Some •aspects of systematic error in the ECMWF model

T. Jung,* A. M. Tompkins and M. J. Rodwell ECMWF, Shinfield Park, Reading, RG2 9AX, UK

*Correspondence to:

Abstract

T. Jung, ECMWF, Reading, Berkshire, UK.. E-mail: jung@ecmwf.int

AQ1

Received: 8 March 2005 Revised: 30 March 2005 Accepted: 30 March 2005 First, systematic errors of short-range and medium-range Z500 forecasts are described along with their changes since the early 1980s. Then systematic cloud error will be described. Finally, the capability of the ECMWF model to simulate the Madden-and-Julian Oscillation is assessed. Copyright © 2005 Royal Meteorological Society

1 I. Introduction

2 3 Two sources of error lead to the development of 4 forecast error: error in the initial conditions and model 5 error. Continuous monitoring at ECMWF reveals that 6 forecast errors have been substantially reduced in 7 recent years (Simmons and Hollingsworth, 2002). This 8 reduction of forecast errors is partly due to improved 9 initial data and partly due to model improvements. In 10 general, however, it is not straightforward to separate 11 the influence of improved analyses from those due to 12 improved model formulations, since models are used 13 in data assimilation schemes to determine the analysis.

14 A relatively simple way to identify aspects of model 15 error is to focus on *systematic* errors of the forecast. To this end, a particular meteorological aspect (e.g. 16 17 the mean circulation) is quantified from a large set 18 of forecasts. The model results are then compared 19 with estimates of the truth, which are obtained from 20 observational data (or reanalyses). At the beginning 21 of 2003, it was decided to carry out a comprehensive 22 study of systematic errors in the ECMWF forecasting 23 system. This decision was motivated by the fact that 24 such a systematic major documentation had not been 25 carried out for some time and that the ECMWF model 26 underwent considerable improvements in recent years 27 (e.g. Andersson *et al.*, 2003). In the following, we shall 28 discuss some of the outcomes of this extensive study 29 (see also, Jung and Tompkins, 2003; Jung et al., 2004; 30 Jung, 2005).

31 32

34

33 2. Results

2.1. Atmospheric circulation 35

36 We start by considering short-range and medium-range 37 systematic Z500 errors of the ECMWF model cycle 38 23R4. This cycle is one of the key model releases 39 at ECMWF used to carry out the ERA-40 reanalysis 40 (Uppala, 2002). It also forms the atmospheric com-41 ponent of the ECMWF operational seasonal forecast 42

system. This model cycle was in operational use at 43 ECMWF for the medium-range from 12 June 2001 44 to 21 January 2002. In the framework of the ERA-45 40 reanalysis project, this cycle has also been used 46 to carry out 10-day reforecast every day from 1960 47 to 2001. The resolution used is $T_L 159 (\approx 1.125^\circ)$ with 48 60 levels in the vertical. The length of the time series 49 50 allows us to quantify systematic errors with unprece-51 dented accuracy.

52 Mean systematic Z500 errors of short-range D + 2forecasts (D + n denotes a n -day forecast) are shown 53 54 in Figure 1 for all four seasons. The first thing to 55 notice is that systematic Z500 errors are very simi-56 lar throughout the annual cycle, both in terms of their spatial structure and their magnitude. The two areas 57 58 that stand out, in particular, are the North Pacific and 59 the central North American continent. In the North 60 Pacific, an anticyclonic bias has developed by D + 2, 61 which leads to an underestimation of the midlatitude 62 westerly winds. Over the North American continent, 63 the model has problems at D + 2 in producing the 64 observed stationary wave structure downstream of the 65 Rocky Mountains. Evidently, this problem is promi-66 nent in all four seasons. The relaxation of the 'convec-67 tive mass-flux limiter' for long time steps introduced in 68 October 2003 led to a significant reduction (the error 69 has been halved) of the North American Z500 bias 70 during the summer months (not shown). 71

Systematic Z500 errors at D + 10 are shown in 72 Figure 2. Evidently, the largest systematic errors in the 73 northern hemisphere occur during the winter season 74 (DJF•). Moreover, as for D + 2 forecasts, systematic 75 Z500 errors at D + 10 show a very similar structure 76 throughout the annual cycle. The spatial correlation 77 (north of 20°N) between the winter pattern and those 78 in spring, summer and autumn amounts to 0.57, 79 0.60 and 0.78 respectively. Notice that most of the 80 systematic errors found at D + 2 also show up at 81 D + 10 (e.g. North Pacific and North America). It is 82 worth pointing out that the thorough investigation of 83 the systematic error structure of one particular model 84

AQ2



Figure 1. Mean Z500 difference (shading in dam) between D + 2 forecast and verifying analysis data for (a) winter, (b) spring, (c) summer, and (d) autumn. Results are based on ERA-40 hindcast and reanalysis data from the period 1960-2001. Also shown is the mean Z500 field from ERA-40 reanalysis data (thin dotted contours in dam)

1 cycle in the far-medium range has been made possible 2 only through the availability of the ERA-40 hindcasts 3 over a long period (40 years in this study). Usually, 4 the operational ECMWF model undergoes changes 5 at least once a year. Therefore, the assessment of 6 systematic errors in medium range usually has to rely 7 on only one realization of each season. As discussed 8 in more detail by Jung (2005), this is problematic 9 since, on average, the skill of Z500 forecasts at D + 10is relatively low. As a consequence of this loss of predictability, the seasonal mean of all individual 12 forecasts is very similar to the climatology and, 13 therefore, the seasonal-mean forecast error resembles 14 the observed Z500 anomaly, except with opposite sign. 15 This makes it difficult to separate true systematic 16 model errors from the usually quite large 'apparent' 17 systematic error.

18 Climatological systematic Z500 errors of model 19 cycle 23R4 are described in detail by Brankovic and 20 Molteni (2004). The spatial structure of climatological 21 systematic Z500 error in the North Pacific is very 22 similar to that at D + 10 for the ERA-40 reforecasts; 23 the magnitude at D + 10, however, amounts only to 24 about half that in the extended-range. This shows that 25 systematic Z500 errors continue to grow beyond the

medium range. Further experimentation has revealed -26 that the North Pacific Z500 bias in the extended 27 range is largely due to the use of an unrealistic 28 aerosol climatology in north Africa and the Middle 29 East (Rodwell and Jung, manuscript in preparation). 30 This shows that the North Pacific Z500 bias is in 31 32 part remotely forced. The fact that systematic Z500 33 errors are also evident in the short range (D + 2), 34 Figure 1) clearly shows that the origin is both remote 35 and local, their relative importance being dependent 36 on the forecast range under consideration.

37 So far, the focus has been on systematic Z500 error 38 of one particular model cycle (23R4). Next, we discuss 39 how systematic Z500 error has changed in operational 40 ECMWF forecasts since the early 1980s. For the win-41 ter season, results have been recently presented by 42 Jung (2005). Here, we go one step further by consider-43 ing all four seasons. In the following, the magnitude of 44 the mean error component is quantified by computing 45 the spatial standard deviation of the difference between 46 mean Z500 forecast errors (forecast minus analysis) 47 north of 30°N for individual years. The resulting time 48 series for operational D + 2 and D + 5 forecasts are 49 shown in Figure 3. The most prominent feature at 50 D + 2 is the pronounced reduction of systematic Z500

2

10 11



Figure 2. Same as in Figure 2(a)-(d), except for D + 10 hindcasts. Note the different contour interval

error around the mid to late 1980, which have been 1 2 traced back to changes in the parametrization of con-3 vection and radiation and to a lesser degree gravity 4 wave drag and vertical resolution (Arpe, 1989). Evi-5 dently, D + 2 forecasts during the winter season benefited the most from these model improvement. After 6 7 almost 10 years of little changes of systematic Z500 error at D + 2, mean errors improved substantially in 8 9 all four seasons since 1999 or so. While the exact 10 reason for this reduction is not known, it is likely that improved parametrizations (see also next sec-11 12 tion) and an increase of the horizontal resolution to 13 T_L511 ($\approx 0.35^\circ$), which took place in autumn 1999, 14 played key roles.

15

Color Figure - Online only

¹⁶ 2.2. Clouds and cloud-related parameters

In recent years, effort has also been expended to 18 improve the representation of moist physical processes 19 20 in the ECMWF model. However, in general, it is not 21 straightforward to relate improvements of the repre-22 sentation of physical processes (so-called parametriza-23 tions) to fields such as geopotential height. This (and 24 their paramount influence on local weather conditions) 25 implies that diagnostics of cloud-related parameters 26 should be preferably included in any detailed study 27 of systematic model error.

Previous assessment of clouds in model cycles used 28 in the late 1990s has revealed that, in general, clouds 29 are well captured, with the following exceptions: The 30 cloud cover is too low in the midlatitudes (in particular 31 too little cloud cover is simulated over Europe in 32 summer) and subtropics; the cloud ice amount is too 33 low, especially in the midlatitudes; the liquid water 34 is too high, especially in the subtropics; the cloud 35 cover in stratocumulus regions is too low; and, finally, 36 there is too much high cloud in regions of tropical 37 deep convection (Jakob, 1999; Hogan et al., 2001; 38 Chevallier et al., 2001; Chevallier and Kelly, 2002). 39

Here, we can only briefly examine one cloud influ-40 enced diagnostic for illustrative purposes, namely, the 41 systematic error in the top-of-atmosphere (TOA) net 42 short-wave budget. For further details, see Jung and 43 Tompkins (2003) and Tompkins et al. (2004). Model 44 cycle 26R1 (operational from 29 April to 6 Octo-45 ber 2003) and cycle 23R4 are validated against Earth 46 Radiation Budget Experiment (ERBE) measurements 47 in Figure 4 to see if model improvements occurred in 48 49 the intervening 2-year period.

The older cycle (left column) reveals signs of some 50 of the characteristic errors identified in the literature. 51 The reflectivity is too high in much of the subtropics 52 and in the tropical Pacific and Atlantic due to excessive liquid water in these regions. In contrast, the lack 54

AQ4



Figure 3. Temporal evolution of wintertime Northern Hemisphere (north of 30° N) mean Z500 errors of operational (a) D + 2 and (b) D + 5 forecasts in winter (solid), spring (dotted), summer (dashed), and autumn (dash-dotted). A three year running average has been used for smoothing. Results are based on the spatial standard deviation of the temporal mean forecast error. Area-weighting has been taken into account

29

30 31

1

of stratocumulus near the West coast of the Americas and Africa is associated with too little reflectivity. Analysis of cloud cover and liquid water path (LWP) retrievals from other instruments confirms this assessment (not shown). In model cycle 26R1 (Figure 4, right column) observed TOA-SW• characteristics are substantially improved in the tropical and subtropical oceans. A summary of the model revisions to convective, radiative and cloud processes that lead to a reduction in LWP from cycle 23R4 to 26R1 is given in Jung and Tompkins (2003). In contrast, the model still fails to capture stratocumulus adequately. This has been addressed by a new diffusion scheme (not shown), which will be implemented operationally in late 2005 (Tompkins *et al.*, 2004).

2.3. Madden-and-Julian oscillation

So far, we have focussed on systematic errors of the mean. However, model problems may also affect the model's ability to simulate variations around the mean. Here, we shall concentrate on the by far most dominant mode of atmospheric intraseasonal variability in the Tropics, which is associated with continentalscale organization of convection propagating eastward across the Indian and western Pacific ocean. Honouring the discoverers of this phenomenon (Madden 28 and Julian, 1972), this mode is nowadays known as 61 the Madden-and-Julian oscillation (MJO). Regarding 62 operational activities at ECMWF, there are at least 63 three reasons why the MJO should be simulated well. 64 First, there is evidence that westerly wind bursts 65 66 can trigger ENSO• events. Therefore, the skill of 67 ECMWF's seasonal ENSO forecasts may crucially 68 depend on the model's ability to simulate the MJO. 69 Second, there is an indication that medium-range 70 forecast skill in the northern hemisphere extratrop-71 ics depends on how well the Tropics in general, and 72 the MJO, in particular, are simulated (Ferranti et al., 73 1990). Finally, the quasi-periodicity of the MJO at 74 periods of 30-60 days implies extended-range pre-75 dictability that might be utilized in monthly fore-76 casts (Vitart, 2004), which have been produced oper-77 ationally at ECMWF every week since October 2004.

78 The MJO has been diagnosed in a set of 6-month 79 long integrations with model cycle 26R1 (at T_L95 80 with 60 levels in the vertical). The integrations were 81 started on 1 October of each of the years 1960-2001 82 using observed SST fields. A dramatic shortcoming 83 of the ECMWF model is that it does not produce the 84 observed spectral peak in the 30-60 day range. This 85 can be inferred from Figure 5, which shows average 86 power spectra of tropical velocity potential anoma-87 lies at 200 hPa for different longitudes. The ERA-88 40 reanalysis data show a clear spectral maximum 89 in the eastern hemisphere, particularly between 60°E 90 and 180°E. The ECMWF model, on the other hand, 91 merely produces red power spectra with no indica-92 tion of quasi-periodicity. As pointed out by Jung and 93 Tompkins (2003), the model also has problems in sim-94 ulating the temporal coherence of slowly eastward 95 propagating anomalies, whereas relatively fast prop-96 agating anomalies are more realistically simulated. 97

Finally, it has been found that in the ECMWF model 98 MJO-related upper tropospheric divergence anoma-99 lies are primarily associated with large-scale precip-100 itation (i.e. convection on the gridscale) instead of 101 subgrid-scale convective precipitation (Tompkins and 102 Jung, 2004). Additional sensitivity experiments with 103 an aqua-planet version of the model have revealed that 104 the quasi-periodicity of the simulated MJO depends 105 on the ratio of large-scale to convective precipitation; 106 changes to the model physics that increase the propor-107 tion of large-scale precipitation amplify the magnitude 108 and increase the periodicity of the simulated MJO. 109 It is hypothesized that this is because the large-scale 110 cloud scheme is constrained to provide latent heat-111 ing in phase with any wave that provides forcing for 112 the cloud, a positive feedback along the lines of the-113 ories of Conditional Instability of the Second Kind 114 (CISK) (e.g. Lindzen, 1974; Kirtman and Vernekar, 115 1993). This is consistent with the propagating mode's 116 resemblance to a moist Kelvin wave. The convection 117 scheme is not so constrained and can provide heating 118 out of phase with the wave, possibly even damping the 119 oscillation (Emanuel, 1994). We should emphasize that 120



59 60



Atmos Squbeta 63000 (2005)

Figure 4. Mean top of the atmosphere downward short-wave radiation (Wm^{-2}) for the period June–August 1987 based on model data (upper panels), observational data from ERBE (middle panels) along with difference: model minus observational data (lower panels). Left: Model cycle 23r4. Right: Model cycle 26r1 and the frame of the servational data (lower panels). Left: Model cycle 23r4. Right: Model cycle 26r1 and the frame of the servational data (lower panels). Left: Model cycle 23r4. Right: Model cycle 26r1 and the frame of the servational data (lower panels). Left: Model cycle 23r4. Right: Model cycle 26r1 and the servational data (lower panels). Left: Model cycle 23r4. Right: Model cycle 26r1 and the servational data (lower panels). Left: Model cycle 23r4. Right: Model cycle 26r1 and the servational data (lower panels). Left: Model cycle 23r4. Right: Model cycle 26r1 and the servational data (lower panels). Left: Model cycle 23r4. Right: Model cycle 26r1 and the servational data (lower panels). Left: Model cycle 23r4. Right: Model cycle 26r1 and the servational data (lower panels). Left: Model cycle 23r4. Right: Model cycle 26r1 and the servational data (lower panels). Left: Model cycle 23r4. Right: Model cycle 26r1 and the servational data (lower panels). Left: Model cycle 23r4. Right: Model cycle 26r1 and the servational data (lower panels). Left: Model cycle 23r4. Right: Model cycle 26r1 and the servational data (lower panels). Left: Model cycle 23r4. Right: Model cycle 26r1 and the servational data (lower panels). Left: Model cycle 23r4. Right: Model cycle 26r1 and the servational data (lower panels). Left: Model cycle 23r4. Right: Model cycle 26r1 and the servational data (lower panels) and the servational data (lower panels). Left: Model cycle 23r4. Right: Model cycle 26r1 and the servational data (lower panels) and the servational data (lower panels). Left: Model cycle 23r4. Right: Model cycle 26r4 and the servational data (lower panels). Left: Model cycle 23r4. Right: Model cycle 26r4 and the serva

Copyright © 2005 Royal Meteorological Society



Figure 5. Average power spectra of tropical $(5^{\circ}S-5^{\circ}N)$ velocity potential anomalies at 200 hPa for ERA-40 reanalysis data (upper panels) and climate runs with model cycle 26R1 (lower panels). The average is based on 40 raw spectra for the autumn and winter months from October to March of the years from 1962 to 2001. Zonal averages are shown in the right-hand panels. The mean annual cycle has been removed before the computation of the spectra

we do not claim that this mechanism is necessarily rel evant to the real atmosphere, merely that it is key to
 the model's simulated propagating mode.

4 Most of the model's deficits in simulating the 5 MJO described above are typical features of earlier 6 model cycles and other atmospheric models as well 7 (Slingo et al., 1996). Given the importance of the 8 MJO for medium-range and extended-range forecast-9 ing improving the model's capability to properly sim-10 ulate the MJO has a high priority in the near future. 11 Possible model improvements, however, are likely to

require a better understanding of the mechanisms governing the MJO — a widely accepted theory for the MJO is still missing.

> 15 16

3. Further remarks

17 18

In this study, some aspects of systematic error in the 19 ECMWF model have been described. The concept of 20 systematic error is a very powerful and straightforward 21 tool to identify the existence of model errors. In 22 general, however, without any further experimentation

and diagnosis, it is difficult to infer the exact source of

7

60

61

AQ6

the model problems, giving rise to systematic model error. In this sense, diagnosing systematic model must be seen as only the first (but important) step in a chain to pinpointing and eradicating model error. From a methodological point of view, the concept of systematic model error is straightforward. There are potential pitfalls, however. First, the datasets used for verification can be associated with considerable uncertainties. This is particularly true for cloud-related parameters and precipitation over the oceans, for

12 13 example, which are notoriously difficult to observe. 14 Moreover, the estimation of systematic errors is a 15 statistical problem. Therefore, the shortness of time series poses serious problems, at least for some 16 17 forecast aspects. In this study, we have made use of 18 40 years of 10-day reforecasts with the same model 19 cycle to infer systematic Z500 errors in the northern 20 hemisphere with, to our knowledge, unprecedented 21 accuracy (see also Jung, 2005).

22 In the past, long time series for the purpose of model 23 assessment have primarily been obtained by carrying 24 out seasonal integrations for a relatively large number 25 of years (e.g. Brankovic and Molteni, 2004). While 26 this is definitely an important part of every model 27 assessment, it is difficult to separate locally from 28 remotely forced errors (Klinker and Sardeshmukh, 29 1992). To circumvent this problem, in our opinion, 30 it is very fruitful to augment climate diagnostics by 31 detailed investigations of systematic errors in the short 32 range and medium range, as has been done for Z500 33 in this study. 34

36 37 **References**

38

35

1

2

3

4

5

6

7

8

9

10

11

- Andersson E, Beljaars A, Bidlot J, Miller M, Simmons A, Thepaut JN. 2003. A Major New Cycle of the IFS: Cycle 25R4. ECMWF
- A1 *Newsletter 97.* ECMWF: Shinfield Park, Reading, Berkshire, UK. Arpe K. 1989. Changes in the ECMWF analysis-forecasting scheme
- 42 and the systematic error of the model. Technical Report 161.
 43 ECMWF: Shinfield Park, Reading, Berkshire, UK.
- Brankovic C, Molteni F. 2004. Seasonal climate and variability in the
 ECMWF ERA-40 model. *Climate Dynamics* 22: 139–155.
- Chevallier F, Kelly G. 2002. Model clouds as seen from space: comparison with geostationary imagery in the 11-micron window
- 47 channel. *Monthly Weather Review* **130**: 712–722.

- Chevallier F, Bauer P, Kelly G, Jakob C, McNally T. 2001. Model 48 clouds over oceans as seen from space: comparison with HIRS/2 49 and MSU radiances. *Journal of Climate* 14: 4216–4229. 50
- •Emanuel KA. 1994. Atmospheric Convection. Oxford University 51 Press: 580.
- Ferranti L, Palmer TN, Molteni F, Klinker E. 1990. Tropicalextratropical interaction associated with the 30–60 day oscillation and its impact on medium and extended range prediction. *Journal* of Atmospheric Science **47**: 2177–2199.
- Hogan RJ, Jakob C, Illingworth AJ. 2001. Comparison of ECMWF 56 winter-season cloud fraction with radar-derived values. *Journal of* 56 *Applied Meteorology* **40**: 513–525. 57
- Jakob C. 1999. Cloud cover in the ECMWF reanalysis. *Journal of* 58 *Climate* 12: 947–959. 59
- Jung T. 2005. Systematic errors of the atmospheric circulation in the ECMWF forecasting system. *Quarterly Journal of the Royal Meteorological Society* In presse.
- Jung T, Tompkins AM. 2003. Systematic errors in the ECMWF 62 forecasting system. Technical Report 422. ECMWF: Shinfield Park, 63 Reading, Berkshire, UK. 64
- Jung T, Rodwell MJ, Tompkins AM. 2004. Systematic Errors in the ECMWF Forecasting System. ECMWF Newsletter 100. ECMWF: Shinfield Park, Reading, Berkshire, UK. 66
- Kirtman B, Vernekar A. 1993. On wave CISK and the evaporation 67 wind feedback for the Madden-Julian oscillation. *Journal of* 68 *Atmospheric Sciences* 50: 2811–2814. 69
- Klinker E, Sardeshmukh PD. 1992. The diagnosis of mechanical dissipation in the atmosphere from large-scale balance requirements. *Journal of Atmospheric Sciences* **49**: 608–627. 71
- Lindzen RS. 1974. Wave-CISK in the tropics. *Journal of Atmospheric* 72 *Sciences* **31**: 156–179. 73
- Madden RA, Julian PR. 1972. Description of global-scale circulation cells in the tropics with a 40–50 day period. *Journal of Atmospheric Sciences* **29**: 1109–1123.
- Simmons A, Hollingsworth A. 2002. Some aspects of the improvement of skill of numerical weather prediction. *Quarterly Journal of Royal Meteorological Society* 128: 647–677.
- Slingo JM, Sperber KR, Boyle JS, Ceron JP, Dix M, Dugas B, Formattion Structure
 Slingo JM, Sperber KR, Boyle JS, Ceron JP, Dix M, Dugas B, Formattion Structure
 Slingo JM, Sperber KR, Boyle JS, Ceron JP, Dix M, Dugas B, Formattion Structure
 Slingo JM, Sperber KR, Boyle JS, Ceron JP, Dix M, Dugas B, Formattion Structure
 Slingo JM, Sperber KR, Boyle JS, Ceron JP, Dix M, Dugas B, Formattion Structure
 Slingo JM, Sperber KR, Boyle JS, Ceron JP, Dix M, Dugas B, Formattion Structure
 Slingo JM, Sperber KR, Boyle JS, Ceron JP, Dix M, Dugas B, Formattion Structure
 Slingo JM, Sperber KR, Boyle JS, Ceron JP, Dix M, Dugas B, Formattion Structure
 Slingo JM, Sperber KR, Boyle JS, Ceron JP, Dix M, Diagonal Structure
 Slingo JM, Sperber KR, Boyle JS, Ceron JP, Dix M, Diagonal Structure
 Slingo JM, Sperber KR, Boyle JS, Ceron JP, Dix M, Diagonal Structure
 Slingo JM, Sperber KR, Boyle JK, Shang M, Shang
- Tompkins AM, Jung T. 2004. Influence of process interactions on MJO-like convective structures in the IFS model. *ECMWF/CLIVAR* 85
 Workshop on Simulation and Prediction of Intra-seasonal Variability 86
 with Emphasis on the MJO. ECMWF: Shinfield Park, Reading, UK; 103–114.
- Tompkins AM, Bechtold P, Beljaars A, Benedetti A, Cheinet S, Janisková M, Köhler M, Lopez P, Morcrette J-J. 2004. Moist physical processes in the IFS: Progress and Plans. ECMWF Technical 90
 Memorandum, 452, available at http://www.ecmwf.int/publications/. 91
- Uppala S. 2002. ECMWF reanalysis 1957–2001. 3. Workshop on Reanalysis. ECMWF: Shinfield Park, Reading, UK; 1–10.
- *analysis.* ECMWF: Shinfield Park, Reading, UK; 1–10. 93 Vitart F. 2004. Monthly forecasting at ECMWF. *Monthly Weather Review* **132**: 2761–2779. 94

IMPOR at the r	TANT NOTE: Please mark your corrections and answers to these queries directly onto the proof elevant place. Do NOT mark your corrections on this query sheet.
Queries AQ1 I AQ2 I AQ3 I AQ4 I AQ5 I AQ6 I	a from the Copyeditor: Please provide the keywords for this article. Please confirm if this abbreviation needs to be spelt out. If yes, please provide the expansion. Please confirm if this abbreviation needs to be spelt out. If yes, please provide the expansion. Please confirm if this abbreviation needs to be spelt out. If yes, please provide the expansion. Please provide the place of publication. Please clarify if the article has since been published. If so please provide the complete details for this
1	reference.
	OF