Microsystems, Germany). Prebleached images were firstly acquired to ensure for the
284 lack of photo-destruction during the observation. A brief laser pulse (200 ms) was then delivered to the
285 cell on a given and fixed position. Images were thereafter recorded at given intervals (440 ms) using a
286 spectral window for fluorescence emission between 500 and 600 nm. The intensity ratio between the
287 extinction laser beam and the monitoring laser beam was fixed to 10^6 . Each fluorescence recovery was
288 recorded for 100 s at 25°C, containing 150 experimental values (Recovery curve was sampled every
289 0.44 s in the beginning and 1 s in the end to avoid photobleaching during the monitoring). Focusing
290 the laser by the microscope objectives produced a Gaussian intensity distribution of the beam in the
291 object plane. This distribution was monitored using NBD-PC labelled DPPC multilamellar preparation
292 at 25°C ($T < T_m$). Since no diffusion occurs at this temperature, the image obtained immediately after
293 the end of the bleaching pulse shows a "hole" in the fluorescent preparation that allow measurement of
294 the laser waist and determination of the intensity profile in the x,y plane. These measurements were
295 confirmed by the use of fluorescent beads with a maximum wavelength of emission at 500 nm [35].
296 The following values were obtained for the waist as a function of the objective used :

Table 2. Size of the different radius measured with the objectives used in this work.

Objective	R (μm)	ΔR (μm)	laser waist (nm)
16x (NA=1.0)	0.74	0.04	370
40x (NA=1.3)	0.44	0.03	222
63x (NA=1.4)	0.37	0.01	185
100x(NA=1.4)	0.32	0.01	160

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303 Abbreviations

304 The following abbreviations are used in this manuscript:

305 FRAP	Fluorescence Recovery After Photobleaching
vrFRAP	variable radii Fluorescence Recovery After Photobleaching
FCS	Fluorescence Corelation Spectroscopy
SPT	Single Particle Tracking
306 CTRW	Continuous-Time Random Walk
aDm	anomalous subdiffusion motion
Bm	Brownian motion
rBm	restricted Brownian motion
PH-EFA6	Pleckstrin Homology domain of Exchange Factor for ARF-6

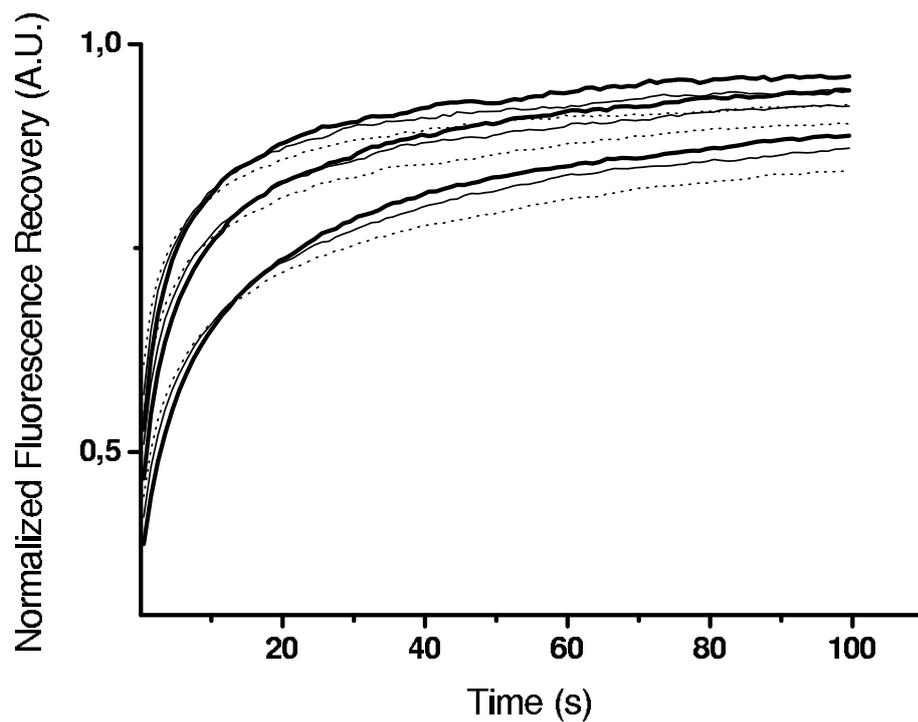


Figure A1. Monte-Carlo simulation of normalized fluorescence recoveries in the case of CTRW anomalous subdiffusion. Different values of D and α have been tested in the simulations. Here, values of $D = 0.5; 1; 1.5$ are represented from bottom to top with different α in each case: 0.6 (dots); 0.7 (thin line); 0.8 (thick line). The Monte Carlo has been constructed with 10^7 individual trajectories

307 Appendix A. Examples of Numerical Generated Fluorescence Recovery Curves

308 Figure A1 depicts some fluorescence recovery curves obtained with our numerical simulations
309 for different D and α inserted in the simulation.

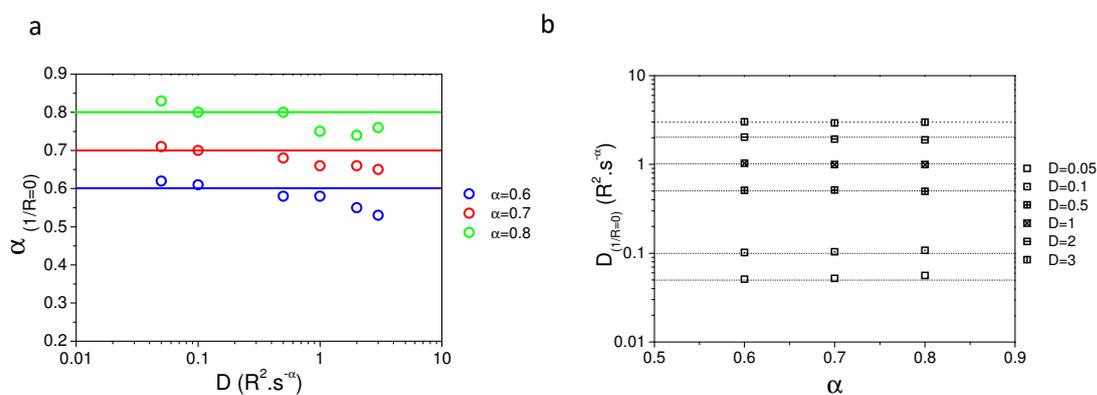


Figure A2. Values of α (fig. A2 a) and D (fig. A2 b) at $1/R = 0$ intercepts.

310 Appendix B. Comparison of input and fitted D and α at variable radii

311 In figure A2 are plotted values of α (fig. A2 a) and D (fig. A2 b) at $1/R = 0$ intercepts. Fig. A2a
 312 and A2b show that extrapolated $\alpha_{(1/R=0)}$ and extrapolated $D_{(1/R=0)}$ are found to be with an error of
 313 less than 5% of their simulation inserted values.

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