Time-Series Analysis of a Catalog of UFO Events: Evidence of a Local-Sidereal-Time Modulation

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Abstract—Analysis of a large database may yield patterns that are not obvious by studying individual cases. We here carry out a time-series analysis of a selection of the 1999 version of the Hatch UFO catalog, comprising 12,100 events, for which geographical and temporal data are well established. We compare our results with those of earlier analyses by Poher and Vallee on smaller datasets, and confirm that the event rate exhibits a modulation related to local sidereal time (LST). Such a modulation could in principle be simply a spurious effect due to the combined influence of a strong time-of-day modulation, and an expected time-of-year modulation. However, we develop a test to distinguish such a spurious pattern from an intrinsic pattern. This test indicates that the LST modulation is intrinsic.

1. Introduction

The UFO problem—that of understanding the cause of causes of UFO reports—is extraordinarily complicated. Many thousands of reports have been filed and investigated by the US Air Force (projects Sign, Grudge, and Blue Book), the French project SEPRA (first known as “GEPAN”), and more recently by projects in Brazil, Chile, Peru, and Uruguay. In addition, many thousands of reports have been filed by non-governmental organizations such as APRO, NICAP, and MUFON in the United States, and similar organizations in other countries. A brief guide to the extensive literature on this subject is given in Sturrock (1999, p. 381). The reader may learn of the early history from Jacobs (1975), and form some idea of the complexity of the subject from three encyclopedias (Clark, 1998; Sachs, 1980; Story, 1980).

It is significant that this complex problem has received scant attention from the scientific community. Possible reasons for this lack of interest were considered by a panel, convened by Mr. Laurance S. Rockefeller, that met at Pocantico, New York, in September 1997 (Sturrock, 1999: 153). It appears that one major reason for this lack of curiosity is the belief that the Air-Force-funded Colorado Project, which led to the “Condon Report” (Condon & Gillmor, 1969), established that there is no significant content to the problem. A panel set up the National Academy of Sciences to review the Condon Report gave the report a clean bill of health (Clemence et al., 1969), when there is much in the report to
be criticized (Sturrock, 1987). For instance, Condon's summary of the project is seriously at odds with the case studies and summaries of the Project staff.

However, a great deal of data has been collected since the Condon Report, and much of these data have been compiled in the form of computer-readable catalogs. Probably the first computerized catalog of UFO cases was developed by Dr. Jacques Vallee in the 1960s. Since that time, other catalogs have been developed, including notably UFOCAT, originally compiled by Dr. David Saunders as part of his work for the Colorado Project, and more recently developed by Don Johnson for the Center for UFO Studies (CUFOS); UNICAT, compiled by Dr. Willy Smith; AIRCAT, a catalog of aircraft cases compiled by Dr. Richard F. Haines; and the *U* Database compiled by Larry Hatch (1999). We have concentrated on the *U* database, since Hatch has kindly made his data available in a form convenient for analysis.

In the scientific study of any phenomenon, the theoretical challenge is to explain experimental or observational facts. In order to determine these facts, we need to identify consistent patterns in the data, and then establish that these patterns are intrinsic to the phenomenon, and not simply a product of the data-acquisition and compilation processes. Poher (1973), in a seminal report, analyzed a dataset of French UFO cases and a dataset of other worldwide cases, and established that a number of patterns are identical in both datasets. Poher and Vallee (1975) carried out a crucial extension of this analysis by combining and comparing the Poher catalog with that compiled independently by Vallee. These studies indicated that there are indeed consistent patterns in UFO data.

The purpose of this article is to extend the analyses of Poher and Vallee by analyzing the Hatch catalog, which is much more extensive and has been compiled independently of the compilations of Poher and Vallee. However, as a first step in this catalog analysis, we restrict the scope of this article to time-series analysis. Perhaps the most intriguing result of the time-series aspects of the Poher-Vallee analyses was the strong suggestion that the rate of occurrence of UFO events depends upon local sidereal time (LST). For this reason, the main focus of our present analysis is to determine whether analysis of the more extensive Hatch catalog confirms this pattern in the Poher and Vallee catalogs and, if so, to try to determine whether this pattern is intrinsic to the phenomenon, or extrinsic in the sense that it may be attributed to other known and understood properties of the dataset.

In Section 2, we analyze the rate of UFO events on a historical time-scale, by day of the week, by time of day, and by time of year. In Section 3, we carry out a power spectrum analysis and find a notable peak at $1y^{-1}$ (one cycle per year). Clearly, this may simply reflect a yearly variation that has a meteorological and/or sociological origin, but it may have an astronomical cause. In order to distinguish these two possibilities, we carry out a “running-wave” spectrum analysis in Section 4. This analysis clearly indicates that the variation has an astronomical origin. We carry out a significance test in Sections 5, and we examine the event rate as a function of LST in Section 6. We analyze two
simulated test cases in Section 7, and give a brief discussion of these results in Section 8.

2. Basic Patterns

Hatch kindly selected from his 1999 database a list of 12,100 cases for which the latitude, longitude, date, and time were known. We refer to each case as an “event,” leaving open the nature of the event or, indeed, whether the event really occurred or was a hoax or misperception of a well known object or phenomenon. The number of events is shown, as a function of year, in Figure 1 for 1890 to 1945, and in Figure 2 for 1945 to 1999. The cumulative count for 1890 to 1999 is shown in Figure 3. We see that there were only a few events before 1945, and there was a sharp onset in 1946. The peak at 1896 to 1897 may be part of the “airship” wave (see, for instance, Clark, 1998:46), and the peak at 1909 may be part of the “phantom aircraft” wave. Although the event rate has varied greatly from year to year since 1945, it is clear from Figure 3 that there was a very sharp increase in 1946. That was the year of the Kenneth Arnold report (see, for instance, Clark, 1998:139), which received worldwide attention and initiated the current “flying saucer” wave.

We now wish to examine the event rate by day of week, etc. We could take the entire database and form the histogram by day of week, etc. However, we wish to know whether there is any consistent pattern in the phenomenon, as compared with a pattern that comes and goes. We have therefore divided the

Fig. 1. Number of events per year in selected database from 1890 to 1945.
database into four time intervals A, B, C, and D, listed in Table 1, that contain equal numbers of events we refer to as "blocks."

Histograms according to day of the week, for the four time-blocks, are shown in Table 1, in which we also list the mean and standard deviation for each day of the week. The histograms are also shown in Figure 4. It is clear from this figure

![Figure 2](image1) Fig. 2. Number of events per year in selected database from 1890 to 1999.

![Figure 3](image2) Fig. 3. Cumulative number of events in selected database from 1890 to 1999.
that there is little evidence for a day-of-the-week pattern. We may check this conjecture by means of the familiar chi-square test. We form the statistic

\[
\chi^2 = \sum_{k=1}^{K} \frac{(n_k - n_k(\text{exp}))^2}{\sigma_k^2},
\]

(2.1)

where \( k \) enumerates days of the week, \( n_k \) is the mean count, and \( \sigma_k \) is the standard deviation of the count, for each day of the week.

If we consider the hypothesis that each event is independent of all other events, and can occur on any day of the week with uniform probability, then the appropriate choice for \( n_k(\text{exp}) \) is 12,100/7, i.e. 1,729, and the appropriate choice for \( \sigma_k \) is \( \sqrt{1729} \), i.e. 41.6. For this case, Equation 2.1 gives \( \chi^2 = 18.1 \). For 6 degrees of freedom, the probability of obtaining this large a value of \( \chi^2 \) or more by chance is 0.006. Hence the hypothesis may be rejected, but not by overwhelming evidence.

We now look to see if there is any pattern in the data that appears consistently throughout the database. We investigate this question by evaluating the

<table>
<thead>
<tr>
<th>Periods Covered by the Four Time-Blocks</th>
<th>A</th>
<th>March 1892 to November 1957</th>
</tr>
</thead>
<tbody>
<tr>
<td>B</td>
<td>November 1957 to March 1973</td>
<td></td>
</tr>
<tr>
<td>C</td>
<td>March 1973 to August 1982</td>
<td></td>
</tr>
<tr>
<td>D</td>
<td>August 1982 to November 1999</td>
<td></td>
</tr>
</tbody>
</table>

Fig. 4. Day-of-week histograms for the four time blocks A–D. The horizontal line is the mean value.
chi-square statistic using the experimental data of Table 2. We evaluate $\chi^2$ using data from columns 7 and 8 in Equation 2.1: if the pattern is consistent, and the scatter in count rate is smaller than the deviation of the count rate from its expected value, then we will expect to obtain a large value of $\chi^2$ (for the given number of degrees of freedom). On the other hand, if the count rate varies substantially from block to block, so that the scatter is as large as or larger than the departure from the expected rate, then we will expect $\chi^2$ to be small.

We show data from columns 7 and 8 in Figure 5, which give no obvious evidence for a consistent departure from the uniform distribution. We find that $\chi^2 = 3.93$. For 6 degrees of freedom, there is a probability of 0.39 of obtaining this large a value or more by chance. Hence, as expected, this test confirms that there is no evidence for a consistent departure from uniformity with respect to day of the week.

We have carried out a similar analysis to study the variation of event rate with hour of the day. The corresponding histograms are shown in Figure 6. It is obvious that there is a huge variation, which is fairly consistent among time-blocks, of the count by time of day. The minimum is at about mid-day, and the maximum at about 10 pm. The variation is so large that a chi-square test is redundant. (The actual value of $\chi^2$ is 684.) The same data are shown as an error-bar display in Figure 7.

It is convenient to carry out an analysis of time-of-year variation similar to our analysis of time-of-day variation. We have therefore divided the year into 24 equal parts (taking time of day into account) of about 15.22 days each. We refer to these divisions as “hours of year” (HOY) and ordinary hours as “hours of
day” (HOD). The histograms are shown in Figure 8, and the error-bar display in Figure 9. This dataset also leads to a very large value of the chi-square statistic (122.7), confirming what is obvious from the plots that there is a significant and fairly consistent variation in the count rate with time of year. There is a minimum around May and a broad maximum around September.

![Fig. 6. HOD histograms for the four time blocks A–D. The horizontal line is the mean value.](image)

![Fig. 7. Mean and standard deviation of the number of events per hour per block. The horizontal line is the mean value.](image)
3. Power Spectrum Analysis

In any investigation of a time series, power spectrum analysis can be informative. Perhaps the simplest procedure for a large but irregular time series is to form the “Schuster periodogram” or “Rayleigh power” (Bretthorst, 1988; Mardia, 1972), which is given by

\[ S_R(v) = \frac{1}{N} \left| \sum_{n=1}^{N} e^{i2\pi vt_n} \right|^2, \]  

(3.1)

where \( t_n \) is the time of event \( n \), \( N \) is the total number of events, and \( v \) is the frequency. The power spectrum as computed from this formula is shown for frequencies in the range 0 to \( 5y^{-1} \) in Figure 10. We see that, apart from the unavoidable strong peak near zero frequency, there are two notable peaks, one at \( 0.42y^{-1} \) with \( S = 308 \), and the other close to \( 1y^{-1} \) with \( S = 297 \). The former has no obvious significance, but the latter is most likely related to the calendar year. It may have a purely calendar origin, due to the variation of the event rate during the year (as shown in Figure 8), but it could conceivably have an astronomical origin, since some fraction of UFO reports may be due to the misperception of celestial objects.

If the events were independent, \( S \) would have an exponential distribution such that the probability of obtaining a power \( S \) or more would be \( e^{-S} \). However, events appear to be highly correlated, occurring in bunches, so we may not interpret the power spectrum in this way. We find that in the frequency range 0 to \( 10y^{-1} \), there are 390 peaks. A semi-logarithmic display of the cumulative

Fig. 8. Histograms formed from HOY for the four time blocks A–D.
distribution of these peaks is shown in Figure 11. We see that, except for the 10 to 15 strongest peaks, the cumulative distribution is close to exponential. We find that if we “normalize” the Rayleigh power as follows,

\[ S = \frac{S_R}{25}, \]

Fig. 9. Mean and standard deviation of the number of events per HOY per block. The horizontal line is the mean value.

Fig. 10. Rayleigh power spectrum.
the cumulative distribution of the normalized power satisfies

$$P(S > S_0) = e^{-S_0}$$

for most of the range. Hence the power at 1y^{-1} corresponds to a “normalized” power $S = 12$. The probability of obtaining this power or more would be about $6 \times 10^{-6}$.

4. Running-Wave Power Spectrum Analysis

In the previous section, we looked into the possibility that the event rate contains one or more oscillations in time. In this section, we ask a different but related question. We look into the possibility that the event rate exhibits a significant pattern in terms of a rotating reference frame. We know that if the frame rotates with respect to the Earth with a one-day rotation period, there is a very strong modulation, shown in Figures 6 and 7. If it were to turn out that many events are due to observations of the planet Jupiter, then one would expect to find a significant pattern in terms of a frame that has the same mean rotation rate that Jupiter has with respect to the Earth.

It is convenient to take as our basic reference frame one in which the Sun has a fixed location. With respect to an observer on Earth, this frame rotates with a period of one day. With respect to an observer on a nearby star, this frame rotates with a period of one year. With respect to this basic frame, Jupiter rotates with a period of 398.88 days (Jupiter’s synodic period).

We denote by $\phi_n$ the angular position of the zenith (the position looking vertically upward) for event $n$, but normalize the angle to run from 0 to 1:

$$\psi = \phi / 2\pi.$$
This quantity is related the hour of day of the event by

$$\psi = (\text{HOD})/24.$$  \hfill (4.2)

We now denote by $\nu$ the angular velocity of the test frame with respect to the basic (Sun-locked) frame, where $\nu$ is measured in cycles per year. Then we can evaluate the possibility that events “cluster” with respect to the test frame by forming the following form of the Rayleigh power:

$$S(\nu) = \frac{1}{N} \left| \sum_{n=1}^{N} \exp[i2\pi(\psi_n + \nu t_n)] \right|^2,$$  \hfill (4.3)

where $t$ is to be measured in years.

We note that $\nu = 0$ corresponds to the Sun-locked frame and $\nu = 1y^{-1}$ corresponds to the star-locked frame, for which the combination $\psi + \nu t$ has a well known astronomical interpretation. If $t$ is measured from 0 hours on January 1 of some year, then $t$ (in $y^{-1}$) is related to the “hour of year” introduced in Section 2 by

$$t(\mod 1) = (\text{HOY})/24.$$  \hfill (4.4)

We see from Equations 4.2 and 4.4 that the combination $\psi + \nu t$ is related to the sum of HOD and HOY. This combination is related to LST by the equation

$$\text{LST} = \text{HOD} + \text{HOY} + 6.67.$$  \hfill (4.5)

We noted in Section 3 that there appears to be a modulation of the event rate with period 1 year. This annual variation may of course be due to a seasonal variation in hours of daylight, social patterns, atmospheric conditions, etc. The interplay of an annual modulation and the strong daily modulation shown in Figures 6 and 7 would lead to a combined effect that may be written schematically as a probability distribution function (pdf) given by

$$P(\text{HOD, HOY}) = P(\text{HOD})P(\text{HOY}),$$  \hfill (4.6)

where $P(\text{HOD})$ and $P(\text{HOY})$ are the pdf’s for events in terms of HOD and HOY, respectively. Since each pdf is periodic, with period 24, the simplest modulations are of the form

$$P(\text{HOD}) = 1 + \Delta_D \sin\left(2\pi \frac{\text{HOD}}{24} + \chi_D\right)$$  \hfill (4.7)

and

$$P(\text{HOY}) = 1 + \Delta_Y \sin\left(2\pi \frac{\text{HOY}}{24} + \chi_Y\right).$$  \hfill (4.8)
We see that Equation 4.6 now yields

\[
P(HOD, HOY) = 1 + \Delta_D \sin \left( \frac{2\pi}{24} HOD + \chi_D \right) + \Delta_Y \sin \left( \frac{2\pi}{24} HOY + \chi_Y \right)
\]

\[
+ \frac{1}{2} \Delta_D \Delta_Y \left[ \cos \left( \frac{2\pi}{24} HOD + \chi_D - \frac{2\pi}{24} HOY - \chi_Y \right) \right.
\]

\[
- \cos \left( \frac{2\pi}{24} HOD + \chi_D + \frac{2\pi}{24} HOY + \chi_Y \right) \right]
\] (4.9)

leading to a term that is a function of the LST and another term, of exactly the same amplitude, that is a function of the quantity

\[
ALST = HOD - HOY + 6.67, \quad (4.10)
\]

which we refer to as the “Alias Local Sidereal Time” (ALST). If the Rayleigh power defined by Equation 4.3 leads to a power spectrum with two equal peaks at \( \nu = 1y^{-1} \) and at \( \nu = -1y^{-1} \), we may infer that the annual modulation found in the analysis of Section 3 was due simply to the seasonal effect. But if we find that there is a strong peak at \( \nu = 1y^{-1} \), and no peak—or only a weak peak—at \( \nu = -1y^{-1} \), we may infer that the annual modulation has some kind of astronomical origin.

The power spectrum in terms of normalized power is shown for frequencies in the range \(-5 \) to \( 5y^{-1} \) in Figure 12, in which the powers for positive frequencies are shown as positive quantities, and the powers for negative frequencies are shown as negative quantities. Ignoring the inevitable large peak at zero
frequency, we see that there are significant peaks at or near $v = 1y^{-1}$ ($S = 125$) and $v = 2y^{-1}$ ($S = 77$). At exactly $v = 1y^{-1}$, $S = 102$, and at exactly $v = 2y^{-1}$, $S = 75$. By contrast, at exactly $v = -1y^{-1}$, $S = 3.5$, and at exactly $v = -2y^{-1}$, $S = 8.5$.

A semi-logarithmic display of the cumulative distribution of peaks for the frequency range $-10$ to $10y^{-1}$ is shown in Figure 13. We again find that, except for a few peaks, the cumulative distribution is close to exponential. If we “normalize” the Rayleigh power as follows,

$$S = S_R/7,$$

(4.11)

the cumulative distribution of the normalized power satisfies

$$P(S > S_0) = e^{-S_0}$$

(4.12)

for most of the range. Hence the peak near $v = 1y^{-1}$ for the forward wave corresponds to a normalized power of 17.9; the probability of obtaining this power or higher is about $1.7 \times 10^{-8}$. The peak near $v = 2y^{-1}$ for the forward wave corresponds to a normalized power of 11.0; the probability of obtaining this power or higher is about $1.7 \times 10^{-5}$.

Since the power spectrum computed from Equation 4.3 is not symmetric in positive and negative frequencies, it appears not to be due to the interplay of the HOD and HOY modulations. The asymmetry indicates that there is a modulation in terms of LST. It is not unreasonable that the modulation should show up at $v = 2y^{-1}$ as well as at $v = 1y^{-1}$, since Equations 4.7 and 4.8 represent only the simplest approximations to the pdfs, and more accurate approximations must be expected to contain terms of higher order.

Fig. 13. Cumulative distribution of powers computed from Equation 4.3.
5. Test of LST Peaks in Running-Wave Spectrum

We saw that the distribution of peak powers differs significantly from the form in Equation 4.12 to be expected of independent random variables. This is because, as mentioned earlier, consecutive counts of events are highly correlated, since events arrive in “bunches.” Hence we need a more reliable procedure for obtaining significance estimates.

A convenient method of obtaining robust significance estimates is the “shuffle test.” We retain the actual lists of times of events and of times of day of events, but we randomly reassign members of one list to members of the other list. For each such “shuffle,” we compute the power using Equation 4.3. We may then determine the fraction of shuffles for which the power exceeds any specified value.

We have carried out this procedure for both \( v = 1y^{-1} \) and \( v = 2y^{-1} \), for 10,000

<table>
<thead>
<tr>
<th>Day</th>
<th>A</th>
<th>B</th>
<th>C</th>
<th>D</th>
<th>Total</th>
<th>Mean</th>
<th>SD</th>
</tr>
</thead>
<tbody>
<tr>
<td>Sunday</td>
<td>459</td>
<td>399</td>
<td>441</td>
<td>464</td>
<td>1763</td>
<td>440.75</td>
<td>29.53</td>
</tr>
<tr>
<td>Monday</td>
<td>458</td>
<td>452</td>
<td>412</td>
<td>486</td>
<td>1808</td>
<td>452.00</td>
<td>30.51</td>
</tr>
<tr>
<td>Tuesday</td>
<td>400</td>
<td>459</td>
<td>420</td>
<td>384</td>
<td>1663</td>
<td>415.75</td>
<td>32.38</td>
</tr>
<tr>
<td>Wednesday</td>
<td>433</td>
<td>409</td>
<td>405</td>
<td>440</td>
<td>1687</td>
<td>421.75</td>
<td>17.35</td>
</tr>
<tr>
<td>Thursday</td>
<td>478</td>
<td>423</td>
<td>440</td>
<td>390</td>
<td>1731</td>
<td>432.75</td>
<td>36.62</td>
</tr>
<tr>
<td>Friday</td>
<td>447</td>
<td>446</td>
<td>476</td>
<td>449</td>
<td>1818</td>
<td>454.50</td>
<td>14.39</td>
</tr>
<tr>
<td>Saturday</td>
<td>350</td>
<td>437</td>
<td>431</td>
<td>412</td>
<td>1630</td>
<td>407.50</td>
<td>39.79</td>
</tr>
</tbody>
</table>

Fig. 14. Results of a test of 10,000 shuffles for \( v = 1y^{-1} \) showing (thick line) the logarithm of the fraction of trials that give a power exceeding a value shown in the abscissa. The thin line is the curve computed from Equation 5.2. The asterisk shows the actual power \((S = 102.3)\) for \( v = 1y^{-1} \).

TABLE 2

Histogram Data for Breakdown by Day of the Week
shuffles each. We show in Figure 14 the log of the fraction (ordinate) of simula-
tions for \( m = \frac{1}{\text{yr}} \) that have power in excess of the value shown in the abscissa. We
find that this curve is similar in shape to that of the “false alarm” probability dis-
cussed by Scargle (1982). For \( M \) independent trials, each corresponding to
a power spectrum analysis of independent normally distributed data, for which the
power spectrum is known to be exponential, Scargle shows that the probability
that the maximum of the \( M \) powers exceeds a given value \( S \) is given by

\[
P = 1 - (1 - e^{-S})^M. \tag{5.1}
\]

We find that, if we generalize this formula to

\[
P = 1 - (1 - e^{-aS})^M, \tag{5.2}
\]

we get a good fit to the results of the shuffle operation for the case \( v = 1\text{yr}^{-1} \) with
the values \( M = 10,000 \) and \( a = 0.23 \). This fit enables us to extend the results of
the shuffle test. From this extrapolation, we find that the shuffle test indicates
that the probability of obtaining a power of 102 or more is \( 10^{-6.19} \), i.e. \( 6.5 \times 10^{-7} \).

We have repeated these calculations for \( v = 2\text{yr}^{-1} \). We show in Figure 15 the
log of the fraction (ordinate) of simulations for \( v = 2\text{yr}^{-1} \) that have power in
excess of the value shown in the abscissa. We get a good fit to the results of the
shuffle operation for the case \( v = 2\text{yr}^{-1} \) with the values \( M = 13 \) and \( a = 0.425 \).
Using this fit to extrapolate the curve, we find that the shuffle test indicates that the probability of obtaining a power of 75.6 or more is $10^{-12.84}$, i.e. $1.45 \times 10^{-13}$.

### 6. LST and ALST Patterns

It is interesting to examine the distribution of events with respect to LST, which is related to HOD and HOY by Equation 4.5. The histogram is shown in Figure 16. It appears to be fairly uniform except in the range 18 to 24 hours, where there is a noticeable bump. The same data are shown as an error-bar display in Figure 17, which indicates that the structure seen in Figure 16 is indeed consistent through the four time blocks. We find that the chi-square statistic computed from data in Figure 17 has the value 82.3. For 23 degrees of freedom, the probability of obtaining this value by chance is only $1.4 \times 10^{-8}$. Hence the result of this test is significant, but not as significant as the result of the running-wave test of Section 5. This is understandable, since the chi-square test treats all "bins" as independent, whereas the Rayleigh-power approach takes account of the phase relationship between bins.

We pointed out in Section 4 that if the LST modulation were due to a convolution of the HOD and HOY modulations, the resulting modulation should be approximately symmetric with respect to LST and ALST defined by Equation 4.10. It is therefore interesting to compare with Figures 16 and 17 the corresponding figures for ALST. These are shown as Figures 18 and 19. We see that these are not at all like Figures 16 and 17. By contrast, Figures 18 and 19 appear to be featureless. Indeed, when we compute the chi-square statistic from Equation 2.1, using data appearing in Figure 19, we find that $\chi^2 = 8.59$. For 23
degrees of freedom, there is a probability of 0.997 of obtaining this value or more. Hence the ALST histogram is consistent with no modulation, but the degree to which this is true is somewhat surprising. This test indicates that the LST modulation is most probably not due to the convolution of the HOD and HOY modulations. This finding supports the results of Section 5.
7. Test Cases

In Section 4, we gave a theoretical argument as to why one would expect that the powers of the forward and reverse waves would be similar if the oscillation is due to an interplay of the time-of-day and the time-of-year distributions. Since this is a crucial argument in support of the case for an LST effect, we now reconsider this argument from a different point of view. We consider two hypothetical catalogs: in one there is an LST effect, and in the other there is not. We expect that the powers of the forward and reverse waves will be similar in the second case and dissimilar in the first case. We retain the actual dates of events, but we modify the time-of-day distributions.

Taking our lead from Figure 17, we consider a random distribution of LST values that is flat over 80% of the phase-of-year range, but is larger by 30% in the other 20% of the phase. We then compute the powers of the forward and reverse waves by the method of Section 4. The result is shown in Figure 20. We see that there is a strong peak for the forward wave at $m = \frac{1}{y}/c_0$, but there is no significant peak for the reverse wave. The forward wave also has a small peak at $m = \frac{2}{y}/c_0$.

For comparison, taking our lead from Figure 9, we consider a random distribution of HOY values that is flat over 70% of the phase-of-year range, but is larger by 20% in the other 30% of the phase. We then compute the powers of the forward and reverse waves by the method of Section 4. The result is shown in Figure 21. We see that the forward and reverse waves have peaks of comparable size at $v = 1y^{-1}$. There are also small peaks at $v = 2y^{-1}$.
This comparison confirms our expectation that the combined effects of time-of-day and time-of-year modulations leads to comparable powers in both the forward and reverse waves, whereas an LST modulation leads to significant power in only the forward wave.

Fig. 20. Running-wave power spectrum, forward waves shown with positive power, and reverse waves shown with the negative of the power, for a model with an LST modulation.

Fig. 21. Running-wave power spectrum, forward waves shown with positive power, and reverse waves shown with the negative of the power, for a model that has a significant HOY modulation but no LST modulation.
8. Discussion

It appears from our analysis that the rate of UFO events exhibits a significant variation in terms of LST. There may be some mundane explanation of this curious fact, but this explanation is not obvious. What seemed to be an obvious interpretation, that the modulation is due to the combined effect of the time-of-day and time-of-year modulations, seems to be ruled out by the running-wave power spectrum analysis since this effect would be symmetric with respect to the forward and reverse waves, which disagrees with our data analysis. The analysis of Section 5 seems to show that the peaks at $v = 1y^{-1}$ and at $v = 2y^{-1}$ in the running-wave analysis are highly significant. Nevertheless, the results of this test cannot be taken at face value, since there are strong patterns in both the HOD distribution and the HOY distribution. It will be important to examine the catalog for systematic effects that might explain the apparent LST effect. It will also be important to devise other significance tests that hopefully will be valid even though there are strong patterns in the HOD and HOY distributions. It is also clearly important to apply the same analysis to an independent catalog such as the UFOCAT catalog.

If the effect does indeed have an astronomical origin, it points to a source with right ascension in the range 21.5 $+$\/-1.5. Based on the analysis carried out to date, the declination is unknown.

The analysis reported in this article appears to support the early findings of Poher (1973) and of Poher and Vallee (1975), which were based on independent and comparatively small catalogs.

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