

Migration of Ca II H bright points in the internetwork

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ABSTRACT

Context. The migration of magnetic bright point-like features (MBP) in the lower solar atmosphere reflects the dispersal of magnetic flux as well as the horizontal flows of the atmospheric layer they are embedded in.

Aims. We analyse trajectories of the proper motion of intrinsically magnetic, isolated internetwork Ca II H MBPs (mean lifetime 461 ± 9 s) to obtain their diffusivity behaviour.

Methods. We use seeing-free high spatial and temporal resolution image sequences of quiet-Sun, disc-centre observations obtained in the Ca II H 3968 Å passband of the Sunrise Filter Imager (SuFI) onboard the Sunrise balloon-borne solar observatory. Small MBPs in the internetwork are automatically tracked. The trajectory of each MBP is then calculated and described by a diffusion index (γ) and a diffusion coefficient (D). We also explore the distribution of the diffusion indices with the help of a Monte Carlo simulation.

Results. We find $\gamma = 1.69 \pm 0.08$ and $D = 257 \pm 32 \text{ km}^2 \text{ s}^{-1}$ averaged over all MBPs. Trajectories of most MBPs are classified as super-diffusive, i.e. $\gamma > 1$, with the determined γ being the largest obtained so far to our knowledge. A direct correlation between D and timescale (τ) determined from trajectories of all MBPs is also obtained. We discuss a simple scenario to explain the diffusivity of the observed, relatively short-lived MBPs while they migrate within a small area in a supergranule (i.e. an internetwork area). We show that the scatter in the γ values obtained for individual MBPs is due to their limited lifetimes.

Conclusions. The super-diffusive MBPs can be described as random walkers (due to granular evolution and intergranular turbulence) superposed on a large systematic (background) velocity, caused by granular, mesogranular, and supergranular flows.

Key words. Sun: chromosphere – Sun: photosphere – methods: observational – diffusion – turbulence

1. Introduction

The study of small-scale, magnetic bright point-like features or magnetic bright points (MBPs) in the lower solar atmosphere has gained in interest over the last two decades because they trace kG magnetic features (Riethmüller 2013), many of which connect the photosphere with higher layers of the atmosphere (e.g. Stenflo 1989). Magnetic bright points are among the smallest potentially spatially resolved structures seen in the photosphere (e.g. Berger & Title 2001; Möstl et al. 2006; Sánchez Almeida et al. 2010; Riethmüller et al. 2010) and chromosphere (e.g. Rutten & Uitenbroek 1991; Steffens et al. 1996; Leenaarts et al. 2006; Jafarzadeh et al. 2013). Their motion is important for the braiding of the magnetic field in the corona; this braiding plays an important role in coronal heating according to Parker (1972, 1983, 1988); cf. Gudiksen & Nordlund (2002, 2005a,b); Peter et al. (2004).

Dispersion of the MBPs, i.e. their non-oscillatory motion on the solar surface, is thought to be due to photospheric flows, for example expansion and evolution of granules and supergranules, differential rotation, and meridional flows (Hagenaar et al. 1999). These motions are commonly described in terms of a diffusion process (e.g. Leighton 1964; Lawrence & Schrijver 1993; Dybiec & Gudowska-Nowak 2009; Ribeiro et al. 2011) whose efficiency is expressed by a diffusion coefficient (D) representing the rate of increase in the area that the MBP diffuses

across per unit time. This process can be characterised by the relation $(x - x_0)^2 \propto t^\gamma$, where $(x - x_0)^2$ represents the squared displacement (sd) of the tracked MBP between its location x at any given time t and its initial position x_0 ; γ is normally named the diffusion index. Motions with $\gamma < 1$, $\gamma = 1$ and $\gamma > 1$ are called sub-diffusive, normal-diffusive (random walk) and super-diffusive, respectively. In these regions sd grows more slowly than linear, linearly, or faster than linear with time, respectively (Dybiec & Gudowska-Nowak 2009).

Normal-diffusion was historically the first known class of diffusive processes. It characterises a trajectory which consists of successive random steps and is described by the simplest form of diffusion theory (e.g. Fick 1855; Einstein 1905; Lemons & Gythiel 1997). Brownian motion is an example of a normal-diffusive process. Leighton (1964) discussed the random walk interpretation of magnetic concentrations in the solar photosphere. He estimated the diffusion coefficient for granules and supergranules to be roughly $10^4 \text{ km}^2 \text{ s}^{-1}$. He found that this rate is comparable with the dispersal rate of the magnetic regions in the photosphere; and hence concluded that convective flows are responsible for the random walk of the magnetic concentrations (cf. Jokipii & Parker 1968; Lawrence & Schrijver 1993; Muller et al. 1994).

Sub-diffusive motion of magnetic elements in an active region in the photosphere was first reported by Lawrence & Schrijver (1993). Later, Cadavid et al. (1999) found that although the motion of network G-band MBPs in the photosphere is random if their lifetimes are larger than 25 min, MBPs with

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lifetimes less than 20 min migrate sub-diffusively. The sub-diffusivity was explained by the trapping of MBPs at stagnation points (i.e. points with zero horizontal velocity; sinks of flow field) in the inter-cellular pattern (Simon et al. 1995). In agreement with Cadavid et al. (1999), Hagenaar et al. (1999) stated that the diffusion index obtained from tracking magnetic elements may depend on their lifetimes.

There are only a few observational reports of super-diffusion in the lower solar atmosphere. Berger et al. (1998) found indications of slight super-diffusivity among otherwise normal-diffusive G-band MBPs in network regions. Later, in two high spatial and temporal resolution image sequences, Lawrence et al. (2001) found a significant number of super-diffusive MBPs between normal- ($\gamma = 1$) and Richardson diffusion (i.e. $\gamma = 3$). Recently, Abramenko et al. (2011) observed super-diffusivity of photospheric TiO MBPs in high spatial resolution time series. They reported the presence of super-diffusivity (as the only observed diffusion regime) for both quiet-Sun and active regions. This super-diffusivity was later confirmed by Chitta et al. (2012) by tracking MBPs observed in wideband $H\alpha$.

Diffusivity of MBPs is thought to be related to the turbulent convection on and below the solar surface (Nordlund 1985). In addition, it has been shown that stronger magnetic fields and larger magnetic elements result in smaller diffusion coefficients (Schrijver 1989; Schrijver et al. 1996).

In summary, the mobility of small MBPs has been described as dependent on temporal and spatial scales and on the strength of magnetic field.

We present measurements of the motion of small MBPs seen in the high-resolution Ca II H images of the SuFI instrument aboard the SUNRISE balloon-borne solar observatory. The MBPs under study are located in the quiet-Sun internetwork areas sampled at a height corresponding roughly to the temperature minimum. In particular, we track the motion of these small and intrinsically magnetic features (Jafarzadeh et al. 2013, hereafter Paper I) in time series of filtergrams and calculate their trajectories. We compute the sd for each trajectory and measure the corresponding power exponent (i.e. the diffusion index) as well as the diffusion coefficient.

In Sect. 2, we outline the dataset used to produce the time series. Section 3 represents the analysis method of the trajectories of the MBPs as well as the diffusion study results. In Sect. 4 we explore the role played by the distribution of diffusion indices with the help of a simple model. The concluding remarks are discussed in Sect. 5.

2. Data

For this study we used part of the datasets described in Paper I. The datasets consist of six time series of intensity images obtained in the Ca II H passband (centred at 3968 \AA with $FWHM \approx 1.8 \text{ \AA}$) of the SUNRISE Filter Imager (SuFI; Gandorfer et al. 2011) on board the SUNRISE balloon-borne solar observatory during its first flight in June 2009 (Solanki et al. 2010; Barthol et al. 2011; Berkefeld et al. 2011).

Table 1 lists the employed image sequences. All images were phase diversity reconstructed employing averaged wavefront errors (see Hirzberger et al. 2010, 2011), i.e. they correspond to level 3 data. For all images the field of view is $\approx 15 \times 41 \text{ arcsec}^2$ (712×1972 pixels) in size, with an image scale of $0.021 \text{ arcsec/pixel}$. All data refer to quiet regions close to the solar disc centre.

Figure 1 illustrates two snapshots from two different time series. We have indicated four MBPs with yellow boxes whose

Table 1. Log of observations.

Date	Time interval (UT)	No. of frames	Cadence
2009 Jun. 9	00:36–00:59	117	12 s
2009 Jun. 9	01:32–02:00	136	12 s
2009 Jun. 11	15:22–15:44	312	4 s
2009 Jun. 11	20:09–20:19	78	8 s
2009 Jun. 11	21:00–21:09	83	7 s
2009 Jun. 13	01:46–01:59	255	3 s

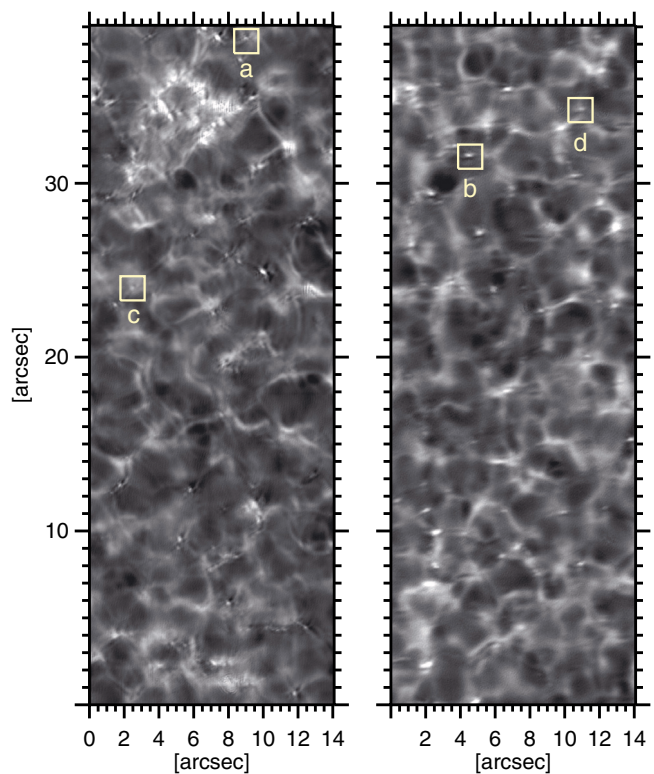


Fig. 1. Two examples of SuFI/SUNRISE Ca II H filtergrams. *Left:* observed on 9 June 2009, at 01:54:43 UT. *Right:* obtained on 11 June 2009, at 21:05:20 UT. The yellow boxes include selected MBPs whose trajectories are shown in Fig. 2.

trajectories will be studied here in detail. The left image in Fig. 1 contains a small network area in the upper part of the field of view that is excluded from our analysis which concentrates on internetwork MBPs.

3. Data analysis and results

We analyse trajectories of the Ca II H MBPs studied in Paper I. According to the definition introduced in Paper I, a Ca II H MBP is an isolated (no merging, no splitting, and no fine-structure resolved) bright, point-like feature in the height range sampled by the SuFI Ca II H 3968 \AA passband (see Jafarzadeh et al. 2013, for the relevant contribution function). The MBPs are included in this study if they (1) are located in an internetwork area around the disc centre; (2) are intrinsically magnetic (for those cases for which we have co-spatial & nearly co-temporal magnetograms); (3) have a lifetime longer than 80 s; (4) have a roughly circular shape with a diameter smaller than 0.3 arcsec ; and (5) are neither due to acoustic waves nor associated with “reversed granulation” (cf. Rutten et al. 2004).

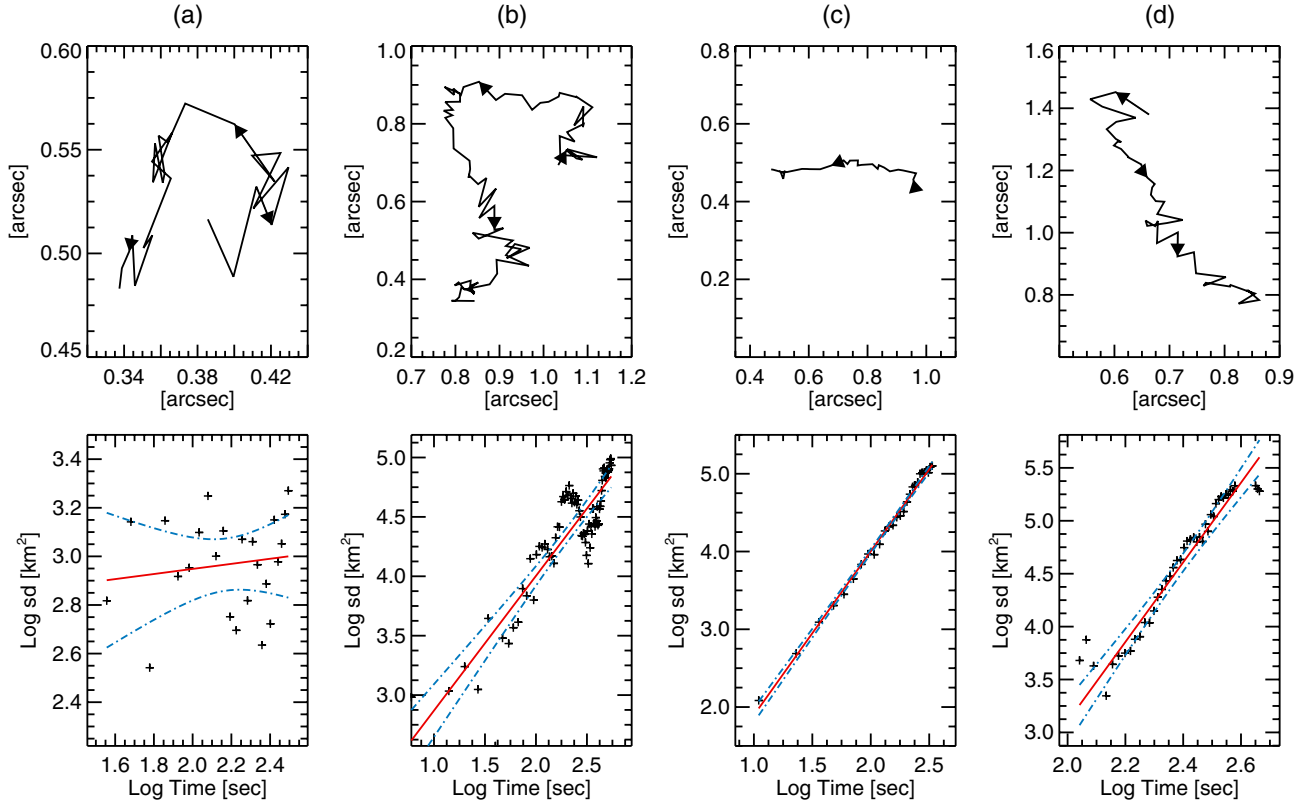


Fig. 2. Trajectories (*top*) and log-log plots of the squared displacement (sd) vs. time (*bottom*) of the four MBPs (a)–(d) identified in Fig. 1, marked at the top of the upper panels. Arrows on the trajectories indicate the direction of the MBPs’ motion. The red solid lines are the best linear fits to the sd data points (crosses) vs. time and the blue dot-dashed lines show the 95% confidence bands around the linear fit. The linear fits result in **a)** a MBP with $\gamma = 0.10 \pm 0.47$, **b)** a slope (or γ) of 1.13 ± 0.18 , **c)** a MBP with $\gamma = 2.10 \pm 0.11$, and **d)** a MBP with a high power exponent γ of 3.76 ± 0.57 . The uncertainties are computed from the 95% confidence bands.

The identified (selected) MBPs are tracked using the algorithm described in Paper I. Furthermore, the trajectories of the MBPs are computed and the modes of motion for each trajectory are investigated.

3.1. Trajectories

The trajectory of MBPs can be reconstructed by linking the string of x and y positions marking the MBPs’ locations in each frame. These MBPs have a mean lifetime of 673 ± 9 s, when measured only for those MBPs whose birth and death times are observable within the course of the time series and inside the images field of view (Jafarzadeh et al. 2013). The observed lifetime was found to be 461 ± 9 s on average when all MBPs are considered. The shortest and longest lifetimes are 167 ± 8 s and 1321 ± 13 s, respectively.

Four examples of trajectories of SUNRISE Ca II H MBPs are illustrated in Fig. 2 (upper panels).

Linking the MBP positions into a trajectory is only possible if the MBPs’ displacement in two consecutive frames is sufficiently smaller than the typical distance between the detected MBPs. This condition ensures that interacting MBPs or those that show apparent merging or splitting are excluded. This criterion may cause us to miss the fastest moving MBPs. Since a particular MBP may disappear and reappear again in a short time interval, such as the so-called persistent flashers (Brandt et al. 1992; Jafarzadeh et al. 2013). Therefore, these possible absent times were considered in the tracking algorithm, while a final visual inspection guaranteed the exclusion of interacting MBPs

in these special cases. For details on the locating and tracking procedures of the MBPs, we refer the reader to Jafarzadeh et al. (2013).

3.2. Diffusion processes

The trajectories of MBPs, $\mathbf{r}(t)$, for a diffusive process can be parametrised by their self-diffusion coefficients D based on the Einstein-Smoluchowski equation (Crocker & Grier 1996),

$$\langle |\mathbf{r}(t_0 + t) - \mathbf{r}(t_0)|^2 \rangle = 2dD\tau, \quad (1)$$

where τ is the elapsed time and d indicates the trajectory’s dimension. The term $\mathbf{r}(t_0 + t) - \mathbf{r}(t_0)$ gives the displacement of the MBP at a given time $t_0 + t$, from its first observed location at time t_0 . In its original definition, the left-hand side of the above equation is the mean of the squared displacement, whose value is computed by averaging all squared displacements (sd) in the system, i.e. all squared displacements over all MBPs. However, this averaging can result in the mixing of different diffusive processes and is not ideal for studying different motion types that the MBPs may have (Dybiec & Gudowska-Nowak 2009).

In its general form, diffusion is characterised by scaling of the variance of positions or alternatively the sd with time (Ribeiro et al. 2011),

$$sd(\tau) = C\tau^\gamma, \quad (2)$$

where C is the constant of proportionality. The power exponent γ (i.e. diffusion index) is the scaling factor of the sd and, by definition, can be used to classify distinct diffusion regimes as,

$$\begin{cases} \gamma < 1, & \text{sub-diffusive process} \\ \gamma = 1, & \text{normal-diffusive process (Random walk)} \\ \gamma > 1, & \text{super-diffusive process.} \end{cases} \quad (3)$$

Since diffusive processes in which the sd grows linearly with time ($\gamma = 1$) are called normal, deviations from linearity result in anomalous diffusion. Trajectories for which the sd grows more slowly or more quickly than linearly with time are said to lie in the sub- or super-diffusion regimes, respectively. Diffusive processes with $\gamma = 2$ and $\gamma = 3$ are often referred to as ballistic and Richardson diffusions, respectively.

In order to compute the γ , we calculate the sd of each MBP in each image (i.e. at every time-step) from its observed position in the first image it is present in. The diffusion index of the MBP can be conveniently captured by the slope of $sd(\tau)$ on a log-log scale. The standard deviation of the slope is also computed within 95% confidence intervals. Therefore, the true value of the slope and hence of γ lies within this confidence interval with a 95% probability.

Once γ is known, the diffusion coefficient D is calculated from the constant of proportionality C in Eq. (2) for each MBP separately. Following Abramenko et al. (2011), the diffusion coefficient, D , representing the rate of area in unit time that a MBP moves across, is estimated as the coefficient of turbulent diffusion described by Monin & Iaglom (1975) as a function of timescale,

$$D(\tau) = \frac{1}{2d} \frac{d}{d\tau} (sd(\tau)). \quad (4)$$

From Eq. (2), it follows that

$$D(\tau) = \frac{C\gamma}{2d} \tau^{\gamma-1}, \quad (5)$$

where $d = 2$ for our $2D$ trajectories.

In practice, C can be computed from the constant term of the linear equation obtained from the least-squares fit to the log-log plot of the $sd(\tau)$ for each MBP (i.e. the lower panels of Fig. 2; $C = 10^{\text{y-intercept}}$), γ is the slope of the fit and τ represents the MBP's lifetime.

Figure 2 displays the trajectory and $sd(\tau)$ plots of four selected MBPs indicated as (a)–(d) in Fig. 1, marked at the top of the upper panels in Fig. 2. The MBP (a) in Fig. 2 represents one of the few MBPs (lifetime 312 ± 13 s) with a small $\gamma = 0.10 \pm 0.47$. Although this is not the best example in our data, we show its trajectory (top panel) and its log-log plot (bottom panel) here, because it was the only MBP with $\gamma < 1$ in the selected frames shown in Fig. 1. The 95% confidence intervals are plotted as the confidence bands (blue dot-dashed lines) around the linear fits (red solid line) to the data points. As can be seen in the upper-left panel, the MBP moves over a small distance and in fact, it stays almost at the same position for some time before making a short jump to the next location. It moves by less than 0.1 arcsec from its initial position, i.e. by less than its own width (size) of 0.16 arcsec. The MBP (b) in Figs. 1 and 2, whose lifetime is 548 ± 8 s, provides an example where the slope of the linear fit (the red solid line in the bottom panel), $\gamma = 1.13 \pm 0.18$, is consistent with normal-diffusion. This particular MBP displays a peculiar behaviour. The MBP first tends to move quickly from its initial position. Later it changes its general direction and comes closer to its initial coordinates again. Had it disappeared

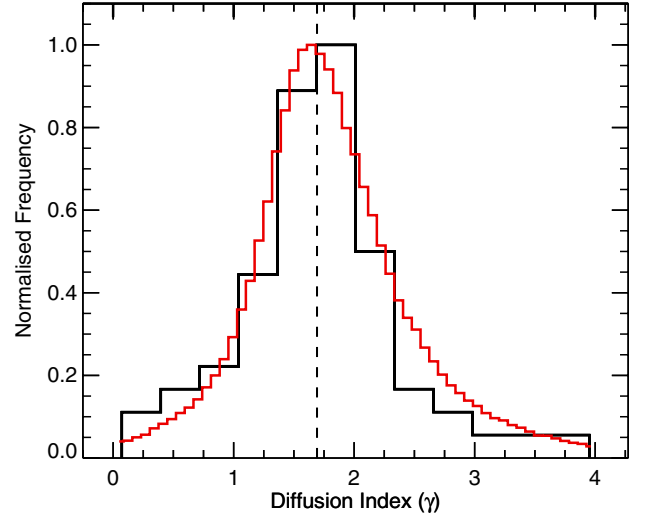


Fig. 3. Distribution of the diffusion index γ obtained from the SUNRISE/SuFI data (black histogram), normalised to unity maximum value. The vertical dashed line indicates the mean value of γ . The red histogram shows the result of a Monte Carlo simulation for the observed MBPs (see Sect. 4 for details).

after 300 s, for example, a larger γ would have been obtained. This hints that realisation noise, caused by the relatively short lifetimes of the MBPs, may result in a different γ . We will investigate this effect using a simple model in Sect. 4. The $sd(\tau)$ plots of the MBPs (c) and (d) in Figs. 1 and 2 (lifetimes 336 ± 13 s and 461 ± 8 s) result in $\gamma = 2.10 \pm 0.11$ and $\gamma = 3.76 \pm 0.57$, respectively. Although the two trajectories look rather similar in the upper panels of Fig. 2, they are different in the sense that the trajectory of MBP (d) displays a larger random component. The larger γ for example (d) comes from the fact that it gets accelerated, i.e. on average it moves faster with time.

3.2.1. Statistics

Figure 3 shows the statistical distribution of the diffusion indices γ of all 103 MBPs in the analysed SUNRISE/SuFI data (black histogram). The histogram is peaked close to $\gamma = 2$ and has a mean value of 1.69 (indicated by the vertical dashed line), with the distribution's standard deviation equal to 0.80. Eighty-eight percent of these MBPs have $\gamma > 1$ and only 12% exhibit $\gamma \leq 1$.

The red histogram illustrates the statistics of diffusion indices resulting from a Monte Carlo simulation. We describe and discuss this simulation in Sect. 4.2.

3.2.2. Diffusion coefficients

Table 2 summarises the computed diffusion coefficients, D , for the MBPs a–d (in Figs. 1 and 2) along with their lifetimes and diffusion indices, γ ; D increases with increasing γ for these four examples.

The computed D obtained from all SUNRISE/SuFI Ca II H MBPs are plotted as a function of γ in a log-linear plot in Fig. 4 (black crosses). A power-law fit to the data (i.e. $\gamma = 28.9 D^{2.8}$) is overplotted as a green dashed line.

The red line is obtained from the same Monte Carlo simulation as the red histogram in Fig. 3. In Fig. 4 the simulated line has been shifted upward to the observed trend (discussed in Sect. 4.2). The blue vertical lines indicate the error bars of

Table 2. Diffusion parameters of four example MBPs.

MBP	Lifetime [s]	γ^*	D^{**} [km ² s ⁻¹]
a	312 ± 13	0.10 ± 0.47	0.08 ± 0.01
b	548 ± 8	1.13 ± 0.18	35.8 ± 1.2
c	336 ± 13	2.10 ± 0.11	203.6 ± 14.7
d	461 ± 8	3.76 ± 0.57	811.7 ± 35.6

Notes. (*) Diffusion index. (**) Diffusion coefficient.

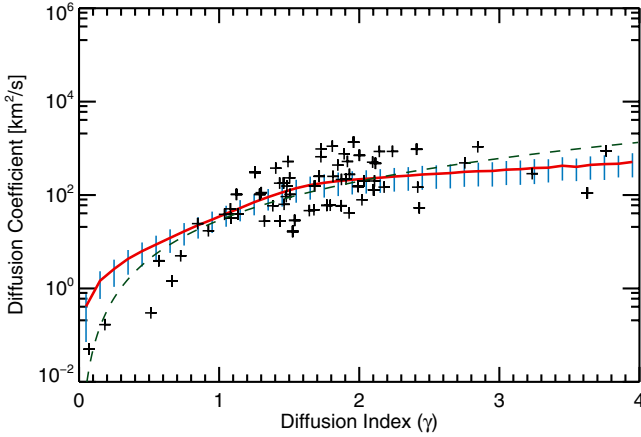


Fig. 4. Log-linear plot of diffusion coefficient, D , versus diffusion index, γ . Black crosses indicate the diffusion parameters calculated from trajectories of the SUNRISE Ca II H MBPs. The green, dashed line shows a power-law fit to the data points (see main text). The red solid curve is obtained from a Monte Carlo simulation. Error bars to the simulated values are depicted by blue vertical lines (see Sect. 4.2 for details).

the simulated trend. A mean value of $D = 257 \pm 32 \text{ km}^2 \text{ s}^{-1}$ was obtained from the dispersal of the SUNRISE (internetwork) Ca II H MBPs.

Plotted in Fig. 5 is $D(\tau)$ on a log-log scale. The black solid line shows the best linear fit to the data points obtained from the SUNRISE Ca II H MBPs. We observe a direct correlation between the D and timescale (τ) in agreement with the results on superdiffusive internetwork MBPs in the quiet Sun found by Lawrence et al. (2001) (red solid line) and Abramenko et al. (2011) (green dashed line). The slopes of the linear trends on the log-log plot, determined from our results and those of Abramenko et al. (2011) and Lawrence et al. (2001), represent power-law exponents of 0.63, 0.53, and 0.28, respectively. Our larger power-law exponent means smaller D in short timescales compared to the other two studies. This may be due to overestimation of short-term motions as a result of seeing in the ground-based data, while on longer timescales seeing effects average out. We also found a direct relationship between D and displacement (\sqrt{sd}) values (not shown), similar to that of Abramenko et al. (2011).

4. Modelling

The histogram of the diffusion index (i.e. Fig. 3) is peaked close to $\gamma = 2$, with wings reaching down to $\gamma < 1$ and up to $\gamma > 3$. Furthermore, more than one third (39%) of the γ computed for SUNRISE/SuFI MBPs are around $\gamma = 2$ (within their uncertainties). The fact that our longest lived MBP has a lifetime of 22 min (with each MBP being observed for ≈ 8 min on average) leads us to speculate that the broad wings of the distribution are due to the relatively small number of individual frames (steps) at which

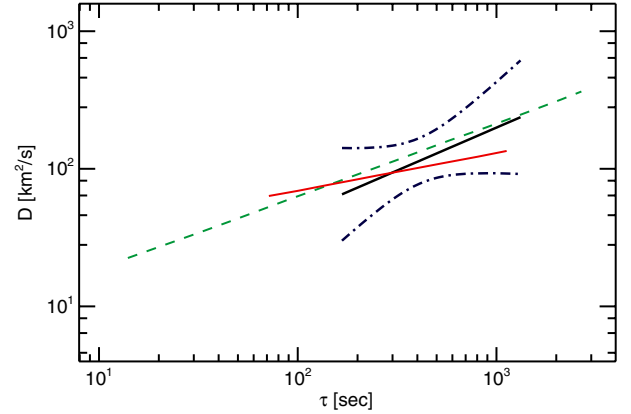


Fig. 5. Log-log plot of diffusion coefficients (D) as a function of timescale (τ). Black solid line: A linear fit to the data points determined from trajectories of the SUNRISE Ca II H MBPs. The 95% confidence bands to the fit are overplotted as dot-dashed lines. For comparison, similar plots (for quiet-Sun regions) from Lawrence et al. (2001) (red solid line) and Abramenko et al. (2011) (green dashed line) are overlaid.

we have observed each MBP, i.e. the width of the distribution is due to realisation noise caused by relatively short lifetimes of the MBPs. Clearly, the sd can be fit more precisely when a longer MBP motion is followed. This can be clarified by making a simple model of the horizontal velocity of a MBP. We note that we are dealing with internetwork MBPs, which can lie anywhere in a supergranule (SG).

4.1. Migration scenario

The MBPs are located in a complex turbulent medium. Correspondingly, the observed horizontal velocity of a MBP can be broken up into several components. In general, a variety of motions advect the MBPs, acting on different time and spatial scales. These motions range from turbulence in intergranular lanes (acting on the order of a few seconds), via the expansion and motion of granules (on the order of a few minutes), meso- and supergranular flows (on the order of one hour and about one day, respectively) to differential rotation and meridional flow. Given the relatively small field of view of all analysed time-series and their locations close to disc centre, the influence of the meridional circulation and differential rotation can be neglected.

The relatively short lifetimes of the MBPs under study implies that they move only over a fraction of a meso-/supergranule in the course of their average lifetime of ≈ 8 min. This in turn implies that the short-lived small-scale motions (i.e. intergranular turbulence and granular evolution) are primarily responsible for the random walk components of the MBPs' motion. Conversely, we expect that the meso- and supergranular flows as well as motions imparted by constantly expanding, contracting, and splitting granules on passively advected MBPs can be considered to be systematic.

We note that granules can contribute in a way to both random (v_r) and systematic (v_s) components of the MBPs' horizontal velocity, since the speeds imparted on MBPs due to granular evolution (slowly) change in the course of a MBP's lifetime.

In addition, it is worth mentioning that a part of the motion of the Ca II H MBPs can be due to kink waves excited at these magnetic elements. However, we do not search for any periodicity in the motion of the MBPs so that the effects of any kink waves running along the field lines underlying the MBPs are assigned to the random motion.

4.2. Monte Carlo simulation

Assuming an explanation as proposed in Sect. 4.1, we explore the statistical distribution of the diffusion index γ (shown in Fig. 3) and its relationship with D (plotted in Fig. 4) obtained from the SUNRISE/SuFI data with the help of a Monte Carlo simulation.

We allow 30 000 randomly generated “MBPs” to move along the perpendicular x and y axes in $2D$ space and in time t . The x and y axes represent the directions of the random motions of a MBP, while the systematic direction is identified with the x axis. As in Sect. 4.1 the latter can be taken to represent mainly meso-granular, supergranular and granular flows. The random motions are thought to be largely associated with granular and turbulent intergranular flows (on short timescales). The random walk is modelled assuming a discrete time-step of Δt (coherence time), over which the velocity is assumed to be constant. The Δt and the ratio of the random to systematic velocities (v_r/v_s) are treated as initial free parameters of the simulation.

We performed the Monte Carlo simulations for MBPs with lifetimes between 160 and 1320 s; 30 000 realisations were analysed for each lifetime. For each run of the simulation (for a particular lifetime), the trajectories of the 30 000 simulated MBPs were determined in the same way as for the observed MBPs, described in Sect. 3.1, and are also analysed to compute their diffusion indices using a method similar to that applied to the SUNRISE/SuFI Ca II H MBPs as described in Sect. 3.2 (i.e. the slope of the log-log plot of the $sd(\tau)$ for each MBP is computed from the simulated MBP trajectory). Therefore, for any set of the two free parameters, we obtain a distribution of diffusion index γ for the 30 000 simulated MBPs with a given lifetime.

Integration of the individual simulated histograms corresponding to each lifetime results in a distribution of γ calculated for MBPs with different lifetimes, similar to the distribution obtained from the observations. To compare with the observed distribution, the histogram resulting from the simulation is re-binned to match the resolution of the one obtained from the observations. Then, the free parameters are tuned until the best match between these two histograms (i.e. one obtained from the observations and one calculated from the Monte Carlo simulation) occurs. The chi-square, χ^2 , is computed as a measure of the best match between the two histograms. Since we compare the distributions of the two binned datasets, the quantity χ^2 is measured as (Press et al. 2007)

$$\chi^2 = \sum_i \frac{(O_i - C_i)^2}{O_i + C_i}, \quad (6)$$

where O_i and C_i indicate the i th observed and computed (simulated) bin of the histograms.

In Fig. 6 the coherence time Δt is plotted as a function of the ratio of velocities v_r/v_s for all χ^2 values. The Δt and v_r/v_s combination giving the smallest χ^2 value was found to be 62 ± 3 s and 1.6 ± 0.03 . We note that the absolute values of the velocity are not constrained by the histogram of γ ; therefore for simplicity we used $v_r = 1 \text{ km s}^{-1}$.

The result of the simulated distribution of γ based on these parameters (integrated over all simulated distributions with different lifetimes) is plotted in Fig. 3 as a red histogram. A comparison with the black histogram, which represents the distribution of diffusion index obtained from the observations (i.e. from the SUNRISE/SuFI data), shows that the simulated distribution is stronger than that of the observations in the (far) wing at $\gamma > 2.5$, but it is weaker for $\gamma < 1$.

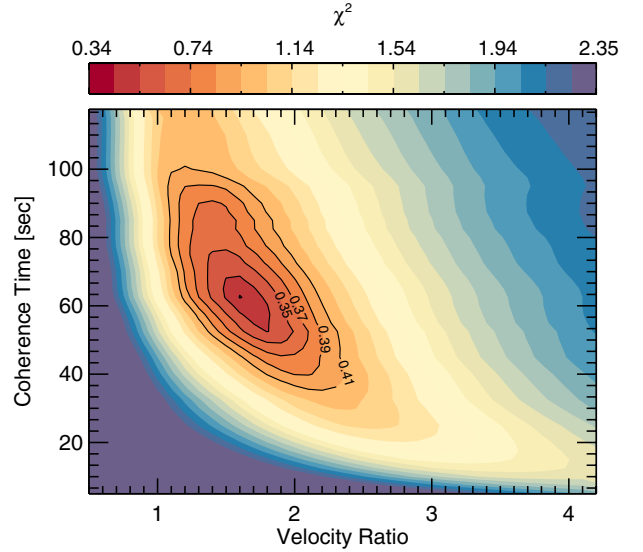


Fig. 6. Results of a chi-square (χ^2) test for finding the best match between the distributions of the diffusion index from observations and Monte Carlo simulations of the MBPs located in a supergranule’s bulk. The contours guide the eye to the best χ^2 value, giving the appropriate coherence time Δt for the random velocity and the ratio of random-to-systematic velocities v_r/v_s of the MBPs under study.

In the next step, we determine the absolute values of these random and systematic velocities using the diffusion coefficient D . Given the determined ratio of the velocity components as well as the coherence time, we compute D in the same way as for the observed data, described in Sect. 3.2.2. Then, we overplot the computed D versus γ for the simulated data on the same log-linear plot as for the observed data. For simplicity, we average the D values in each γ bin and overplot its standard deviation as an error bar for each point. The plot of $\log(D)$ versus γ in Fig. 4 shows similar trends for both simulated and observed data, but originally with an offset in the direction of $\log(D)$.

We have only one free parameter left in the simulation, namely v_s , with which to shift the $\log(D)$ versus γ curve. We note that an arbitrary value for v_r had been used so far. Therefore, we tuned v_r , while keeping v_r/v_s and Δt fixed, until we found the best agreement between the observed and computed plots of $\log(D)$ versus γ . The output of the simulation giving the best agreement with the observations is overplotted as a red line (along with the error bars; vertical blue lines) on the observed data (black crosses) in Fig. 4. The v_r giving the best fit was found to be $1.2 \pm 0.1 \text{ km s}^{-1}$ on average. Consequently, the systematic flow tends to move with a velocity of $0.75 \pm 0.06 \text{ km s}^{-1}$.

Finally, we take these best-fit values and do the simulation once again, but for a larger range of lifetimes, i.e. from 80 s, which represents the criterion for minimum lifetime of the MBPs under study, to three times longer than the maximum lifetime of the observed MBPs from the SUNRISE/SuFI data. A plot of γ versus lifetime resulting from this simulation is illustrated in Fig. 7. The background brightness follows the number density of MBPs with a particular γ . A vertical cut at a given lifetime indicates the histogram density of the γ distribution at the chosen lifetime. The solid lines mark percentiles, i.e. fixed values of the integrals over the histogram density in the vertical direction starting at the bottom. Each yellow cross represents an observed MBP from the SUNRISE/SuFI data. At the left side of this plot, where most of our observed data points are located, the lifetimes are shorter and correspond to much fewer time-steps than the right side. The

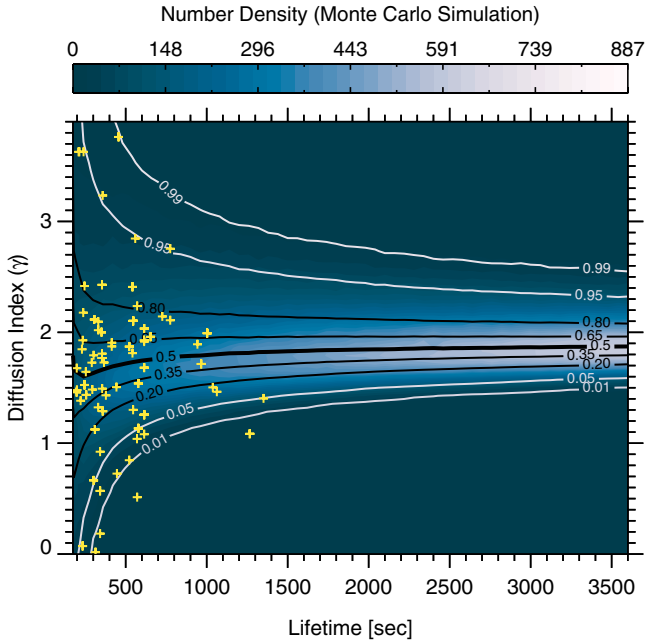


Fig. 7. Evolution of the diffusion index histogram with increasing lifetime of the MBPs. The background brightness is a measure of the number density of MBPs with a certain γ obtained from a Monte Carlo simulation. The solid lines indicate the percentiles of the distributions (see main text for details). The observed data points from the SUNRISE/SuFI data are overlaid on the simulated plot and are marked by yellow crosses.

left part of the plot displays a wider distribution of γ , with the histograms having extended tails, similar to the tails of the observed histogram in Fig. 3. Insufficient sampling on the left part of the diagram enhances the effect of the random velocity, so that an individual MBP can display a γ rather far removed from the expectation value.

There are only few data points (as seen in Fig. 7) that have γ values that are highly unlikely according to the simulations, outside the 5% and 95% percentile curves, for example. One source of these outlying MBPs may be the fact that we considered a uniform systematic flow dragging the MBPs with it, whereas in reality the flow speed (caused by the contraction, expansion or explosion of granules as well as the meso- and supergranular flows) is expected to vary. However, the simulated histogram includes almost all of the data points and so can explain most of the observations. Interestingly, the four data points whose locations lie outside the outermost (1%) contour of the simulation in Fig. 7 all have small γ . One possibility to explain the excess of low γ MBPs is that these are lying at the borders of the SGs, where sub-diffusive and random walk MBPs are expected and are due to inflows from opposite directions (from neighbouring SGs).

To summarise, the observations are consistent with the MBPs being random walkers superposed on a systematic flow. Therefore, while the motion of MBPs has a random component due to intergranular turbulence as well as the birth and death of neighbouring granules, the MBPs are transported by a systematic velocity on a larger spatial scale, e.g. due to the constant evolution of long-lived granules, or the contributions of meso- and supergranules to the large-scale velocity field felt by the magnetic element underlying the MBP.

We found a coherence time of 62 ± 3 s which is comparable with a value of 68 s computed from the given size (length scale) of our features (i.e. 150 km on average) as well as the mean horizontal velocity of the MBPs under study (i.e. 2.2 km s^{-1})

reported by Jafarzadeh et al. (2013). This implies that on average MBPs move a distance corresponding to their own diameter before being forced to move in another direction.

5. Discussion and conclusions

We analysed the trajectories of 103 isolated (i.e. displaying no merging or splitting) internetwork MBPs (≈ 0.2 arcsec in diameter on average; Jafarzadeh et al. 2013) in the quiet Sun observed in the Ca II H 3968 Å passband of SUNRISE/SuFI. We performed a diffusion analysis on the trajectories of MBPs to distinguish between MBPs with different types of motions. We did not, however, search for oscillations or wave-like motions.

In order to avoid mixing MBPs in different diffusion regimes and to get a better insight into the character of their proper motion, we performed the diffusion analysis on all individual trajectories separately (Dybiec & Gudowska-Nowak 2009).

We used the same MBPs as in Paper I where they were identified using stringent criteria. This gives us confidence that these MBPs are good tracers of small-scale magnetic elements in the upper layers of the photosphere and the lower chromosphere. These intrinsically magnetic features can be considered to represent the cross-sections of nearly vertical flux tubes (Jafarzadeh 2013) whose positions are influenced by different external forces and are restricted to intergranular lanes. Therefore, although they are not perfect tracers of the horizontal flows, the spatial and temporal scaling of the MBPs' dynamics can still indicate the presence of turbulence in their paths (Lawrence et al. 2001).

We should note that the relatively small field of view ($\approx 15 \times 41$ arcsec), short time series (less than 28 min), and the restrictive criteria applied to select these small MBPs have limited our sample to a relatively small number of 103 isolated MBPs (a number density of 0.03 (Mm)^{-2} ; Paper I). However, selecting only point-like features has allowed us to locate the MBPs more accurately.

5.1. Diffusion index

We found a mean diffusion index (or diffusion power-law exponent), γ , of 1.69 ± 0.08 averaged over all MBPs. The γ histogram is peaked close to this value and ranges from $\gamma \approx 0$ to $\gamma \approx 4$ (see Fig. 3). We suspect, however, that this wide range in the γ distribution does not reflect the presence of different diffusion regimes (i.e. sub-, normal- or super-diffusive; as described in Sect. 3.2), but that it is possibly due to realisation noise (around the mean value representing a single diffusion category) due to our relatively short-lived MBPs (mean lifetime ≈ 8 min). The short lifetime may in turn be due to the fact that we consider only small (< 0.3 arcsec diameter) internetwork MBPs.

We also explored the distribution of the diffusion index with the help of a Monte Carlo simulation (see Sect. 4.2). It demonstrated that the migration of internetwork MBPs is consistent with a random walk (due to intergranular turbulence and granular evolution) superposed on a systematic velocity (caused by granular as well as meso- and supergranular flows). We note that a part of the motion of the Ca II H MBPs can be due to kink waves excited in the underlying magnetic elements, whose effects were not separated in our study from that of random motion. The simulation clarified and confirmed that the deduced γ values of almost all MBPs are consistent with a single underlying γ of 1.69 ± 0.08 and that the large scatter of γ values of individual observed MBPs is indeed caused by the short lifetimes of these MBPs and the associated realisation noise. Hence, the

Table 3. Comparison of the mean values of diffusion index γ and diffusion coefficient D of small magnetic elements obtained in this study with some of those in the literature.

Reference	Origin of data	Telescope/Spacecraft	Spatial resolution	Feature ^a	Lifetime ^{b,c} [s]	γ^c	$D^{b,c}$ [km ² s ⁻¹]
This study	Stratospheric balloon	SUNRISE/SuFI	0'.14	IMBP	461	1.69	257
Chitta et al. (2012)	Ground	SST/CRISP	–	IMBP	180–240	1.59	(\approx 90)
Abramenko et al. (2011)	Ground	BBSO/NST	0'.11	IMBP ^d	(10–2000)	1.48	(19–320)
Manso Sainz et al. (2011)	Space	Hinode/SOT	0'.32 ^e	IMBP	<900	0.96	195
Utz et al. (2010)	Space	Hinode/SOT	0'.22 ^f	IMBP	150	(\approx 1)	350
Chae et al. (2008)	Space	Hinode/SOT ^g	–	ME ^d	–	–	0.87
Lawrence et al. (2001)	Ground	SVST	0'.23	NMBP	9–4260	1.13	–
Cadavid et al. (1999)	Ground	SVST	0'.23 ^h	NMBP	18–1320	0.76	–
Hagenaar et al. (1999)	Space	SOHO/MDI	2'.3 ⁱ	MFC	1500–3450 >3.0 × 10 ⁴	1.10 (\approx 1)	– 70–90 220–250
Berger et al. (1998)	Ground	SVST	\approx 0'.2	NMBP ^d	(100–3800)	1.34	60
Lawrence & Schrijver (1993)	Ground	BBSO	–	ME ^d	–	0.92	250

Notes. ^(a) IMBP: Internetwork Magnetic Bright Point; NMBP: Network MBP; MFC: Magnetic Flux Concentration; ME: Magnetic Element. ^(b) Mean values; otherwise a range is given whenever it was available. ^(c) Values in parentheses are not directly reported in the papers, but are computed from the given plots/tables. ^(d) See their paper for other types of regions. ^(e) From Martínez González & Bellot Rubio (2009). ^(f) From Table 1 of Abramenko et al. (2011). ^(g) From SOT magnetograms with the smallest pixel size of 116 km. ^(h) From Lawrence et al. (2001). Cadavid et al. (1999) reported a much lower resolution element of 0'.4. ⁽ⁱ⁾ For five-minute averaged and spatially smoothed high-resolution magnetograms.

dispersal of the studied Ca II H MBPs in the quiet Sun is mainly super-diffusive.

We note that excluding the network regions in the present study most likely has also excluded most MBPs located at stagnation points and hence, sub-diffusive MBPs may be under-represented in our sample, compared to the entire quiet Sun.

Our main result is in rough qualitative agreement with that of Lawrence et al. (2001), who found $\gamma = 1.13 \pm 0.01$ (on average) for MBPs, using high spatial and temporal resolution images in a network region of enhanced magnetic activity. However, the features we studied display a much stronger super-diffusive behaviour. Super-diffusivity has also been reported by Abramenko et al. (2011) and Chitta et al. (2012) in recent investigations. They deduced values of the diffusion index in a given type of region by averaging the displacements over all MBPs. Abramenko et al. (2011) obtained $\gamma = 1.48$ for quiet Sun, $\gamma = 1.53$ for plage, and $\gamma = 1.67$ for coronal hole areas, while Chitta et al. (2012) reported a diffusion index of 1.59 for their relatively short-lived MBPs (with a mean lifetime of about 3–4 min) observed in quiet-Sun regions. The mean γ for the quiet Sun obtained in both studies are roughly consistent with the average $\gamma = 1.69 \pm 0.08$ that we found in the present work, although their values are slightly smaller. This may have to do with the fact that earlier γ values were obtained from ground-based observations and could be affected by differential seeing-induced deformations (introducing artificial turbulences), while we determined $\langle\gamma\rangle$ using the SUNRISE/SuFI data unaffected by seeing.

Lepreti et al. (2012) used the same datasets as Abramenko et al. (2011) to determine γ (for timescales < 400 s) for pair dispersion of MBPs (i.e. from measuring the mean-square separation of pairs of MBPs). They found the same γ (\approx 1.48) for MBP pairs observed in all the three regions (i.e. quiet-Sun, plage, and coronal hole areas), from which they interpreted the diffusivity properties as the results of the local correlations in the turbulence's inertial range. In addition, Lepreti et al. (2012) concluded that the diffusivity of individual MBPs studied

by Abramenko et al. (2011) (that differs in the three regions) depends on the detailed structure of the flows.

Manso Sainz et al. (2011) found in contrast to other recent studies a mean value of 0.96 (nearly corresponding to normal-diffusion) averaged over a wide distribution of γ for magnetic internetwork elements. We suspect that such a small value is due to their criterion of tracking only the footpoints of small-scale magnetic loops. The motion of this freshly emerged field is only partly driven by flows at the solar surface, while to a significant extent it also reflects the dynamics and subsurface structure of the emerging field. Hence their results may not be directly comparable with ours.

Some published values of γ , of the diffusion coefficient (D), and of the mean lifetimes of the investigated features are compared with those obtained in this study in Table 3. Since the effect of atmospheric seeing may be important, it is indicated if the observations are space, ground, or balloon-based. The spatial resolutions of the observations are provided where known. The types of solar regions that were investigated can be deduced from the names of the features. With the exception of the investigations finding super-diffusive motions, discussed earlier, almost all other authors interpreted their results as indicative of normal- or sub-diffusive processes. Berger et al. (1998) did find a $\gamma = 1.34 \pm 0.06$. However, they interpreted it in terms of normal-diffusion with a slight indication for super-diffusivity, since the area coverage of their network MBPs as a function of time could be well explained by a Gaussian model. Cadavid et al. (1999) found normal-diffusion ($\gamma = 1.10 \pm 0.24$) for timescales longer than 25 min and sub-diffusive MBPs ($\gamma = 0.76 \pm 0.04$) on timescales shorter than 22 min, both based on tracking MBPs in a network area.

5.2. Diffusion coefficient

We also determined a mean diffusion coefficient (D ; the area that a MBP moves across per unit time) of 257 ± 32 km² s⁻¹

averaged over all studied MBPs. The diffusion coefficient values reported in the literature and summarised in Table 3 lie between $0.87 \text{ km}^2 \text{ s}^{-1}$ and $350 \text{ km}^2 \text{ s}^{-1}$. One source of such a large range could be that different values could refer to different features. Schrijver et al. (1996) noted the flux dependence of the diffusivity of flux concentrations, with the smaller concentrations moving faster compared to the larger ones. Tracking small features on relatively short timescales is another source of bias, since the potential effects of larger scales (e.g. supergranular flows) are detected more clearly when measuring for a sufficiently long duration (Schrijver et al. 1996). By tracking magnetic features in the large FOV of MDI magnetograms, Hagenaar et al. (1999) found that the diffusion coefficient, $D = 70\text{--}90 \text{ km}^2 \text{ s}^{-1}$, determined for intervals of time less than $1.0 \times 10^4 \text{ s}$ is smaller than that measured on timescales longer than $3.0 \times 10^4 \text{ s}$, which gives $220 \leq D \leq 250 \text{ km}^2 \text{ s}^{-1}$. They interpreted this difference (using a model) as the effect of supergranular flow that acts as a negligible drift on short timescales. Berger et al. (1998) determined $D = 60.4 \pm 10.9 \text{ km}^2 \text{ s}^{-1}$ for network MBPs by assuming, for simplicity, Gaussian (normal) diffusion (i.e. $D = sd/2d\tau$), whereas the γ they obtained, $\gamma = 1.34 \pm 0.06$, corresponds to the super-diffusive regime. They also found a $D = 285 \text{ km}^2 \text{ s}^{-1}$ in quiet-Sun internetwork regions by tracking markers (so-called corks) added on frames of image-sequences. Cadavid et al. (1999) discussed the mismatch between different reported diffusion coefficients in the literature due to the assumption of normal-diffusion for all investigated features (i.e. determining D by assuming $\gamma = 1$ in Eq. (5) for simplicity, regardless of actual diffusion indices) in most of the studies to that date. Roudier et al. (2009) estimated $D = 430 \text{ km}^2 \text{ s}^{-1}$ from Hinode/SOT observations using floating corks on a relatively large scale (larger than 2.5 Mm; since they used granules to determine the flow). Chae et al. (2008) reported the smallest value of $D = 0.87 \pm 0.08 \text{ km}^2 \text{ s}^{-1}$ from Hinode/SOT magnetograms. They estimated D by modelling the change of magnetic field in individual pixels between two frames (taken 10 min apart in time), based on solving the equation of magnetic induction. Their unusually small value, however, could likely be caused by the use of individual pixels instead of clearly separated magnetic features in their somewhat different method of diffusivity measurements. Later, Utz et al. (2010) found $D = 350 \pm 20 \text{ km}^2 \text{ s}^{-1}$ for internetwork MBPs observed by the same telescope (i.e. Hinode/SOT). They obtained this value in the framework of the normal-diffusion process. Manso Sainz et al. (2011) used the same instrument and found $D = 195 \text{ km}^2 \text{ s}^{-1}$ for footpoints of small-scale internetwork magnetic loops. We speculate that this relatively small value compared to that found in the present work is due to the fact that they considered only freshly emerged small loop footpoints, which may not follow surface flows to the same extent as our magnetic features.

Cameron et al. (2011) described the decay of the magnetic field by turbulent diffusion through 3D radiative MHD simulations. They characterised the decay in terms of diffusion coefficients and found D to lie in the range of $100\text{--}340 \text{ km}^2 \text{ s}^{-1}$. This range encompasses the mean value of $D = 257 \pm 32 \text{ km}^2 \text{ s}^{-1}$ we obtained from the dispersal of SUNRISE Ca II H MBPs.

5.3. A relationship between diffusion coefficient and timescale

Furthermore, we found a direct correlation between D and timescale (τ) computed from trajectories of all MBPs (see Fig. 5). We note that the dependence of $D(\tau)$ on γ is not surprising though, since it follows the relationship expressed in

Eq. (5). The $D(\tau)$ trend we obtained from our super-diffusive MBPs tends to be steeper than those measured by Lawrence et al. (2001) and Abramenko et al. (2011) for internetwork super-diffusive features. Abramenko et al. (2011) compared such a relationship with other types of regions (e.g. network areas, active regions and a coronal hole area) and other diffusion regimes reported in the literature. They showed that such a direct correlation between D and τ is only observed for the case of super-diffusion ($\gamma > 1$). Values of D independent of τ for $\gamma = 1$ (Schrijver et al. 1996; Berger et al. 1998; Hagenaar et al. 1999) as well as anti-correlations between D and τ (Berger et al. 1998; Cadavid et al. 1999) for $\gamma < 1$ are also observed.

5.4. Migration of MBPs over a supergranule

As stated earlier, our Monte Carlo simulation describes the motion of our MBPs as the superposition of a directed systematic velocity v_s and a random velocity v_r . A comparison of the model output with observations of the horizontal motion of the Ca II H MBPs allowed us to determine both the random and systematic components as well as the effective coherence time of the random walk flows. We determined the mean values of these velocities to be $v_r \approx 1.2 \pm 0.1 \text{ km s}^{-1}$ and $v_s \approx 0.75 \pm 0.06 \text{ km s}^{-1}$. A coherence time of $62 \pm 3 \text{ s}$ was found for the random velocity.

The random component of the velocity obtained here is consistent with that computed for the rms value of the horizontal transport of the magnetic concentrations in intergranular areas from 3D radiative MHD simulations (Cameron et al. 2011).

Horizontal flows in supergranules can be measured through the tracking of corks, for example. Spruit et al. (1990) tracked corks over a mesogranule and found a cork velocity of 1.0 km s^{-1} as they move towards the mesogranular boundaries (on timescales of about 10–30 min). They showed that the flow speed decreases when approaching the mesogranular boundaries, reaching a minimum of about 0.5 km s^{-1} . This range of systematic velocities caused by the combination of granular, meso- and supergranular flows is comparable with that found in the present work. The v_s obtained in our investigation is also in good agreement with that of Del Moro et al. (2007), who found a horizontal flow speed of $0.75 \pm 0.05 \text{ km s}^{-1}$ inside a supergranule via cork tracking. Our determined systematic velocity is, however, slightly larger than the mean horizontal flow velocity of $\approx 0.4 \text{ km s}^{-1}$ within supergranules, reported by Title et al. (1989), Wang et al. (1995), and Hathaway et al. (2002) and the $\approx 0.3 \text{ km s}^{-1}$ reported for mesogranular flows by Leitzinger et al. (2005).

Recently, Orozco Suárez et al. (2012) studied the horizontal velocity of both convective flows and internetwork magnetic elements (IMEs) over a SG during a 13 h uninterrupted observing campaign with Hinode/NFI. They found that the IMEs are almost at rest at the centre of the SG where they start accelerating radially outward. The IMEs also tend to decelerate while approaching the SG boundaries.

We speculate that a hypothetical long-lived internetwork MBP (that may appear or disappear at any location in the body of a SG) would gently accelerate over the SG's bulk. This is caused by the increasing velocity with radial distance of the supergranular flow profile due to mass conservation assuming a constant upflow over most of the SG's area. The supergranule's flow systematically advects all MBPs within its bulk towards its borders. In addition, granular and mesogranular flows (that act on shorter spatial scales) would impart the MBP with additional velocity components that change as these convection cells evolve. The trajectory of such a MBP is expected to follow a super-diffusive

regime, although with a lower γ than expected from the SG flow alone, because of the random motions imparted mainly by granular evolution (which for a very long-lived MBP contributes mainly to v_r) and intergranular turbulence. This MBP would start decelerating when approaching the SG boundary. Once the MBP finds itself in the network region, it is trapped in the sinks (stagnation points) due to inflows from opposite directions, i.e. from neighbouring SGs. The ever-evolving granular flows will keep acting on small spatial scales. The trajectory of the MBP at this stage of its evolution is probably best explained by a normal- and/or sub-diffusion processes.

This scenario may explain why the sub- and normal- diffusive MBPs were almost the only diffusion regimes reported in the literature before ≈ 2000 (Berger et al. 1998; Cadavid et al. 1999; Lawrence et al. 2001); older studies most likely concentrated on network areas in which MBPs are easier to observe compared to internetwork regions. For ground-based observations, another effect is also important: the residual aberrations and distortions due to variable seeing introduce an artificial turbulent motion into image time series. The artificial turbulence motion leads to slower growth of sd with time which in turn results in artificially small γ s (see Eq. (2)). It is worth noting that lower spatial resolution observations need not lead to smaller γ values and the larger scale flows (e.g. supergranular flows) would be detected more easily compared to smaller scale motions. This effect results in a larger γ value.

5.5. Summary

To summarise, we characterised the motions of SUNRISE Ca II H MBPs by turbulent diffusion theory. The MBPs (mean lifetime ≈ 8 min) were observed in seeing-free high-resolution image sequences in an internetwork area of the quiet Sun. A mean diffusion index of $\gamma = 1.69 \pm 0.08$ and a mean diffusion coefficient of $D = 257 \pm 32 \text{ km}^2 \text{ s}^{-1}$ were obtained. The γ corresponds to super-diffusion which describes the MBPs as features whose squared displacement (sd) from the first observed location grows faster than linearly with time. It is, to our knowledge, the largest γ value for MBPs reported in the literature so far. The parameter D lies within the range of decay rate of the magnetic field from MHD simulations and is among the largest D values obtained for small-scale magnetic features found in the literature. We found that D increases as the timescale increases, but generally lies in the range of those obtained by other investigations for larger spatial extent and longer durations. The migration of relatively short-lived features such as MBPs is composed of a superposition of random motions due to granular evolution and intergranular turbulence and systematic motions due to more steady granular evolution, mesogranular, and supergranular flows.

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