**Probabilistic Precipitation Anomalies Associated with ENSO**

Simon J. Mason* and Lisa Goddard+

1. Introduction

El Niño–Southern Oscillation (ENSO) warm events are associated with climate extremes over many areas of the globe (Bradley et al. 1987; Ropelewski and Halpert 1987, 1989, 1996; Kiladis and Diaz 1989; Halpert and Ropelewski 1992). Because of the relatively high predictability of the ENSO phenomenon (Cane and Zebiak 1985; Cane et al. 1986; Latif et al. 1998; Neelin et al. 1998; Stockdale et al. 1998a,b), prior knowledge of the expected state of the equatorial Pacific Ocean provides a significant source of predictability of seasonal climate variability over much of the Tropics, and farther afield (Palmer and Anderson 1994). Knowing what typically has occurred during previous ENSO events gives some indication of what is likely to happen during a present or upcoming event. For many parts of the world this knowledge provides a better estimate of the probable future climate than the assumption that seasonal conditions will be the same as average.

A composite of climate anomalies observed during past ENSO events provides some useful information about typical ENSO impacts (e.g., Bradley et al.
Given the limitations of the composite and linear measures of ENSO influences for forecasting future impacts, it is of value to estimate the probabilities of climate anomalies conditional upon the phase of ENSO (Ropelewski and Halpert 1996). A simple approach is to use contingency tables. This approach can provide a measure for identifying the influence of an independent variable on the probability of a predefined climate event occurring. The contingency table is used to determine whether the events are more or less likely to occur given different values of the independent variable. For example, warm and cold ENSO episodes have been associated with extremes in global precipitation eigenmodes (Lau and Sheu 1991), in regional precipitation (Shukla and Paolina 1983; Parthasarathy et al. 1988; Barnett et al. 1991; Lau and Sheu 1991; Hutchinson 1992; Webster et al. 1998), and in annual numbers of hurricane days in the North Atlantic (Gray and Sheaffer 1991). The contingency table method was used by Ropelewski and Halpert (1987, 1989) to identify changes in median precipitation conditional upon the phase of the Southern Oscillation.

In this paper, the influence of extreme ENSO states on global precipitation anomalies is investigated using contingency tables. This paper extends the analyses of Ropelewski and Halpert (1987, 1989) to consider three categories rather than the two categories used in their analyses of station data. Thus precipitation anomalies that are only weakly positive or negative are considered “near normal” and are not counted in the climate impacts. The results provide useful indications of the global impacts of El Niño and La Niña events in probabilistic terms and using newly available high-resolution datasets. By using gridded and spatially interpolated precipitation data that provide complete coverage of global land areas, a first attempt can be made to estimate the proportion of the earth’s land surface that has a significant ENSO-related precipitation signal.

2. Data and methods

a. Data

Gridded monthly precipitation data at a resolution of $0.5^\circ \times 0.5^\circ$ were obtained from New et al. (1999, 2000). The almost 12 000 stations represented in this gridded dataset were expanded from the original Climatic Research Unit–Department of Energy data described in Eischeid et al. (1991), and developed by Hulme (1994) and the Climate Research Unit at the
University of East Anglia. After extensive quality control the station data anomalies were averaged into 0.5° grids and then interpolated using thin-plate splines. Data are available from the beginning of the century, but only the 45-yr period 1951–52 to 1995–96 has been considered in this paper because of diminished quality of both the gridded precipitation data and the sea surface temperature data, described below, prior to about 1950. For each grid point seasonal precipitation totals were calculated and assigned to the below-normal, normal, or above-normal tercile. The analyses were repeated using the lower-resolution (2.5° × 3.75°) dataset of Hulme (1994) as a quality check on the results obtained by using the high-resolution data.

b. Identification of ENSO-related climate impacts

The historical impacts of El Niño and La Niña events on climate can be illustrated effectively by calculating the observed percentage of times that seasonal precipitation or temperature has been in the upper, middle, and lower climatological terciles during ENSO extremes (Mason et al. 1999). A 3-month mean Niño-3.4 index was calculated by averaging Kaplan et al.’s (1998) sea surface temperature anomalies over the central equatorial Pacific Ocean (5°N–5°S, 170°–120°W), and the warmest eight El Niño events between 1951–52 and 1995–96 were identified for each season. The selected El Niño years are indicated in Table 1a, which shows the warmest 11 events. By selecting the eight warmest events in each season the Niño-3.4 index exceeds one standard deviation in all cases except for June–August 1992. Similarly, the eight coldest La Niña events were selected and are listed in Table 1b. Again the selection of eight events ensures that in most cases the Niño-3.4 index exceeds one standard deviation. Although there are some differences in the sea surface temperature threshold used between seasons and between ENSO phases, the standardized thresholds are reasonably similar. In addition, because of the small sample sizes and highly discrete form of the distributions used in the significance tests, as discussed below, the constant number of selected cases facilitates comparison of results. However, to test the robustness of the results, all analyses were repeated using only the 5 strongest ENSO events and using the 11 strongest events. The strongest 5 events correspond with a minimum Niño-3.4 anomaly of about 0.6°C, while the strongest 11 events correspond with a minimum anomaly of about 0.35°C. Changing the number of selected ENSO events between 5, 8, and 11 effected only incidental differences in the results, and so results are presented only for the strongest eight events. Where notable differences do occur, they are mentioned.

The numbers of times that the observed precipitation anomalies during these warm- and cold-event years were in each tercile were tabulated for each grid point. These frequencies give some indication of the likelihood of observing a climate anomaly in each of the categories during El Niño and La Niña events, respectively. Areas experiencing a “dry season” (defined as a 3-month period in which, on average, the precipitation total is less than 15% of the annual total and less than 50 mm) and areas in which the seasonal precipitation total was zero in at least one-third of the years were masked. Antarctica has been excluded from the analysis because of a lack of observational data, although there probably is a detectable precipitation signal there (Bromwich et al. 2000).

c. Calculation of significance levels

The significance of the number of times that the observed precipitation or temperature anomalies during ENSO extreme events were above or below some predefined threshold can be calculated using contingency table theory. From a total of \( n \) years, of which \( b \) are “dry,” and from which \( r \) years are selected at random (the \( r \) strongest El Niño years, e.g.), let the number of dry years that are selected, denoted by \( X \), be equal to \( x \) [where \( 0 \leq x \leq \min(r,b) \)]. This information can be represented in a \( 2 \times 2 \) contingency table, as shown in Table 2, or extended to a three-category problem, as shown in Table 3. Assuming there are a total of \( r \) El Niño years and that \( x \) of them are dry, then the significance is defined as the probability of selecting \( x \) or more dry years in a random sample of \( r \) years. This probability is equivalent to the right tail area of the hypergeometric distribution (Agresti 1990, 1996) and is given by

\[
p_x(X \geq x) = H(x;r,b,n) = \sum_{k=x}^{\min(r,b)} \binom{b}{k} \binom{n-b}{r-k} \binom{n}{r}.
\]  

The test for significance defined by Eq. (1) is known as Fisher’s exact test (Fisher 1935, 1970; Irwin 1935) (appendix A). In our case, where \( n = 45, b = 15, \) and \( r = 8, \) five selected events is a significantly high number beyond the 90% level of confidence, six events is significant beyond the 95% level, and seven or more events is significant beyond the 99% level. Zero is a significantly low number of selected events beyond the
95% level of confidence. Further discussions on the significance levels used in this paper are presented in appendix B.

When calculating more than one significance test, such as for different seasons or spatial grid points, the calculated significance level normally should be adjusted to account for multiplicity, using the bonferroni inequality if the tests are independent (Katz 1988; Brown and Katz 1991; Katz and Brown 1991). If the tests are not independent, because of spatial correlation for example, it would be necessary to resort to Monte Carlo methods to estimate field significance (Livezey and Chen 1983). Field significance tests were not conducted in this paper because the aim is to identify where the ENSO-related climate signal is locally significant, not to demonstrate that an ENSO signal exists. On the strength of considerable earlier research, it was assumed a priori that the ENSO phenomenon does have an impact on climate around the globe. Multiplicity problems do arise, however, when estimating the spatial extent of the ENSO signature. These issues are addressed below and in appendix B.

d. Assumptions

The accuracy of the significance tests is dependent on the validity of assumptions very similar to those

<table>
<thead>
<tr>
<th>Year</th>
<th>Dec–Jan–Feb (σ = 0.87°C)</th>
<th>Mar–Apr–May (σ = 0.54°C)</th>
<th>Jun–Jul–Aug (σ = 0.57°C)</th>
<th>Sep–Oct–Nov (σ = 0.83°C)</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>SSTA</td>
<td></td>
<td>SSTA</td>
<td></td>
</tr>
<tr>
<td>1982–83</td>
<td>1.92</td>
<td>2.21</td>
<td>1992</td>
<td>1.20</td>
</tr>
<tr>
<td>1957–58</td>
<td>1.49</td>
<td>1.71</td>
<td>1983</td>
<td>1.19</td>
</tr>
<tr>
<td>1972–73</td>
<td>1.48</td>
<td>1.70</td>
<td>1987</td>
<td>0.93</td>
</tr>
<tr>
<td>1991–92</td>
<td>1.44</td>
<td>1.66</td>
<td>1958</td>
<td>0.78</td>
</tr>
<tr>
<td>1965–66</td>
<td>1.20</td>
<td>1.38</td>
<td>1969</td>
<td>0.77</td>
</tr>
<tr>
<td>1994–95</td>
<td>1.08</td>
<td>1.24</td>
<td>1993</td>
<td>0.69</td>
</tr>
<tr>
<td>1986–87</td>
<td>1.06</td>
<td>1.22</td>
<td>1966</td>
<td>0.63</td>
</tr>
<tr>
<td>1968–69</td>
<td>1.04</td>
<td>1.20</td>
<td>1991</td>
<td>0.58</td>
</tr>
<tr>
<td>1987–88</td>
<td>0.83</td>
<td>0.95</td>
<td>1995</td>
<td>0.54</td>
</tr>
<tr>
<td>1990–91</td>
<td>0.74</td>
<td>0.85</td>
<td>1994</td>
<td>0.47</td>
</tr>
<tr>
<td>1969–70</td>
<td>0.71</td>
<td>0.82</td>
<td>1972</td>
<td>0.44</td>
</tr>
<tr>
<td>Avg 5</td>
<td>1.51</td>
<td>1.73</td>
<td>Avg 5</td>
<td>0.97</td>
</tr>
<tr>
<td>Avg 8</td>
<td>1.34</td>
<td>1.54</td>
<td>Avg 8</td>
<td>0.85</td>
</tr>
<tr>
<td>Avg 11</td>
<td>1.18</td>
<td>1.36</td>
<td>Avg 11</td>
<td>0.75</td>
</tr>
</tbody>
</table>
that are required for a Monte Carlo test for significance, in which the observational data are repeatedly scrambled, or the selected years are repeatedly redefined (Livezey and Chen 1983). In both cases it is assumed that the observation in one year is independent of the observations in another year, which can be tested using autoregressive methods (Wilks 1995). In practice this assumption of independence will be invalid if there is a trend or significant low-frequency variability in the data. If the selected years are reasonably well distributed throughout the period of interest, the effects of autocorrelation and trend are probably not too serious, and the calculated significance levels will err on the conservative side. If the selected years are clustered, the effects will be more severe, and significance

| Year       | SSTA | $|\sigma|_{rel}$ | Year       | SSTA | $|\sigma|_{rel}$ | Year       | SSTA | $|\sigma|_{rel}$ | Year       | SSTA | $|\sigma|_{rel}$ |
|------------|------|-------------|------------|------|-------------|------------|------|-------------|------------|------|-------------|
| 1973–74    | -1.66| 1.91        | 1974       | -0.84| 1.56        | 1975       | -1.07| 1.88        | 1955       | -1.71| 2.06        |
| 1970–71    | -1.51| 1.74        | 1971       | -0.83| 1.54        | 1988       | -1.02| 1.79        | 1975       | -1.35| 1.63        |
| 1975–76    | -1.44| 1.66        | 1955       | -0.83| 1.54        | 1973       | -0.81| 1.42        | 1973       | -1.29| 1.55        |
| 1988–89    | -1.35| 1.55        | 1989       | -0.73| 1.35        | 1955       | -0.75| 1.32        | 1988       | -1.21| 1.46        |
| 1955–56    | -1.02| 1.17        | 1975       | -0.59| 1.09        | 1956       | -0.69| 1.21        | 1970       | -1.11| 1.34        |
| 1954–55    | -0.99| 1.14        | 1956       | -0.54| 1.00        | 1964       | -0.68| 1.19        | 1964       | -1.01| 1.22        |
| 1984–85    | -0.71| 0.82        | 1976       | -0.50| 0.93        | 1954       | -0.67| 1.18        | 1954       | -0.85| 1.02        |
| 1964–65    | -0.71| 0.82        | 1964       | -0.46| 0.85        | 1970       | -0.57| 1.00        | 1971       | -0.77| 0.93        |
| 1962–63    | -0.62| 0.71        | 1985       | -0.43| 0.80        | 1971       | -0.60| 1.05        | 1974       | -0.69| 0.83        |
| 1971–72    | -0.51| 0.59        | 1968       | -0.36| 0.67        | 1974       | -0.45| 0.79        | 1956       | -0.69| 0.83        |
| 1967–68    | -0.48| 0.55        | 1967       | -0.35| 0.65        | 1959       | -0.35| 0.61        | 1962       | -0.47| 0.57        |
| Avg 5      | -1.40| 1.60        | Avg 5      | -0.76| 1.41        | Avg 5      | -0.87| 1.52        | Avg 5      | -1.33| 1.61        |
| Avg 8      | -1.17| 1.35        | Avg 8      | -0.67| 1.23        | Avg 8      | -0.78| 1.37        | Avg 8      | -1.16| 1.40        |
| Avg 11     | -1.00| 1.15        | Avg 11     | -0.59| 1.09        | Avg 11     | -0.70| 1.22        | Avg 11     | -1.01| 1.22        |

Table 2. Two-by-two contingency table for verification of a binary forecast system.

<table>
<thead>
<tr>
<th>Observations</th>
<th>Yes</th>
<th>No</th>
<th>Total</th>
</tr>
</thead>
<tbody>
<tr>
<td>Dry</td>
<td>$x$</td>
<td>$b-x$</td>
<td>$b$</td>
</tr>
<tr>
<td>Not dry</td>
<td>$r-x$</td>
<td>$n-r-b+x$</td>
<td>$n-b$</td>
</tr>
<tr>
<td>Total</td>
<td>$r$</td>
<td>$n-r$</td>
<td>$n$</td>
</tr>
</tbody>
</table>
is likely to be overestimated. By purposely clustering the selected years, tests for climate change can be conducted. For example, significantly high frequencies of events in the last few years of a series may provide evidence for trends (Wigley and Jones 1987).

e. Estimating the area of ENSO signals

Percentages of total land area not experiencing a dry season or dry conditions (as defined in section 2b) and that experienced significantly high frequencies of precipitation in either the above- or below-normal tercile category were calculated. Grid areas were weighted by the cosine of the latitude. In calculating these areas an 85% confidence level has been used to identify grid points with significantly high frequencies of above- or below-normal precipitation. The estimated areas with ENSO impacts should therefore be considered upper limits, and may be overestimated by 10% or more, as discussed in appendix B.

All area estimates were compared against the coarser-resolution Hulme (1994) dataset. These results are in close agreement with those from the New et al. (1999, 2000) dataset, but suggest that areas with observed above- or below-normal precipitation in more than half of the selected El Niño or La Niña events may be underestimated by at most 5%. However, apart from the smoothing of detected signals over larger grid areas at a coarser resolution, the Hulme (1994) dataset also provides a disproportionate weight to the Pacific islands and coastal areas, since it is assumed that any grid containing land is entirely land. It is therefore likely that the Hulme (1994) dataset may lead to slight overestimates of the area of ENSO signals, and so the area estimates from the New et al. (1999, 2000) dataset are considered more realistic.

Since the areas affected by ENSO change from season to season, a more meaningful indication of the global extent of ENSO teleconnections may be obtained by calculating total areas with ENSO-related seasonal climate anomalies at any time of year. Area estimates were made for all twelve 3-month seasons and all areas with a significant signal in at least one of the seasons were identified. Because of problems of multiplicity when performing tests for more than one season and for the thousands of grid points (Katz 1988; Brown and Katz 1991; Katz and Brown 1991; Wilks 1995), only areas with above- or below-normal rainfall in at least 75% of the eight El Niño or La Niña events were counted. The test

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**Fig. 1.** Frequencies of (top) above- and (bottom) below-normal Dec–Feb precipitation during the eight strongest El Niño events from 1951–52 to 1995–96, defined using the Niño-3.4 index (1957–58, 1965–66, 1968–69, 1972–73, 1986–87, 1982–83, 1991–92, 1994–95). Frequencies are shown only where they are significantly high or low at a 90% level of confidence.
therefore considers only areas that have had a significantly high frequency of either above- or below-normal precipitation at the 95% level of confidence (see appendix B).

### 3. Precipitation anomalies associated with El Niño

Areas where there has been a significantly high or low number of years with above- and below-normal seasonal precipitation during the eight strongest El Niño episodes since 1951–52 are shown in Figs. 1–4. Results for the near-normal category are not shown because of an absence of spatially coherent signals, although there are few areas where the probability of near-normal precipitation increases significantly during ENSO events. Areas of significant decreases in the probability of near-normal precipitation are almost invariably reflected by significant increases in one of the other two categories. The main features are described in the following sections, and further details and references are reserved for appendix C.

#### a. December–February

There are surprisingly few areas where the frequency of above-normal December–February precipitation has been significantly high during El Niño events (Fig. 1a). Of all nondry land areas excluding Antarctica, only about 8% experienced a significantly high frequency (in more than half the cases) of above-normal precipitation, which is only slightly more than would be expected by chance (6.8%). Despite the weak global impact, there are some regions where El Niño has been associated with significantly high frequencies of above-average precipitation. The signal appears to have been most reliable

![El Niño, March-May, above-normal](image)

![El Niño, March-May, below-normal](image)

**Fig. 2.** Frequencies of (top) above- and (bottom) below-normal Mar–May precipitation during the eight strongest El Niño events from 1951 to 1995, defined using the Niño-3.4 index (1958, 1966, 1969, 1983, 1987, 1991, 1992, 1993). Frequencies are shown only where they are significantly high or low at a 90% level of confidence.

<table>
<thead>
<tr>
<th>Table 3. Three-by-two contingency table for verification of a binary forecast system.</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>El Niño years</strong></td>
</tr>
<tr>
<td><strong>Dry</strong></td>
</tr>
<tr>
<td><strong>Not dry, not wet</strong></td>
</tr>
<tr>
<td><strong>Wet</strong></td>
</tr>
<tr>
<td><strong>Total</strong></td>
</tr>
</tbody>
</table>

**Table 3.**
El Niño is associated with areas with high probabilities of below-normal December–February precipitation (Fig. 1b) rather than above-normal precipitation. Significantly high frequencies of below-normal precipitation occurred over about 11% of non-dry areas, which is almost twice the area expected by chance. Over 3% of land areas experienced below-normal precipitation in more than 75% of the cases, which is over three times the area expected by chance (1.1%). Below-normal precipitation occurred during El Niño most frequently over central Indonesia, the southern Philippines, and parts of Micronesia and the Marshall Islands, but was also widespread over northern South America, southern Africa, and parts of northern Africa. Below-normal precipitation over land during El Niño conditions in December–February was therefore much more widespread than above-normal precipitation. In total, about 19% of nondry land areas have experienced significant precipitation impacts during December–February.

b. March–May

Although probabilities of anomalous March–May precipitation conditions during El Niño events (Fig. 2) weakened in most areas compared to December–February (Fig. 1), with the notable exception of the Philippines, total land areas experiencing anomalous precipitation changed only minimally. There was a small expansion of the area experiencing below-normal precipitation to more than 12%, again almost twice the area expected by chance. Below-normal precipitation at this time of year was widespread throughout southeastern Asia, northern Australia, parts of Africa and the Middle East, and along the north coast of Brazil. The expansion of the area experiencing below-normal precipitation was more than offset by a fairly substantial contraction of areas of above-normal precipitation.
The signal for above-normal precipitation was reasonably strong over parts of central Russia and the former Soviet states, but less than 6% of land areas experienced above-normal precipitation in more than half of the cases, and less than 1% experienced above-normal precipitation in 75% of the cases or more. In both instances these areas are less than would be expected by chance. The March–May El Niño–related signal therefore is predominantly one of high probabilities of below-normal precipitation over global land areas (Fig. 1b) (Bradley et al. 1987). Areas that have experienced below-normal precipitation were more than twice as widespread as areas experiencing above-normal precipitation. Considering both above- and below-normal precipitation, about 18% of nondry land areas are significantly affected by El Niño in March–May.

Despite a substantial weakening in the average Niño-3.4 anomaly (Table 1a) between December–February (1.34°C) and March–May (0.85°C) there is no observed decrease in land areas experiencing significantly high frequencies of either above- or below-normal precipitation. The difference between the average standardized Niño-3.4 anomalies of December–February (1.55) and March–May (1.57) is negligible, and so the atmosphere must remain highly sensitive to the weakening positive sea surface temperature after the mature phase of an El Niño event, which most of the selected events represent (Table 1a).

c. June–August

Areas of significantly high frequencies of above-normal precipitation during El Niño events in boreal summer (Jun–Aug) (Fig. 3) again have been constrained to only a few areas. Only about 5% of land areas experienced high frequencies of above-normal precipitation in more than half of the cases. Most of these areas are concentrated in the Americas. Again, below-normal precipitation has been widespread, with over 14% of nondry land areas experiencing below-normal conditions in more than half of the cases, which is more than twice the area expected by chance. Similarly, just over 4% of areas experienced below-normal precipitation in at least 75% of the cases, four times the area expected by chance. Areas with strong below-normal signals include eastern Australia, Indonesia, and parts of eastern and southern Asia. Thus, again, approximately 19% of nondry land areas observed significantly high probabilities of either above- or below-
normal precipitation during El Niño events in June–August.

For this season, most of the selected warm events represent the growth phase of an El Niño (Table 1a). The average Niño-3.4 anomaly (0.86°C) is about the same as for March–May (0.85°C), and the standardized anomalies are very similar also (1.50 compared with 1.57). Given that the occurrence of below-normal precipitation during these early stages of developing El Niño events is widespread, the importance of early warnings of developing warm events is highlighted.

d. September–November

During September–November (Fig. 4), the widespread high frequencies of below-normal precipitation persisted, and strengthened in areas such as eastern Australia and Indonesia. The total nondry land areas that experienced below-normal precipitation in more than half of the cases was over 13%, which is little changed from the previous season. Areas where below-normal precipitation occurred in all eight El Niño years reached their maximum extent (just under 0.1%) in this season, but are largely confined to the western Pacific. Similarly, areas where above-normal precipitation occurred in all eight years reached their maximum extent (0.06%) in September–November and are focused in the central Pacific. Over continental areas, high frequencies of widespread above-normal precipitation were confined largely to the Middle East, parts of east Africa, and the United States. Almost 9% of land areas experienced above-normal precipitation during more than half of the El Niño episodes.

In September–November the extent of the El Niño-related precipitation signal appears to reach a maximum, even though the El Niño itself typically reaches maturity only in the following season. The average Niño-3.4 index was 1.23°C, which is a 10th of a degree (celsius) cooler than the December–February average (Table 1a), but the averaged standardized index, at 1.49, is similar to that for the other three seasons. The area where below-normal precipitation occurred in more than half of the cases was slightly less than during June–August, but areas where below-normal precipitation was experienced in over 85% of the El Niño cases were most widespread in September–November. Similarly, the area where the frequency of above-normal precipitation was significantly high reached a maximum in this season (about 8% or 9%). This amounts to a total of about 22% of land areas during El Niño associated with above- or below-normal precipitation in September–November.
4. Precipitation anomalies associated with La Niña

Some symmetry exists between the precipitation anomalies associated with La Niña and El Niño episodes (cf. Figs. 1–4 with 5–8). In this section, only the important exceptions to the symmetry are highlighted.

a. December–February

In the United States, La Niña appears to have had a weaker influence on precipitation than El Niño in the southwest of the country, but a stronger influence in the southeast (Fig. 5b). Similar differences between the strength of the El Niño and La Niña signals are found for southeastern South America, parts of southern Africa, and for southeastern China. As opposed to El Niño, which was associated mainly with below-normal precipitation over land areas, the areas with high frequencies of below-normal precipitation (less than 8%) during La Niña events were less than the areas of above-normal precipitation (about 12%). Thus approximately 20% of nondry land areas exhibited precipitation impacts during La Niña events in December–February.

b. March–May

During March–May, there is again reasonable symmetry between the El Niño (Fig. 2) and La Niña signals (Fig. 6). Noteworthy exceptions are Australia, where precipitation impacts were more robust during La Niña, and northeastern Brazil and southeastern Asia, where the La Niña signal was weaker. Overall, the impacts of La Niña were experienced over approximately the same area (19%) as those of El Niño (18%), but the preference for above-normal precipitation characteristic of La Niña in the December–February season is no longer evident. The areas with high frequencies of above- and below-normal precipitation during La Niña events were about 10% in each case. The total area of significant anomalous precipitation in either direction was only marginally less than in the other seasons, even though the absolute sea surface temperature anomalies were considerably weaker than during September–February (Table 1b). If only the strongest five cold events are considered, there are few areas outside of Australia that indicate a widespread impact of La Niña on precipitation during March–May.

c. June–August

The preference for above-normal precipitation to occur during cold extremes of the ENSO phenomenon during the eight strongest La Niña events from 1951 to 1995, defined using the Niño-3.4 index (1950, 1955, 1956, 1971, 1974, 1975, 1976, 1989). Frequencies are shown only where they are significantly high or low at a 90% level of confidence.
in December–February is reestablished during July–August. Above-normal precipitation occurred much more frequently than below normal throughout the continental Tropics, and occurred over more than 12% of all nondry land areas, compared to about 8% for below normal.

d. September–November

Again there is reasonable symmetry between the El Niño (Fig. 4) and La Niña signals (Fig. 8) during September–November, although there are some notable exceptions. The most marked asymmetry is probably that the strength of the below-normal signal over southeastern South America was much stronger during La Niña than the opposite signal was during El Niño. In northeastern Argentina 100% of the selected La Niña events were associated with below-normal precipitation. The impact of El Niño events over this region apparently is delayed until the December–February season when a high frequency of above-normal seasonal precipitation has been observed (Fig. 1a). In total about 12% of nondry areas experienced high frequencies of above-normal precipitation.

High frequencies of below-normal precipitation are evident over western Saudi Arabia, the Kyrgyzstan area, and the central near-equatorial Pacific islands so that 0.28% of all nondry land areas had below-normal precipitation in all the selected La Niña events. The area that experienced equally high frequencies of above-normal rainfall also reached a maximum extent (0.32%) at this time of year. These areas were concentrated mainly in the Indonesia–northern Australia region. As was shown for El Niño, therefore, the global impact of La Niña on seasonal precipitation appears to reach a maximum in September–November. The area with high frequencies of below-normal precipitation during La Niña events reached a maximum (almost 13%) in this season.

The total area experiencing high frequencies of above- or below-normal precipitation in September–November is about 25% of nondry land areas. As for El Niño, this area is at a maximum during September–November, even though ENSO events typically reach maturity only in the following season. However, the average Niño-3.4 index during the eight strongest La Niña events was −1.16°C, which is indistinguishable from the December–February average of −1.17°C (Table 1b), and the averaged standardized index, at −1.40, is slightly stronger than in the other seasons.
5. Discussion and conclusions

Previous analyses of precipitation anomalies associated with El Niño and La Niña events have indicated the global nature of ENSO-related teleconnections (Bradley et al. 1987; Ropelewski and Halpert 1987, 1989, 1996; Kiladis and Díaz 1989). In this paper a newly available high-resolution dataset (New et al. 1999, 2000) was used to extend the results of earlier analyses. Areas with significantly high frequencies of either below- or above-normal precipitation during ENSO events were identified, and were indicated in Figs. 1–8. Only data after 1951 were considered to ensure that data quality is maximized, and only the strongest ENSO events were investigated. An attempt has been made to obtain estimates of the percentage of global land areas with a strong ENSO-related precipitation signal.

The results confirm that climate anomalies associated with El Niño and La Niña events do occur on a global basis, but suggest that caution should be exercised to not exaggerate the local significance or the global coverage of the ENSO influence. During the eight strongest El Niño and eight strongest La Niña episodes over the period 1951–52 to 1995–96, precipitation anomalies beyond either the above- or below-normal terciles occurred in more than half of the selected ENSO events over no more than about 15%–25% of land areas in any 3-month season. The spatial coverage decreases to about 5% when considering precipitation anomalies that occurred in at least two-thirds of the cases, and to an average of only about 0.1% for anomalies in all eight cases. The limited extent of these areas highlights the event-to-event variability in observed precipitation anomalies during ENSO episodes, and indicates a definite need to emphasize the probabilistic nature of the climate response rather than focusing on a composite response to past ENSO events (Ropelewski and Halpert 1996).

The areas affected by ENSO change from season to season. The total area experiencing ENSO-related climate anomalies in at least one of the twelve 3-month seasons gives a more meaningful indication of the global extent of ENSO teleconnections, but the exact significance levels of this are difficult to determine because of multiplicity problems. However, areas with high frequencies of anomalous seasonal precipitation at some time of year that are locally significant beyond the 95% level of confidence constitute only about 20–30% of total land areas. Results using the 11 strongest El Niño or La Niña events suggest that seasonal pre-

![Fig. 8. Frequencies of (top) above- and (bottom) below-normal Sep–Nov precipitation during the eight strongest El Niño events from 1951 to 1995, defined using the Niño-3.4 index (1950, 1954, 1955, 1964, 1970, 1973, 1975, 1988). Frequencies are shown only where they are significantly high or low at a 90% level of confidence.](image-url)
cipitation over as much as 30%—45% of land areas may be affected at some time in the year. It should be emphasized that the difference in these estimates is because of a sensitivity to unavoidable changes in the significance level used, and not because of a sensitivity to the number of selected ENSO events per se. Estimates of the spatial extent of ENSO impacts are therefore sensitive to the strength of the signal that is considered, but what can be concluded is that areas with strong ENSO-related seasonal precipitation signals are restricted in extent, but areas with a weak signal are much more widespread (see appendix B for a more detailed discussion).

The danger of overemphasizing the climate response to strong ENSO events is particularly apparent in the case of El Niño, which is usually the ENSO extreme that receives greatest attention. The results presented in this paper and supported elsewhere (e.g., Hoerling et al. 1997; Vega et al. 1998) suggest that the impact of La Niña is at least as widespread as that of El Niño. It is estimated that La Niña affects precipitation over about 5%—15% more of the land surface than El Niño. Given that the magnitude of the SST anomalies defining the La Niña events considered here was weaker than those of El Niño, the extent of the La Niña–related climate impacts is probably a result of the more localized anomalous tropical convection relative to that during El Niño events.

It is difficult to identify a clear seasonal cycle in the proportion of land areas that have experienced a strong ENSO-related precipitation signal. The magnitude of SST anomalies associated with ENSO events does differ between seasons, but the differences in the standardized anomalies are small. This suggests that the relative magnitude of the anomalous SST is what matters for the seasonal climate. There are, however, some indications that the spatial extent of the precipitation signal is greatest in the September–November season. For El Niño, the precipitation signal is predominantly below-normal precipitation over land areas year-round (Bradley et al. 1987). For La Niña, some seasonality exists in the precipitation signal with more regions experiencing above-normal precipitation during June–August and December–February. It should be emphasized that these results apply only to land areas, and do not imply a global increase or decrease in precipitation in either ENSO extreme. The impact of ENSO events on precipitation over much of the central Pacific is sampled by only a few islands and so the spatial extent of precipitation anomalies in this region are vastly underrepresented.

Where ENSO events do have a significant impact on precipitation anomalies, it seems reasonable to use historical relative frequencies, such as those shown in Figs. 1–8, to derive probabilistic estimates of a precipitation anomaly for a season in which an ENSO event is anticipated. During the 1997–98 El Niño and the 1998–99 La Niña, for example, the composite frequencies of ENSO impacts were often used to suggest the likely outcomes for the following months, and were an important input to the International Research Institute for Climate Prediction “net assessments” (Mason et al. 1999). However, obtaining a forecast from contingency tables of historical impacts is complicated. If the probabilities obtained using a contingency table approach are to be used in a forecasting environment, the table should be constructed so that the probabilities are contingent upon forecasts of ENSO events rather than on observed events. Given that there is still much uncertainty involved in forecasting ENSO events operationally (Barnston et al. 1999; Landsea and Knaff 2000), probability estimates obtained from Figs. 1–8 are likely to be overconfident. There is additional uncertainty arising from the modulating influence of sea surface temperatures in other ocean areas. Providing reliable estimates of forecast probabilities therefore has to consider uncertainties in the evolution of global sea temperatures, as well as errors in estimating the historical ENSO impacts.

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Appendix A: Calculation of the hypergeometric equation

With only a slight loss of accuracy because of floating point errors, hypergeometric probabilities can be calculated using

\[ h(x; r, b, n) = \exp \left[ \sum_{i=r}^{b} \ln i + \sum_{i=n-r+1}^{n-b} \ln i - \sum_{i=r}^{n} \ln i - \sum_{i=r}^{b} \ln i \right] \]

where \( b \) is the number of cases of \( n \) with the characteristics of interest and \( x \) is the number of \( r \) selected
cases with the characteristic. The right-tail area [Eq. (1)] can then be calculated using

\[ p_x (X \geq x) = H(x; r, b, n) = \sum_{k=x}^{\min(r, b)} h(k; r, b, n). \] (A2)

Further rounding errors may be minimized by reversing the summation in Eq. (A2) so that \( k = \min(r, b) \) is calculated first, and \( k = x \) is calculated last. For greater accuracy, the problem can be reformulated using prime number factorizations to minimize the number of computations (Wu 1993), but for most climatological applications, Eqs. (A1) and (A2) should be sufficiently accurate.

### Appendix B: Confidence levels

Because of the discrete form of the hypergeometric distribution, the values of \( p_x \) become highly discrete with small samples. Significance tests for these distributions therefore can become conservative. For example, given the parameters used in this paper for the hypergeometric distribution \((n = 45, b = 15, \text{ and } r = 8)\), at least five below-normal cases have to be selected for below-normal precipitation to be significantly frequent at the 90% level of confidence. However, there is only a 6.8% probability of selecting five or more cases (in a one-tailed test). Similarly, at least six cases have to be selected at the 95% level of confidence, but there is only a 1.1% probability of selecting six or more cases. The probability of selecting seven or more cases is less than 0.1%. Significantly low frequencies are indicated in addition, but mainly for ease of comparing the El Niño and La Niña signals. There is a 2.7% probability of selecting no cases.

Since there is interest in high frequencies of both above- and below-normal precipitation, a two-tailed test is more appropriate. A test conducted with a 90% level of confidence then becomes highly conservative. At least six or more below-normal or above-normal cases would have to be selected to be significant at a 90% level of confidence, even though there is only a 2.2% probability of selecting six or more of one of the two cases. The next step is to adopt an approximately 85% level of confidence and to count areas with at least five below-normal or five above-normal cases; there is a 13.6% probability of selecting five or more of one of the two cases. In Figs. 1–8, areas with above- or below-normal precipitation in more than half of the eight strongest El Niño or La Niña years are indicated, and thus an 85% level of confidence has been used. Since an 85% confidence level is liberal, the percentage areas with significant ENSO impacts in any of the four seasons should be considered upper limits. An approximately 95% level of confidence can be used if the 11 strongest El Niño or La Niña events are considered. The areas with significant ENSO impacts are then estimated to be about 11% or 12% less than the quoted areas.

These significance tests make no correction for problems of multiplicity, which should be considered acute given the thousands of grid points over which the test is conducted. The problem of multiplicity becomes even more serious when estimating the total area with an ENSO-related precipitation signal at some time of the year. For this reason, the 85% confidence level used to mask areas in Figs. 1–8 was considered too liberal for identifying areas with ENSO signals in each of the 12 seasons. As discussed above, the next step is to count only those areas with six or more of the eight selected events having below-normal or above-normal precipitation, which is equivalent to having a confidence level of about 98%. This level of confidence has been used. The results are compared with those obtained by counting those areas with 7 or more of the 11 selected events having above- or below-normal precipitation. There is a 4.0% probability of selecting seven or more cases. The confidence level is then reduced from about 98% to about 96%, and indicates an increase in the area affected by about 10%–15%. Conversely, the estimated area decreases dramatically if a slightly stricter confidence level is selected. The implication is that areas with strong ENSO-related seasonal precipitation signals are restricted in extent, but areas with a weak signal are widespread.

### Appendix C: Areas with significant ENSO-related precipitation anomalies

Brief descriptions are provided of the extent of ENSO-related precipitation anomalies in each of the four seasons considered. The interested reader is referred to the cited references for more details.

#### a. December–February

Noteworthy areas where there has been a significantly high frequency of above-normal seasonal precipitation for December–February during El Niño
events and that do not typically experience a dry season at this time of year include the following: Kiribati and the northern parts of French Polynesia and the Cook Islands (He and Barnston 1996); the southwestern and southeastern United States (Livezey et al. 1997; Montroy 1997; Mo and Higgins 1998; Montroy et al. 1998; Masutani and Leetmaa 1999; Minnich et al. 2000); southern Brazil and neighboring parts of Paraguay and Argentina (Wagner 1987; Grimm et al. 1998, 2000; Lucero 1998; Montecinos et al. 2000); southwestern and southeastern United States (Gong and Wang 1999; Wang et al. 2000); and parts of central Russia. Over Europe and the Middle East changes in storm tracks are thought to effect small changes in the probability of above-normal winter precipitation in some areas (Fraedrich 1994; Rodo et al. 1997; Price et al. 1998; Kadioglu et al. 1999). Below-normal precipitation was prevalent over much of northern South America; central Indonesia, the southern Philippines, parts of Australia, Micronesia, and the Marshall Islands, extending southeastward to the southern Cook Islands (Fig. 1a) (Quinn et al. 1978; Allan 1991; Alory and Delcroix 1999); and parts of southern Africa (Mason and Jury 1997).

There are some notable asymmetries in the precipitation signals between warm and cold ENSO phases. In the United States, below-normal precipitation was more frequent in the southeast during La Niña than was above-normal precipitation during El Niño (Livezey et al. 1997). Other areas in which December–February precipitation appears to have been more strongly affected during La Niña episodes than during El Niño episodes include central eastern Brazil, southern Africa, eastern Australia, and parts of Europe (Fraedrich 1994; Davies et al. 1997). Over southern Africa, the asymmetry was notably strong when only the strongest five events were selected: above-normal precipitation occurred during all five La Niña events over large areas of South Africa and Namibia, whereas, a signal in the five strongest El Niño events was barely detectable. In some places, such as the southwestern United States (Livezey et al. 1997; Gershunov 1998) and southeastern South America, the signal during El Niño events was stronger.

b. March–May

The strong signals of above-normal December–February precipitation over the United States–Mexico border region and over southern Brazil during El Niño events weakened considerably during March–May (Fig. 2a). Similarly, in northern South America, high frequencies of below-normal precipitation north of the equator weakened considerably (Xavier et al. 1995) but redeveloped over northeastern Brazil (Fig. 2b), although the influence of the tropical Atlantic or rainfall here may be more important (Ward and Folland 1991; Moron et al. 1995; Enfield 1996; Kane 1997, 1999b; Uvo et al. 1998; Fontaine et al. 1999; Nobre and Cavalcanti 1999). An equally strong change occurred over southeastern Asia, the Philippines, and islands farther east. During December–February, a strong signal of below-normal precipitation was evident over the Philippines, Palau, and Micronesia (Fig. 1b), but appeared to shift westward in March–May while weakening over Micronesia and developing over southeastern Asia. Signals over Australia strengthened a little, but remained weak.

In other areas, the signal was little changed in March–May compared to December–February. Over southern Africa, high frequencies of below-normal precipitation persisted, but with some weakening over Zambia and Zimbabwe (Fig. 2b), suggesting that an early end to the southern African rainfall seasons is likely during El Niño events (Waylen and Henworth 1996; Mason and Tyson 1999). Elsewhere in Africa, drier than normal conditions persisted in the Mediterranean states, but a below-normal similar signal is evident for La Niña (Fig. 6b) and so may be spurious. In northeastern Europe areas of significantly high frequencies of above-normal precipitation remained patchy (Fraedrich 1994; Rodo et al. 1999; van Oldenborgh et al. 2000). The stronger signal in the Kazakhstan area shifted southward reaching into northern India.

Changes in the areas with significant La Niña–related climate anomalies between December–February and March–May are broadly consistent with those of El Niño. Only the most noteworthy features are highlighted here. The most notable change is the development of high probabilities of above-normal precipitation over a large part of northern and eastern-central Australia (Fig. 6a), which developed more strongly than the signal for below-normal conditions during El Niño. In contrast, above-normal precipitation over southeastern Asia during the selected La Niña cases was not as frequent as below-normal during the eight strongest El Niño years. Apart from the persistence of strong signals over many of the tropical Pacific islands, signals elsewhere were relatively weak.

c. June–August

In the central equatorial Pacific over Nauru and Kiribati, above-normal precipitation conditions oc-
curred frequently during El Niño events, but were not experienced at all over parts of Indonesia and Papua New Guinea, the Solomon Islands, Vanuatu, New Caledonia, Fiji, and much of eastern Australia. During September–November, the strength of both the El Niño– and La Niña–related precipitation anomalies appear to reach a maximum, even though the associated sea surface temperature anomalies typically continue to develop well into the December–February season.

Over continental areas, the Middle East had high frequencies of above-normal precipitation during the selected El Niño events, indicating a high probability of a good start to the rainfall season (Nazemosadat and Gordery 2000). The signal seems to be displaced during the La Niña events, where high frequencies of below-normal precipitation occurred in the vicinity of the Red Sea, and also over Kyrgyzstan and surrounding areas.

Farther southeast, there is evidence that the late rains of the Indian monsoon were frequently poor (good) (Fig. 4b) during El Niño (La Niña). In Africa, the Short Rains of Kenya and neighboring areas were frequently above normal during El Niño events (Farmer 1988; Beltrando and Camberlin 1993; Mutai et al. 1998), but the signal changes sign over western and central Ethiopia and the southern half of Sudan, where the September–November season can contribute late-season rainfall. The opposite signals are evident for La Niña (Fig. 8). Over South America El Niño–related signals for September–November precipitation were not strong, but below-normal precipitation occurred very frequently over northeastern Argentina and neighboring areas during La Niña (Fig. 8b). In North America the El Niño impact seems to have been stronger than that of La Niña: over much of the central and western United States, including Alaska (Livezey et al. 1997), above-normal rainfall was fairly widespread, and below-normal rainfall was not observed in many of these areas (Fig. 4b).


d. September–November

During September–November, the highest frequencies of above-normal precipitation during El Niño events occurred over the near-equatorial Pacific islands east of about 165°E (Fig. 4a). Similarly high probabilities of below-normal precipitation occurred during La Niña events in the same areas (Fig. 8a). The ENSO signal in precipitation was not much weaker, but of the opposite sign, over parts of Indonesia, Papua

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