Neighboring pallidal neurons do not exhibit more synchronous oscillations than remote ones in the MPTP primate model of Parkinson’s disease

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INTRODUCTION

The basal ganglia are a set of subcortical nuclei which play an important role in the control of behavior. This can be seen in both normal physiology, as well as in several movement disorders, the most prominent being Parkinson’s disease (PD). The basal ganglia receive inputs from cortical and thalamic areas, and in a mostly feed-forward manner innervate the brainstem motor nuclei and the motor thalamic nuclei which close the loop back to the motor cortices (Haber and Gdowski, 2004). Anatomically, there is a high level of convergence of inputs from the cortex to the input nuclei of the basal ganglia, the striatum and the subthalamic nucleus. Further convergence exists in the next stages of the basal ganglia, which consist of both segments of the globus pallidus (GP) as well as the substantia nigra pars reticulata (SNr). The external segment of the globus pallidus (GPe) is an internal nucleus within the basal ganglia network, whereas the internal segment of the globus pallidus (GPI) and the SNr form the network’s output nuclei. Both the GPe and the GPI are relatively small nuclei, and have approximately two orders of magnitude fewer neurons than their major input – the striatum (Percheron et al., 1987; Bar-Gad et al., 2003b).

In the healthy primate, the firing pattern of most GP neurons is irregular (DeLong, 1971). Despite the high level of convergence of input innervations (Percheron et al., 1984) and inhibitory collaterals (Kita and Kita, 1994; Sato et al., 2000), GP neurons exhibit negligible synchrony (Nini et al., 1995; Raz et al., 2000). This is the case even between neighboring cells (Bar-Gad et al., 2003a), despite their high probability of sharing at least some of the input and/or having anatomical lateral connections.

In both human PD patients and monkeys treated with the neurotoxin 1-methyl-4-phenyl-1,2,3,6-tetrahydropyridine (MPTP), these properties change drastically. Synchrony appears between pairs of GP neurons, and the firing of many of them is dominated...
by episodes of periodic bursts. Typically, these periodic oscillations occur within two narrow frequency bands, approximately the tremor frequency (around 5–7 Hz) and double-tremor frequencies (Nini et al., 1995; Raz et al., 2000; Levy et al., 2002). In the rodents unilaterally injected with 6-hydroxy-dopamine (6-OHDA), the level of synchrony was found to moderately decrease with the distance (Mallet et al., 2008). However, no similar study has apparently been conducted in Parkinsonian primates.

Traditional studies of single neuron activity in awake behaving animals involve extracellular electrophysiological recordings which are subjected to detection and sorting of action potential (spike) waveforms (Abeles and Goldstein, 1977; Lewicki, 1998). However, the detection of spikes of neighboring neurons recorded by the same electrode, is affected by the shadowing effect (Bar-Gad et al., 2001); namely, when two neurons fire approximately together, the sorting process fails to identify both action potentials (Figure 1A). Although stereo-recording (e.g., by tetrode) might reduce the shadowing effect, they don’t completely eradicate it (Harris et al., 2000). The shadowing effect distorts the spike to spike cross-correlation function of the two neurons, and may lead to false appearance of synchronization (Bar-Gad et al., 2001).

Bar-Gad et al. (2003a) developed a method to estimate the cross-correlogram of an unsynchronized pair of neighboring neurons (with shadowing effect), assuming the autocorrelation functions are known (Bar-Gad et al., 2001). They concluded that cross-correlograms which are significantly different from the expected function are therefore synchronous. However, in the Parkinsonian state, the autocorrelation functions might be severely affected by the shadowing effects, and therefore the real (i.e., unshadowed) autocorrelation functions are unknown. We cannot assume that the autocorrelation functions are flat (as done in the normal state) because of the known pathological oscillatory firing. Therefore a different approach is required.

Here we confirm that the shadowing effect can create false synchronisation, as reported by Bar-Gad et al. (2001). However, we also show that in other cases, this undesired alteration in data can also conceal existing synchrony. Therefore, one cannot assume that the extent of correlation found under the shadowing effect represents an upper bound for the real correlations. We thus present a new method which enables a balanced comparison of close and remote pairs by artificially shadowing the remote pairs. This method creates two populations of pairs, which differ only in the distance between the cells, but not in the shadowing effect. Thereby, we can compare them and estimate the spatial distribution of the synchrony in the GP of the MPTP treated primate. Understanding the spatiotemporal pattern of the pallidal synchrony in the Parkinsonian primate can shed new light on the basal ganglia functional connectivity. Furthermore, these patterns may help clarify what role these pathological oscillations play downstream to the basal ganglia.

**MATERIALS AND METHODS**

**RECORDING PROCEDURES AND MPTP INJECTIONS**

The experiments were carried out on two vervet monkeys (Cu and S, Cercopithecus aethiops aethiops, females, weighing 3.8 and 3.6 kg, respectively). Primates’ care and surgical procedures were in accordance with the *NIH Guide for the Care and Use of Laboratory Animals* (1996) and the Hebrew University Guidelines for the Use and Care of Laboratory Animals in Research, supervised by the Institutional Committee for Animal Care and Use.

Both monkeys were trained to sit in a monkey chair; monkey Cu was also trained to perform a self-initiated button-pressing task to obtain a liquid reward. After training, a recording chamber was attached to the monkey’s skull under surgical procedure. Details of the surgery and data recording of monkey Cu methods were given previously (Heimer et al., 2002, 2006; Bar-Gad et al., 2003a; Elia et al., 2007). Surgery and data recording details of monkey S were identical to those performed on monkey Cu, unless stated otherwise. Briefly, a Cilux recording chamber (Cu – 18 mm round base; S – 27 mm square base) was tilted 50° laterally in the coronal plane and was positioned by a stereotactic device to cover most of the pallidal area (Contreras et al., 1981; Szabo and Cowan, 1984; Martin and Bowden, 2000).

The exact position of the chamber was verified using an MRI scan and electrophysiological mapping.

On each recording day, glass-coated tungsten electrodes (127 μm diameter tungsten rods coated with a layer of approximately 100 μm glass, yielding a total diameter of ∼350 μm for a length of 8–10 cm), confined to a concentric guide tube, were inserted into the brain. The electrodes’ relative and absolute location was a consequence of the chamber’s tilted angle. Therefore, distances on the plane tangent to the camber (and therefore the cranium) are defined “horizontal”, and distances perpendicular to the cranium are defined “vertical”.

The horizontal distance between the different electrodes was defined by the spatial arrangement of the electrodes within the guide: for monkey S the four electrodes were confined in a 570-μm inner diameter guide (i.e., 2 × 2 square with a 410-μm side). For monkey Cu the eight electrodes were confined in a 1420-μm inner diameter guide (i.e., one central electrode with a distance of 505 μm from each of the seven surrounding ones). Layout design can be found at: http://www.alphaomega-eng.com/pr_site/mt_layout/mt_designs.htm.

A mechanical micro-drive (EPS – Electrode Positioning System 1.28, Alpha-Omega Engineering, Nazareth, Israel) was used to lower each electrode individually through the dura into the brain while tracking its depth. Inside the recording target, we manipulated the electrode depth until an optimal and stable signal to noise ratio of spiking activity had been achieved. Vertical distances between the electrodes were therefore variable and typically ranged between 0 and 2000 μm. Distance between electrodes tips was estimated based on depth and horizontal distances. However, it should be stressed that this depth is subject to inaccuracies resulting mainly from the electrodes flexibility. This might be more pronounced in our settings, in which long (8–10 cm), and thin (350 μm diameter) electrodes were used, in order to minimize the spatial spread of the recording tips of the microelectrodes.

After recordings in the normal state, the monkeys were rendered Parkinsonian by five intramuscular injections of 0.4 mg/kg of MPTP–HCl dissolved in ∼0.5 ml of normal saline (Cu: Aldrich, Milwaukee, WI, USA; S: Sigma, Israel) over a period of 4 days. The monkeys were clinically assessed on a regular basis using a modified primate clinical staging scale (Benazzouz et al., 1995; Imbert et al., 2000). The monkeys developed severe Parkinsonism 5 days after initiation of MPTP treatment, and received an average score...
and other anatomical/physiological structures identified along the
electrode trajectory (e.g., border cells). Pallidal cells which could not
be classified as GPe or GPI neurons were considered GP neurons.

**RECORDING AND SPIKE SORTING**
The electrode output was sorted and classified in real time by a
template-matching algorithm (Cu: MSD, S: ASD; Alpha-Omega
Engineering, Nazareth, Israel). In addition, the electrode output
was also bandpass filtered (Cu: 300–6000 Hz, S: 250–5000, MCP-
plus; Alpha-Omega Engineering) and continuously sampled (Cu:
24 kHz, S: 40 kHz, AlphaMap; Alpha-Omega Engineering). The
continuous sampling of the electrode output of monkey Cu was
also subjected to an offline spike-sorting procedure (AlphaSort;
Alpha-Omega Engineering) based on principal-component analy-
sis of the spike patterns (Abeles and Goldstein, 1977). We applied
two sets of principal components (PCs), a default based on cortical
recordings (Abeles and Goldstein, 1977) and a set of PCs that were
created by a library of 131 waveforms of well isolated pallidal spikes
recorded by the same setup (Bar-Gad et al., 2003a). Offline sorting
was performed using both cortical and pallidal PCs, and the best
result was taken for each pair.

Both offline (monkey Cu) and online (monkey S) sorting was
verified by the existence of a refractory period in the inter-spike
interval histogram (cells in which more than 2% of the total inter-
spike intervals were less than 2 ms were excluded) and by the stabil-
ity of the firing rate of the cells (inspected by human observer). In
the stability analysis, the instantaneous firing rate of the neurons as
a function of time was displayed for the entire period of recording,
and the largest segment of stable data was selected for additional
analysis. Spike times were saved at the sampling rate of the analog
data (Cu: 24 kHz; S: 40 kHz).

**DATA ANALYSIS**
All data analysis was done on custom software using Matlab (The
MathWorks). Only pairs of neurons which were recorded simul-
taneously and fulfilled the isolation criteria (refractory period and
stability of discharge rate) for at least 150 s were included. For
analysis of the shadowing effect, which is a phenomenon at the
sub-millisecond order, we used the original sampling rate. For the
inspection of synchronous oscillations, we down sampled the data
to a sampling rate of 1 kHz.

We used auto and cross-correlation functions for time domain
analyses. The cross correlation of all pairs of cells recorded by the
same electrode showed a clear shadowing effect (Figure 1A). We
used a global shuffling method for the frequency domain analysis
(Rivlin-Etzion et al., 2006), which reflects this domain more reliably
than standard spectrum or coherence functions by compensating
for the effect of the refractory period. This method is especially use-
ful for bursting oscillations, which are typical at the Parkinsonian
pallidum. Similar to previous work (Heimer et al., 2006), we used
a frequency resolution of 0.25 Hz; shuffling was repeated 20 times;
the confidence interval was set based on the flat high frequency
range (270–300 Hz); and the range of interest for significant oscil-
lations was 4.5–30 Hz. Since we used shadowed spike trains, which
rendered noisy spectrograms especially in the low frequencies, we
defined a higher threshold than described previously (Rivlin-Etzion
et al., 2006). To define a significant oscillation in this study, a seg-
ment of three consecutive samples were required to cross a con-
fidence level of \( p = 10^{-3} \). Furthermore, we required that the local
maximum of an interval which crossed the confidence level would
be within the range of interest (4.5–30 Hz).

Single neurons were defined “oscillatory” if they had at least
one segment in their auto-spectrum function which fulfilled the
criteria detailed above. Similarly, close pairs of neurons were defined
“synchronized” if they had at least one segment in their cross-
spectrum which fulfilled these criteria. Remote pairs of neurons
were graded between 0 and 1 (synchronization index), according to
their probability of being detected as “synchronized” had they been
affected by the shadowing effect (see below). For further analysis, a
non-synchronization index was also defined, such that it completed
the synchronization index to 1 (i.e., their probability of not being
detected as synchronized had they been affected by shadowing).
For a synchronized pair, the oscillation frequency was defined as
the frequency of the maximal peak in the cross-spectrum (within
the range of interest).

**ESTIMATING THE SHADOWING EFFECT**
In order to mimic the shadowing effect on remote cells, we first
estimated the shadowing effect of the close pairs (see details in
the Appendix). Briefly, we used previously derived mathemati-
cal equations (Eqs 1 and 2 in the Appendix) which estimate the
recorded cross-correlation function and recorded firing rate,
given the probability to miss a spike in each time sample (termed
“shadowing vector”) and the original “true” firing rates of the
two cells (Bar-Gad et al., 2001). Based on the same formula we
estimated the shadowing vector (as well as the original firing
rates), given the recorded firing rate and cross correlation. This
can be done under several assumptions, which are necessary
for mathematically extracting the shadowing vector (see details
in the Appendix). First, we assume that the shadowing effect is
shorter than the refractory period. Second, the shadowing vectors
of a pair of neurons are symmetric; i.e., the shadowing vector of
the first cell at time \( t \) equals the shadowing vector of the second
cell at time minus \( t \). Finally, we assume that during complete
shadowing (i.e., at time samples in which the shadowing vector,
or the lost spike probability, equals 1), no spike was detected at
either of the cells. This process yielded one shadowing vector
per close pair.

To test the shadowing effect estimation method, we used data
recorded from the two monkeys, where spikes are detected and
sorted with different methods. Offline sorting (monkey Cu) yielded
a longer shadowing period than online sorting (monkey S), appar-
ently reflecting the longer “dead time” required by the offline sort-
ing algorithm. The averaged shadowing vectors of monkeys Cu and
The representative vectors were chosen in the following way: all the shadowing vectors were pooled together, and for each time sample, five values were chosen such that they divided the amplitude of the shadowing vector at each time sample into six evenly spaced shares. This approach (known as six-quantiles), is a natural extension of the median concept (achieved by dividing the data into two equal shares), and still provides a good representation of the population diversity. As stated above, this was done separately for each monkey.

Since shadowing is stochastic in nature (there are points in which the probability of missing a spike lies between 0 and 1), we repeated the artificial shadowing with a given shadowing vector five times and averaged the resulting time and frequency correlation functions. This yielded a single correlation function per shadowing vector. All in all, we had five cross-spectrum functions for any given pair of remote neurons, one from each representative shadowing vector. The relative part out of these five functions which was considered synchronous (according to the parameters S are shown in Figure 1B. During the analysis of the remote pairs of each monkey, we used the shadowing vectors derived from the close pairs of the same monkey.

**MIMICKING THE SHADOWING EFFECT**

Artificial shadowing of a pair of spike trains recorded by different electrodes, by a given shadowing vector, was done by dropping spikes from these spike trains with the probability derived from the shadowing vector. A simplified approach would have been to take the average shadowing vector and use it to shadow the remote pairs. However, this would not have fully represented the diversity of the different shadowing vectors. Ideally, we should have artificially shadowed each remote pair with all possible shadowing vectors (as calculated for all the close pairs of a particular monkey). However, repeating the artificial shadowing and all further analyses with all shadowing vectors for all remote pairs was computationally heavy. Therefore, we chose five representative shadowing vectors (per monkey in this study), and applied them to the remote pairs.

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mentioned above), was defined as synchronization index of that pair. This index represents the probability \((p = 0, 0.2, 0.4, \ldots, 1)\) of a remote pair to be detected as synchronized, had it been affected by the shadowing effect. A complementary non-synchronization index was also defined, such that it completes the synchronization index to 1. Close pairs were classified as either synchronized or non-synchronized (which is equivalent to a synchronization index of 1 or 0 respectively). Figures 1C,D illustrates the artificial shadowing of a remote pair by one of the five shadowing vectors (for simplified visualization). Applying the shadowing vectors yielded the short term artifacts observed in the close pairs, as planned.

**STATISTICAL COMPARISON OF CLOSE AND REMOTE PAIRS**

Comparison of two populations (close vs. remote pairs) divided into two categories (synchronized vs. non-synchronized) is typically done by a \(2 \times 2\) contingency table. This method was easily expanded in our continuous case by replacing the discrete number of synchronized pairs by the sum of the synchronization index, which could be a fraction. Similarly, the number of non-synchronized pairs was replaced by the sum of the non-synchronization index. This yielded a continuous \(2 \times 2\) contingency table per recording session.

For each pallidal population, we obtained an \(N \times 2 \times 2\) hypermatrix, where \(N\) represents the number of recording sessions (\(N = 16\) and 33 recording sessions for monkey Cu and S respectively). First, we summed the results over the \(N\) sessions. This yielded, for each population (GPe, GPi, GPi-POOLED), a \(2 \times 2\) continuous version of the contingency table previously described. To test the hypothesis that both close and remote pairs originate from the same distribution, we used a two-tailed statistical Fisher’s exact test, with the null hypothesis that close pairs are as synchronized as remote ones (the \(\chi^2\) test was not suitable because of the relative small \(N\) for the GPi population).

However, since the pallidal oscillations in the Parkinsonian monkey are episodic in nature, this analysis method might distort the results. For example, if during oscillatory episodes more remote pairs were recorded, this might create the impression that they have a higher probability for synchronous oscillations. Furthermore, in an analysis of three or more simultaneously recorded neurons, not all comparisons are independent (e.g., if there is synchrony between the neuron pair A–B and B–C, one would probably find a synchrony between the A–C pair). Therefore, counting correlated pairs might again bias the results.

To overcome this, we used additionally a different statistical method. We averaged the synchronization index of the close and remote populations, over all \(N\) sessions. This method yielded the average synchrony percentage per recording session (for close and remote pairs separately), rather than the total percentage of synchronized pairs. Here, we used a statistical two-tailed Student’s \(t\)-test, comparing the session’s synchronicity. As before, our null hypothesis was that in an average session, the fraction of synchronized pairs of close neurons was identical to that of the remote pairs.

**RESULTS**

**SHADOWING EFFECT IN THE TIME AND FREQUENCY DOMAIN**

Previous studies (Bar-Gad et al., 2001) have revealed that the shadowing effect can induce apparent synchrony between neurons recorded simultaneously from the same electrode. These effects are visible in the time domain, but are clearer in the frequency domain. Figure 2A presents the temporal (first row) and frequency (second row) correlation of a remote pallidal pair. The insets present the autocorrelation and auto-spectrum of the two cells. In both domains, the pair shows no significant synchronous oscillations (although one of the cells shows oscillations). The effect of artificial shadowing on the correlation and the cross-spectrum functions of this pair is shown in Figure 2B. Both functions (in the time and frequency domains) present clear and statistically significant (here calculated only for the frequency domain) synchronous oscillations at approximately 10 Hz.

Here we show that the shadowing effect can also have the opposite effect and mask existing synchrony. This opposite phenomenon is presented in Figure 3. Here, a remote pair displays significant synchronous oscillations (Figure 3A). However, after artificial shadowing by one of the five possible shadowing vectors (Figure 3B), the peak in the cross-spectrum function is smoothed and the cross-spectrum does not meet the significance criteria. Thus, the artificial shadowing by this shadowing vector results in no-detection of synchronous oscillations of this pair of neurons. Application of the other four representative shadowing vectors here also masked the synchrony and yielded a synchronization index of 0.

Thus, the net effect of the shadowing on the detection of neuronal synchronous oscillations is not unidirectional and two error types are possible. First, there are false positive errors; i.e., detection of significant synchronous oscillations in the shadowed pair which were not there in the original spike trains. Second, there are false negative errors; i.e., synchronous oscillations that were missed because of the shadowing effect. The fraction of the two errors was found to differ between the two monkeys, probably because of the difference in sorting method of the close pairs and the resulting shadowing vectors (Figure 1B). For monkey S \((n = 91\) pairs, sorted online, and \(-2\) ms duration shadowing effect), P (false positive) = 0.074, and P (false negative) = 0.171, whereas for monkey Cu \((n = 349\) pairs, sorted offline, and \(-3.5\) ms shadowing effect), P (false positive) = 0.27, and P (false negative) = 0.309. Apparently, the longer duration of the shadowing vector originating from the offline sorting process led to the higher error values in this case.

Next, we wanted to further understand the connection between these types of errors and the firing properties of each of the neurons in a given pair. To do so, we calculated the probabilities of having zero, one, or two of the cells significantly oscillatory in each of the pairs. This was done given the probability of having one of the possible types of errors, as well as the two correct possibilities (correct detection and rejection of synchronous oscillations). These results are summed at Table 1. The most striking result is that all cases in which the shadowing effect created apparent synchrony occurred when only one of the cells in the pair was significantly oscillatory. On the other hand, about half of the cases in which synchrony was concealed by the shadowing effect occurred when none of the cells in the pair was considered oscillatory.

Thus overall, the shadowing effect on neighboring neurons is not one-sided, and it induces both false detection of synchronous oscillations, as well as misses of existing synchronous oscillations. Therefore, a proper comparison of the close pairs to the remote ones should balance this effect.
SYNCHRONY IN THE DIFFERENT PALLIDAL POPULATIONS

We divided the pallidal cell population into three groups – GPe, GPi, and GP (i.e., unclassified cells within the GP). We analyzed separately the GPe pairs, and the GPi pairs, and then pooled all three pallidal populations together (“GP-pooled”).

We used two different statistical approaches in the comparison of close and remote pairs (see details in the Materials and Methods). The results of the contingency table analysis, which pools all pairs and ignores the sessions in which they were recorded, are summarized in Table 2. In the pooled data (as in most sub-groups), the fraction of synchrony in the close pairs (20.5%) was slightly higher than in the remote one without shadowing (17.7%). After application of the shadowing effects to the remote pairs (to balance this effect on the close pairs) we found that the fraction of synchrony in the close pairs (20.5%) was actually lower than that in the remote ones (32.2%). Thus, to our surprise, we failed to reject the null hypothesis that the fraction of synchrony in the close pairs is equal to that found at the remote ones (after artificial shadowing).

Similar trend was achieved in the second analysis method of averaging the synchrony over the recording sessions. This method has the advantage of compensating the episodic nature of the oscillations (see Materials and Methods). These results are summarized in Table 3. In the pooled data, the average synchrony of the close pairs (17.2%) was smaller than that of the remote ones, both before and after artificial shadowing (29.2 and 35.4%, respectively).

Both analysis methods yielded similar results. After artificial shadowing, the remote pairs tend to be more synchronized than the close ones. This was the case for both monkeys, in both the GPe and the GP-pooled population. In the GPi we observed opposite trends in the two monkeys, but the number of close pairs in our

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**FIGURE 2 | Shadowing effect induces artificial synchronous oscillations.**

(A) A remote pair (same pair as in Figures 1C,D) which does not exhibit significant synchronous oscillations. First row – cross-correlation function (inset: auto-correlation functions); second row – cross power spectral density (CPSD, inset: auto-spectral densities). Note that one of the cells is oscillatory whereas the other one is not. (B) The same pair as in a, after artificial shadowing by one of the representative shadowing vectors. At the time domain (first row) we can see the large deflection around $t = 0$ which is caused by the shadowing effect. Both time domain and frequency domain (second row) demonstrate synchronous oscillations (insets show auto-correlations and auto-spectrum after shadowing). Confidence levels in the frequency domain are marked with a broken line. Significant samples (at least three continuous) are marked with a circle.
Figure 3 | Shadowing effect conceals existing synchronous oscillations. (A) A remote pair (monkey Cu) with significant synchronous oscillations in both the time (first row) and frequency (second row) domains. Insets are auto correlation and auto spectrum functions. Both cells exhibit oscillatory firing patterns. (B) The same pair as in (A), after artificial shadowing by one of the representative shadowing vectors (same one as in Figure 2). Similarly to Figure 2B, we can see the deflection at t = 0 (in the time domain, first row) caused by the shadowing effect. In both the time domain and the frequency domain (second row), the shadowing effect decreased the ~10 Hz synchronous oscillations to an insignificant level. Conventions are the same as in Figure 2.

Table 1 | The shadowing effect and oscillatory activity of the single cells.

| Number of significantly oscillating cells in the pair | Synchronous oscillations before and after artificial shadowing ("hit") (%) | Artificial shadowing concealed existing synchronous oscillations ("miss") (%) | Shadowing created apparent synchronous oscillations ("false alarm") (%) | No synchronous oscillations before and after artificial shadowing ("correct rejection") (%) |
|------------------------------------------------------|-------------------------------------------------------------------------|--------------------------------------------------------------------------|---------------------------------------------------------------------|-----------------------------------------------------------------------------------|
| None                                                 | 1.7                                                                     | 52.1                                                                     | 0.0                                                                 | 88.1                                                                              |
| Only one                                             | 36.0                                                                    | 38.1                                                                     | 100.0                                                               | 10.0                                                                             |
| Both                                                 | 62.3                                                                    | 9.8                                                                      | 0.0                                                                 | 0.9                                                                               |

Relationships between firing properties of each of the single neurons in a given pair, and the outcome of the shadowing effect. The left column gives the number of cells within a pair that were defined as oscillatory. Each of the next columns addresses one of the four possibilities of detection of synchronous oscillations given the shadowing effect (signal detection theory). That is, synchronous oscillations were detected before and after shadowing ("hit"); synchronous oscillations were concealed by the shadowing ("miss"); shadowing created apparent synchronous oscillations ("false alarm"); or there were no synchronous oscillations both before and after synchronous oscillations ("correct rejection"). We present here the probability of having none, one, or both oscillatory cells within the pair, given each one of the four possibilities (so that each column is summed to 100%). We can see that all “false alarms” occurred when only one of the cells in the pair was oscillatory. On the other hand, about half of the “misses” occurred when none of the cells in the pair were significantly oscillatory.
Similarly to the previously described comparisons of close and remote pairs, we first compared the ratio of synchronous pairs (Table 4). For the GPe, GPi, and GP-pooled populations the closer (i.e., less than 750 μm apart) pairs were slightly but insignificantly less synchronized than the further distant ones (15.8% and 20.0% of the GP-pooled pairs, respectively). Finally, we compared the average percentage of synchrony per recording session (Table 5). Again, no significant difference was found between closer and further distant pairs in GPe, GPi, and GP-pooled populations (for GP-pooled population, average synchrony of 29.9% in the closer pairs, in comparison with 28.5% in those further distant).

To further validate our results, we divided the remote pairs’ population into two subpopulations. This was done according to the estimated distance between the electrode tips, with a cutoff between categories at 750 μm. We preferred to use two categories, rather than continuous spectrum, since this distance estimation suffers a certain noise, resulting from the electrodes flexibility. Since both subpopulations were recorded by different electrodes, we compared the synchrony of the original, unshadowed signal.

| Monkey | Ratio of synchronous oscillating pairs out of total population of pairs: comparison between neighboring and remote cells. | Fisher’s exact test p-value (unshadowed in parentheses) |
|--------|-------------------------------------------------|-----------------------------------------------|
|        | Monkey | Ratio of synchronous oscillating pairs out of total population of pairs |        | Close | Remote shadowed (unshadowed in parentheses) |        | Close | Remote shadowed (unshadowed in parentheses) |        |
| GPe    | Cu     | 3/19  | 15.8 | 29.4 (7)/118 | 24.9 (5.9) | 0.562 (0.144) |
|        | S      | 2/13  | 15.4 | 2.8 (4)/53  | 5.3 (75)   | 0.253 (0.337) |
|        | Both   | 5/32  | 15.6 | 32.2 (11)/171| 18.8 (6.4) | 0.806 (0.142) |
| GPi    | Cu     | 3/3   | 100  | 15.0 (22)/23 | 65.2 (95.7)| 0.520 (1)    |
|        | S      | 0/2   | 0    | 10.8 (10)/14| 77.1 (71.4)| 0.018 (0.125)|
|        | Both   | 3/5   | 60   | 25.8 (32)/37 | 69.7 (86.5)| 0.637 (0.188) |
| GP-pooled | Cu   | 6/24  | 25   | 119.2 (57)/349| 34.2 (16.3)| 0.503 (0.265) |
|         | S     | 2/15  | 13.3 | 22.6 (21)/91 | 24.8 (23.1)| 0.513 (0.515) |
|         | Both  | 8/39  | 20.5 | 141.8 (78)/440| 32.2 (177) | 0.151 (0.664) |

Ratio of synchronous oscillating pairs out of total population of pairs in the different nuclei. Each artificially shadowed pair was scored between 0 and 1 based on the synchronization index. The sum of this score for the whole population represents the number of pairs which were significantly oscillatory. For the remote pairs we also present, in parentheses, the results of the original, unshadowed pairs. A two-tailed Fisher’s exact test was performed under the null hypothesis of an identical fraction of synchrony in the close and the remote pairs.

| Monkey | Average percentage of synchronous oscillating pairs per session: comparison between neighboring and remote cells. | t-Test p-value (unshadowed in parentheses) |
|--------|-------------------------------------------------|-----------------------------------------------|
|        | Monkey | Number of sessions | Average percentage of synchronous oscillating pairs per session | t-Test p-value | Close | Remote | Close | Remote shadowed (unshadowed in parentheses) |        | (unshadowed in parentheses) |
| GPe    | Cu     | 10   | 10.0 | 23.5 (6.5) | 0.179 (0.674) |
|        | S      | 13   | 20   | 6.9 (12.0) | 0.477 (0.790) |
|        | Both   | 23   | 30   | 13.0 | 12.5 (10.2) | 0.941 (0.725) |
| GPi    | Cu     | 2    | 3    | 100.0 | 678 (97.4) | 0.108 (0.423) |
|        | S      | 2    | 10   | 0.0 | 79.3 (70.0) | 0.0002 (0.001) |
|        | Both   | 4    | 13   | 50.0 | 76.6 (76.3) | 0.437 (0.446) |
| GP-pooled | Cu   | 14   | 16   | 21.4 | 39.0 (20.3) | 0.141 (0.932) |
|         | S     | 15   | 33   | 13.3 | 33.6 (33.5) | 0.098 (0.105) |
|         | Both  | 29   | 49   | 17.2 | 35.4 (29.2) | 0.040 (0.188) |

Percentage of synchronous oscillating pairs in an average session in the different nuclei (regardless of the number of pairs recorded in it). The percentage of significantly synchronous oscillating pairs per session is presented in the second and third columns from the right. For the remote pairs, the percentage of pairs after artificial shadowing is presented, with the percentage of original, unshadowed pairs in parentheses. Similar to Table 1, the two-tailed t-test hypothesis is that close pairs have the same percentage of synchrony as remote ones.

database was very small (three and two close pairs in monkeys Cu and S, respectively). We conclude that in the MPTP treated primate, close pairs of GP neurons have a similar tendency for synchronous oscillations as remote pairs.

To further validate our results, we divided the remote pairs’ population into two subpopulations. This was done according to the estimated distance between the electrode tips, with a cutoff between categories at 750 μm. We preferred to use two categories, rather than continuous spectrum, since this distance estimation suffers a certain noise, resulting from the electrodes flexibility. Since both subpopulations were recorded by different electrodes, we compared the synchrony of the original, unshadowed signal.
FREQUENCY DISTRIBUTION OF SYNCHRONOUS OSCILLATIONS

Next, we compared the frequency of the synchronous oscillations in the two populations (close and remote pairs). As previously mentioned, most oscillations appeared in two frequency bands: approximately 5–7.5 and 10–12.5 Hz, which are roughly the tremor frequency and its double. Again, for the artificially shadowed remote pairs, we calculated the oscillation frequency under each of the representative shadowing vectors, and weighted them accordingly.

Figure 4 shows the distribution of peak frequency in the oscillatory pairs in the close and the artificially shadowed remote population. Since the population of close pairs which showed significant synchronous oscillations was relatively small, we pooled

Table 4 | Ratio of synchronous oscillating remote pairs in two distance categories.

| Monkey | Ratio of synchronous oscillating pairs out of total population of pairs | Fisher’s exact test p-value |
|--------|---------------------------------------------------------------|---------------------------|
|        | <750 μm | >750 μm |                             |                           |
|        | Ratio    | %       | Ratio    | %       | p-value |
| GPe    |          |         |          |         |         |
| Cu     | 5/69     | 7.2     | 2/49     | 4.1     | 0.38    |
| S      | 2/40     | 5.0     | 2/13     | 15.4    | 0.25    |
| Both   | 7/109    | 6.4     | 4/62     | 6.5     | 0.61    |
| GPI    |          |         |          |         |         |
| Cu     | 5/5      | 100.0   | 17/18    | 94.4    | 1.00    |
| S      | 9/12     | 75.0    | 1/2      | 50.0    | 0.51    |
| Both   | 14/17    | 82.4    | 18/20    | 90.0    | 0.42    |
| GP-pooled |       |         |          |         |         |
| Cu     | 25/180   | 13.9    | 32/169   | 18.9    | 0.13    |
| S      | 13/60    | 21.7    | 8/31     | 25.8    | 0.42    |
| Both   | 38/240   | 15.8    | 40/200   | 20.0    | 0.16    |

Ratio of synchronous oscillating pairs out of the entire remote pairs population in the different nuclei. Similarly to Table 2, a two-tailed Fisher’s exact test was performed under the null hypothesis of an identical fraction of synchrony in the two distance categories. Distances were estimated according to the assumed three dimensional locations of the electrode tips, when these were available.

Figure 4 | Frequency distribution of the pairs with significant synchronous oscillations. The close pairs are compared with the artificially shadowed remote pairs; distributions are presented in a 2.5-Hz resolution. Left – GPe, based on five close pairs and an average of 32.2 remote ones; Middle – GPI – based on three close pairs and an average of 25.8 remote ones; Right – GP-pooled population, based on eight close pairs and an average of 110.6 remote ones. Fractional number of remote pairs is the result of different shadowing vectors induced on each remote pair during the artificial shadowing. Most pairs oscillate synchronously at the range of 5–7.5 Hz (approximately the tremor frequency, GPe and GP-pooled only) and 10–12.5 Hz (approximately double tremor frequency). In all three populations, the trend was similar in both the close and artificially shadowed remote pairs.

Table 5 | Average percentage of synchronous oscillating pairs per session in two distance categories.

| Monkey | Number of sessions | Average percentage of synchronous oscillating pairs per session | t-Test p-value |
|--------|--------------------|---------------------------------------------------------------|----------------|
|        | <750 μm | >750 μm | <750 μm | >750 μm |                             |                           |
| GPe    |          |         |          |         |                             |                           |
| Cu     | 9       | 8       | 3.3      | 8.3     | 0.55                        |
| S      | 14      | 8       | 2.9      | 25.0    | 0.10                        |
| Both   | 23      | 16      | 3.0      | 16.7    | 0.09                        |
| GPI    |          |         |          |         |                             |                           |
| Cu     | 2       | 3       | 100.0    | 97.2    | 0.50                        |
| S      | 9       | 2       | 77.8     | 50.0    | 0.48                        |
| Both   | 11      | 5       | 81.8     | 78.3    | 0.88                        |
| GP-pooled |      |         |          |         |                             |                           |
| Cu     | 15      | 13      | 21.0     | 19.1    | 0.88                        |
| S      | 27      | 15      | 34.8     | 36.7    | 0.90                        |
| Both   | 42      | 28      | 29.9     | 28.5    | 0.90                        |

Percentage of synchronous oscillating pairs in an average session in the different nuclei (regardless of the number of pairs recorded in it, similar to Table 3). The percentage of significantly synchronous oscillating pairs per session is presented in the second and third columns from the right. As before, the two-tailed t-test hypothesis is that pairs in the two distance categories have the same percentage of synchrony.
DISCUSSION
We analyzed correlations of close pairs of GP neurons recorded by the same electrode and compared their synchrony to that of pairs recorded by different electrodes (200–2000 μm apart) in the Parkinsonian MPTP treated primates. In addition we compared intermediate distant pairs (less than 750 μm) to further distant ones (over 750 μm). Our results indicate that pathological synchronization of pallidal activity in the MPTP treated monkeys does not fall off across a few millimeters of tissue. Furthermore, the frequency distribution of the significant synchronous oscillations of close and remote pairs was found to be similar. These findings are in line with our previous report of emergence of synchronous oscillations between striatal cholinergic interneurons (TANs) and pallidal neurons after MPTP treatment. In the normal state TANs have no correlation with pallidal activity. The high level of oscillatory correlated activity between random pairs of TAN and GP neuron suggests that coherent, homogeneous oscillations of the whole basal ganglia circuitry underlie the clinical features of the MPTP primates model of PD (Raz et al., 2001).

Along the line of previous work in the rodent 6-OHDA model of PD (Mallet et al., 2008), our results suggest that the pallidal synchrony spans most of these nuclei. However, Mallet et al. found that synchronization decreases with distance. One possible explanation for this discrepancy is the delicate method of multi-contact silicone probes which was used in that study. These probes cover a range of distances from a hundred of microns to a few millimeters, whereas our methods divide the population of neuron pairs in a binary fashion. Furthermore, the close pairs which were recorded in our study by the same electrode were probably within less than a 100-μm apart (Abeles, 1974; Gray et al., 1995). On the other hand, the primate pallidum is larger than the rodent pallidum, and therefore our “remote” units, even those in the subcategory of over 750 μm, might still be in the closer pallidal territories than the remote rodent ones. This difference in scales may underlie the discrepancy between the studies.

Alternatively, the discrepancy could be the result of differences in the model of PD. First, our recordings were made on awake animals, whereas Mallet et al. used urethane anesthetized animals. Next, our dopamine depletion model was systemic and therefore bilateral, whereas the 6-OHDA model was unilateral. Finally, MPTP treated vervet monkeys demonstrate all major Parkinsonian motor symptoms, including rigidity and Parkinsonian tremor. Unilaterally injected 6-OHDA rodents, on the other hand, mainly display motor asymmetry, which is manifested in an increased tendency for turning (Ungerstedt, 1968). The homogeneous level of synchronous oscillations found in the MPTP primate model could cause a broad effect of synchronous oscillations downstream to the basal ganglia; e.g., in the motor thalamus and motor cortex. This, in turn, could lead to a more pronounced clinical effect in the Parkinsonian vervets in comparison to rodents (Beal, 2001).

The homogeneous spatial pattern of pathological GP synchrony may assist in finding the origin of the phenomenon. We might conjecture that if the synchronous oscillations emerge within the GP (rather than upstream in the basal ganglia), there would be higher synchrony between close pairs than between remote ones. However, it was theoretically shown (DeLiang, 1995) that local interactions can cause global, homogeneous synchrony. Therefore, we believe that the attempt to derive the origin of the synchronous oscillations from our results requires a separate theoretical analysis. Such theoretical approach should take into consideration the specific properties of the GP, including its high firing rate and its predominantly inhibitory innervation.

We therefore suggest that future research should focus on two directions. First, in order to minimize the shadowing effect and maximize the separation quality of the different units, tetrodes, rather than single electrodes, should be used (Harris et al., 2000; Ecker et al., 2010). This is especially important since we cannot rule out the possibility that artificial shadowing have a different effect than the “natural” shadowing. It should be mentioned, though, that the shadowing effect still exists on tetrodes recordings. Hence, for a complete suppression of the shadowing effect one should use more than four contacts and/or a more advanced sorting algorithm (Takahashi and Sakurai, 2009; Franke et al., 2010). Second, to complete the comparison between rodent’s and primate’s spatial distribution of pallidal pathological synchronous oscillations, recordings within verified distances inside the primate and human (during DBS surgeries) pallidum should be performed (e.g., by multi-contact silicone probes).

Our work focused on the second order level of synchronous neural activity, i.e., pair wise correlations. At this level, no difference was found between close and remote pairs in the primate GP. However, this does not rule out the possibility that higher orders of synchrony differ between close and remote populations of neurons. Indeed, such a result was recently found in the primate visual cortex (Ohiorhenuan et al., 2010). Such an analysis with our data is not possible since recording of more than two neurons by a single electrode is extremely rare in the primate GP.

To conclude, our findings strongly imply that the pathological synchronous oscillations in the Parkinsonian MPTP treated primates have a homogeneous spatial pattern. This pattern probably contributes to the robust effect of the pallidal oscillations on their target downstream to the basal ganglia. This robust effect can underlie the severe motor abnormalities in PD.

ACKNOWLEDGMENTS
We thank G. Goelman for MRI, V. Sharkansky, S. Freeman, and A. Shaposhnikov for technical assistance, Y. Ben Shaul, M. Joshua, and Y. Prut for suggestions on early stages of this work and E. Singer for language editing. This study was partly supported by the Hebrew University Netherlands Association (HUNA) “Fighting against Parkinson” and the Vorst family foundation grant (HB), the Dekker foundation fellowship (BR); the Foulkes Foundation fellowship, and a fellowship from MYERS-JDC-BROOKDALE Institute of Gerontology and Human Development and ESHEL, the Association for the Planning and Development of Services for the Aged in Israel (SE); the Israel Science Foundation grant (YR) and by the “Hoffman Leadership and Responsibility” fellowship program at the Hebrew University (RM).
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Conflict of Interest Statement: The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

Received: 31 December 2010; paper pending published: 14 February 2011; accepted: 13 June 2011; published online: 30 June 2011.

Citation: Mitelman R, Rosin B, Zadka H, Slovik M, Heimer G, Ritov Y, Bergman H and Elias S (2011) Neighboring pallidal neurons do not exhibit more synchronous oscillations than remote ones in the MPTP primate model of Parkinson’s disease. Front. Syst. Neurosci. 5:44. doi: 10.3389/

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APPENDIX
THE MATHEMATICS OF ARTIFICIAL SHADOWING

Extracellular recording of spiking activity is based on detection of instantaneous events (spikes) in the high pass filtered analog signal and sorting them according to their pattern. This process is known as “spike sorting.” The underlying assumption is that different cells, recorded by the same electrode (neighboring cells), can be detected according to their different shape. However, spikes occurring within a few milliseconds cannot be detected, because of their shape overlap. This phenomenon was termed the “shadowing effect” by Bar-Gad et al. (2001).

The probability to miss a spike within a given spike train, as a function of the spike’s lag from a spike in the neighboring spike train, was termed the “shadowing vector.” Given the actual firing rate \( p \), the shadowing vector \( (S_{nm}) \) and the shadowing duration \( (\alpha) \) the observed firing rate \( (p') \), can be calculated as follows (Bar-Gad et al., 2001):

\[
p'_n = p_n \left(1 - p_n \sum_{t=-\alpha}^{\alpha} S_{nm}(t)\right)
\]

Furthermore, given the neurons’ actual firing rate \( (P_n, P_m) \), their autocorrelation \( (\alpha) \), and the shadowing vector \( (S) \), the cross-correlation function \( (c'_{nm}) \) of two uncorrelated spike trains recorded by the same electrode can be calculated as follows (Bar-Gad et al., 2001, Eq. 5):

\[
c'_{nm}(t) = \frac{P_n}{1 - P_n \sum_{t=-\alpha}^{\alpha} S_{nm}(\tau)u_{\alpha}(t+\tau)} \left(1 - \sum_{t=-\alpha}^{\alpha} S_{nm}(\tau)u_{\alpha}(t+\tau)\right) \left(1 - \sum_{t=-\alpha}^{\alpha} S_{nm}(\tau)u_{\alpha}(t-\tau)\right)
\]

We assume that the refractory period of the neurons is longer than the shadowing period, similar to Bar-Gad et al. (2003a). This means that the autocorrelation equals 1 at \( t = 0 \), and 0 for the rest of the shadowing period. This assumption further simplifies the cross-correlation estimation to:

\[
c'_{nm}(t) = \frac{P_n}{P_n} (1 - S_{nm}(-t))(1 - S_{nm}(t))
\]

We further assume that the shadowing period is symmetric, i.e., that \( S_{mn}(t) = S_{nm}(-t) = S(t) \), and then get Eq. (4):

\[
c'_{nm}(t) = \frac{P_m P_n}{P_n} (1 - S(t))^3
\]

The shadowing vector \( S(t) \) can be calculated from Eq. 4. Since \( S(t) \) is a probability vector, its values are limited to the interval \([0,1]\), hence we get a single solution:

\[
S(t) = 1 - \sqrt{\frac{P_m c'_{nm}(t)}{P_n P_m}}
\]

This allows an approximated solution, if \( P_m = P_n \) and \( c' = P_n \), as follows:

\[
S(t) = 1 - \sqrt{\frac{c'_{nm}(t)}{P_n}}
\]

Furthermore, we have a satisfied set of second degree equations. We have the following empirical data: \( P_n, P_m, \) and \( c'_{nm}(t) \) for \( -\alpha \leq t \leq \alpha \), whereas we are interested in \( S(t) \) for \( -\alpha \leq t \leq \alpha \), as well as \( P_n \) and \( P_m \). This yields \( 2\alpha + 1 \) equations (Eq. 5) for \( -\alpha \leq t \leq \alpha \), and two more equations (Eq. 1) for \( P_n \) and \( P_m \) — altogether \( 2\alpha + 3 \) equations for \( 2\alpha + 3 \) variables.

A set of equations of the second degree does not necessarily have an algebraic solution. Therefore, we used two numerical approaches. First, we used an iterative solution, which initiated in the approximated solution (Eq. 6), and iteratively used Eq. 5 to improve the solution. Second, we used the Matlab \texttt{fsolve} function for the solution of the equation. Both methods yielded similar solutions.

This solution assumes independence between the two spike trains. However, this term might be violated in our data. Therefore, we verified this method by estimating a shadowing vector from an artificially shadowed cross-correlogram. Different shadowing vectors were applied on all 349 pairs recorded by different electrodes (i.e., not originally distorted by the shadowing effect) from monkey Cu. Then we used the above method to estimate the shadowing vector, and calculated the error between known and estimated shadowing vectors. The absolute error per index in the vector was \( 0.017 \pm 0.036 \) (average \pm SD, values representing the probability of missing a spike, hence they vary between 0 and 1). Thus, the estimation of the shadowing effects is satisfactory for our purposes despite the possible violation of the assumption of independent spike train.