



1	A new tool for model assessment in the frequency domain -					
2	Spectral Taylor Diagram : application to a global ocean general					
3	circulation model with tides					
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13	Abstract					
14	We introduce a new tool - the Spectral Taylor Diagram (STD) - for the comparison of time					
15	series in the frequency domain. The STD provides a novel way of displaying the					
16	squared-coherence, power, amplitude, phase, and root-mean-squared difference of discrete					
17	frequencies of two time-series. Each STD summarises these quantities in a single plot, for					
18	multiple targeted frequencies. The versatility of STDs is demonstrated through a series of					
19	sea-level comparisons between observations from tide gauges, and model results from a					
20	global eddy-permitting ocean general circulation model with explicit tidal forcing.					
21	Keywords: Model evaluation; Spectral Taylor Diagrams; Ocean tides; global ocean modelling					
22	Highlights:					
23	A new tool to evaluate tides is introduced: Spectral Taylor Diagram					
24	Ocean General Circulation model with explicit tidal forcing.					





25 1. Introduction

26 Ocean and climate modelling are widely used for research, forecasting, and climate 27 projections. An important step in the application of an ocean or climate model is model 28 assessment. This is commonly done by comparing model results to observations and 29 reanalysis (Landerer and Glecker, 2014, Chiyuan Miao et al 2014), or to other models (Flato 30 et al., 2013). Such comparisons often involve single variable to multi-variables, 31 multi-processes, and multi-phenomena employing methods of mathematical statistics for 32 quantitative evaluations as well as parametric or nonparametric tests of significance. 33 Traditional statistical comparisons in the time-domain typically include calculation of 34 correlations, and comparison of standard deviations, means, and trends (e.g., Schiller and 35 Brassington, 2011). Second order comparisons often compare some variant of empirical 36 orthogonal functions (e.g., Erofeeva et al., 2003). Comparisons in the frequency domain 37 typically involve considerations of the cross-spectral density, squared-coherence, phase, and 38 amplitude at a range of frequencies. In cases where a specific frequency band is targeted -39 to isolate a particular process, for example - analysis techniques such as wavelets, or 40 complex demodulation are commonly employed (Flinchem and Jay, 2000). Each type of 41 comparison has strengths and weaknesses. A challenge for any model assessment is the 42 concise depiction of multiple statistical metrics for easy interpretation. Taylor Diagrams (TDs; 43 Taylor, 2001) are now commonly used to concisely present multiple statistical properties from 44 the comparison in the time-domain. In this paper, we introduce a tool that is analogous to 45 TDs but for spectral comparisons - the Spectral Taylor Diagram (STD). 46 We demonstrate the versatility of STDs through a series of assessments of a global,

47 eddy-permitting ocean general circulation model with explicit tidal forcing. To date, most 48 global ocean models and climate models do not include explicit tidal forcing - with a few 49 notable exceptions (Schiller 2004; Schiller and Fiedler 2007; Arbic et al., 2012; Müller et al., 50 2012; and Ngdock et al. 2016) - largely because of the understanding that tidal energy is 51 completely dissipated in shallow waters (Wunsch 2000). However, many observations of 52 tides indicate that tides might be more important than previously thought. For example, Lee 53 et al. (2006) shows that barotropic tidal energy in coastal regions is several orders of 54 magnitude greater than the deep ocean. Moreover, Munk and Wunsch (1998) concluded that 55 mixing driven primarily by dissipation of tidal energy could contribute to one half of the power 56 required to return the deep waters to the surface.

57 The accuracy of ocean models, forced with tides, remains limited by uncertainties in a range 58 of model parameters (Schiller 2004; Schiller and Fiedler 2007; and Ngodock et al. 2016), 59 such as inaccurate bathymetry, bottom friction, model resolution, inaccurate estimation of





internal tides, and misrepresentation of the self-attraction and loading term (SAL). Another
 important source of barotropic errors in ocean modelling is introduced by an inaccurate
 estimation of phase information. In this study, we compare results from two model
 configurations with tides, to observations. We use version 5 of the Modular Ocean Model
 (MOM5) with it's default tidal configuration (DFT) and with the addition of phase information
 (T8).

⁶⁶ This paper is organised as follows. The STDs are described in section 2, the model is ⁶⁷ described in section 3, and applications of STDs are presented in section 4. We conclude ⁶⁸ and summarise our findings in section 5.

69 2. Spectral Taylor Diagrams – STDs

70 Taylor Diagrams (TDs; Taylor 2001) are often used to intercompare results from different 71 models with observations (e.g., Oke et al. 2012). TDs (e.g., Figure 1) represent unbiased 72 Root-Mean-Squared Error (RMSE; i.e., RMSE with the mean removed), the cross-correlation 73 between observed and modelled estimates, and the standard deviation of the analysed 74 time-series. Some presentations of TDs also include a metric of skill-score (e.g., Divikan et 75 al. 2012). These diagrams nicely summarise a number of statistical comparisons in a single 76 plot. TDs exploit the relationship between three statistical quantities that compose the law of 77 cosines. The correlation coefficient (R), the standard deviations of the test (σ_f) and 78 reference (σ_r) fields, and the centered root-mean-square difference (E') between these two 79 fields create a two-dimensional diagram through the following formula: 80 $E'^2 = \sigma^2 + \sigma^2 = 2\sigma \sigma R$ (4)

	$E = O_f + O_r - 2O_f O_r K$	(1)
81	which resembles the law of cosines.	

82 $c^2 = a^2 + b^2 - 2 a b \cos \varphi.$ (2)

⁸³ This geometric relationship is represented graphically in Figure 2.

In this paper, we introduce a new tool for inter-comparing different time-series against observations, based on a variant of TDs that we call STDs. Instead of calculating the correlation of an entire time-series with the observation, we select a frequency (or a band of frequencies) to be assessed. Here, a transformation is used to convert the time domain signal to the frequency domain. Furthermore, the choice of the three statistical quantities has to satisfy the law of cosines, such that the correlation is replaced by spectral coherence, and the standard deviation of the power replaces the time series standard deviation.





The squared-coherence is analogous to time domain measure of correlation, and is employed here since it also measures the strength of the linear relationship between two time series – ranging from 0 to 1 (the first quadrant of the diagram). Two time-series are considered highly coherent for a given frequency if the squared-coherency is close to 1 and the phase is close to 0 (Emery and Thomson, 2001). This is represented by

96
$$\gamma_{12}^{2}(f_k) = |G_{12}(f_k)^2|/G_{11}(f_k) G_{22}(f_k)$$
, (3)

where $\gamma_{12}{}^2(f_k)$ is the squared-coherency, $G_{11}(f_k)$ the one-sided Fourier spectrum of the first time-series for all frequencies $(f_k, k = 0, 1, ..., N - 1)$, $G_{22}(f_k)$ is the one-sided Fourier spectrum of the second time-series, and $G_{12}(f_k)$ is the cross-spectrum between the first and the second time-series (Emery and Thomson, 2001).

The standard deviation of the power measures the amplitude of the signals while the centered RMS difference provides information about the centered pattern error, derived from the geometric relationship. The signals should combine higher coherence with enough energy to be considered co-oscillating.

The STD is like the TD, where the radial distances provide the standard deviations of the power, the azimuthal position gives the squared-coherence, and the concentric labeled lines indicate the centered RMS difference. The radial lines represent the cosine of the angle made with the abscissa thus consistent with Figure 2. The reference point (usually the observation data), marked with a black dot or star, is placed on the x-axis, whereas it's the one with the maximum coherence. The test data (e.g., the model's simulations) are assessed for the ability to represent the reference data.

Although the mathematical relationship applies to two quadrants of the STD, as in the TD,
the STD is only meaningful in the first quadrant; since a negative coherence is not
applicable. The best performance is given by the test with lowest centered RMS difference,
higher coherence, and similar energy.
Figure 3 shows an example of an STD for artificial time series where there is a difference in

the amplitude and phase of the "model" results that are being inter-compared with "observations". The frequency in this case is fixed, but it is also possible to consider frequency bands. The tests are normalized by the reference standard deviation of the power. The amplitude changes are proportional to radial distances, except in the case where the test amplitude is a multiple of the reference amplitude, where the pattern described is horizontal and exactly positioned along the abscissa axis. An increase in amplitude by multiple values is expressed in the diagram as multiples of the standard deviation of the





124 normalized power. The coherence, in this case, is equal to 1, but horizontally shifted showing 125 the amplification of the signal. For example, varying the amplitude by one third (solid blue 126 line) or reducing by two thirds (dashed blue line), the coherence is reduced to 0.98 and 0.91, 127 the centered RMS difference increases to 0.80 and 0.90, while the normalized standard 128 deviation is extended to 1.77 and 0.11, respectively. 129 The coherence is highly dependent on the phase. Therefore, keeping the amplitude 130 unchanged and only varying the phase values from one quarter (dashed red line) to plus one 131 half (red line), the normalized standard deviation stays fixed, at 1, while the coherence is 132 reduced from 0.76 to 0.27, and the centered RMS difference increases from 0.69 to 1.21. 133 The power spectrum shown in Figure 3D does not clearly demonstrate the contrast when 134 varying the amplitude and phase of the time series. The STD displays both coherence and 135 power, therefore, highlighting the co-oscillating frequencies overcoming the limited 136 information contained in a power spectrum analysis. The need for better representing the 137 degree of correspondence between simulated and observed fields for a given frequency (or 138 frequency bands) is fulfilled by this novel tool inspired by relevant tide features.

139 3. Model Description

140 MOM5 (Griffies et al. 2012) is a hydrostatic (z-model), primitive equation model with free 141 surface. The model configuration used here has a global grid of 1/4°X1/4° horizontal 142 resolution, comprised of 720X1400 grid points and with 50 vertical levels. The first vertical 143 level is 10 m from the surface and vertical resolution of 10 m down to 220 m. Below this 144 depth, the levels are discretized by 166 m to the bottom. This horizontal resolution is 145 eddy-permitting and permits representation of barotropic tides. However, the model 146 resolution is insufficient to resolve internal tides. Here, we focus on the explicit barotropic 147 tidal forcing and its relevance for the contemporary HighResMIP for CMIP6 (Coupled Model 148 Intercomparison Project 6) experiments that use global climate models with a similar 1/4°X1/4° resolution for ocean models (Haarsma et al., 2016). 149

The model topography is derived from ETOPO5¹; the Boussinesq approximation is employed, and the vertical grid uses a *z** coordinate. The surface fields are extracted from the Coordinated Ocean-ice Reference Experiments (CORE1) using climatological forcing for temperature and salinity from Levitus and Boyer (1994). Surface heat fluxes, precipitation, wind stress, and river fluxes are from CORE1. Surface salinity is restored to monthly averaged climatology with a timescale of 60 days. The vertical viscosity and diffusivity are parameterized by the KPP scheme, updated from MOM4.0 to MOM4p1 to resolve the free

¹⁵⁷ ¹ <u>http://www.ngdc.noaa.gov/mgg/global/global.html</u>





surface undulation. Bryan-Lewis background diffusivity is turned off to prepare the model to
 use the barotropic dissipation from Lee et al. (2006) and baroclinic dissipation from Simmons
 et al. (2004) for future studies.

161 Three individual 20 years tidal simulations are run, which is sufficient to span the nodal tide 162 period of 18.6 years. Each 20-year simulation is initialised from the final state of a 60-year 163 spinup. We consider the spinup period to be sufficiently long for an evaluation of the upper 164 ocean. The three experiments performed include: a control run without the tidal potential, 165 hereafter referred to as CNTRL; a run equivalent to CNTRL, but with explicit tidal forcing 166 using the eight principal lunisolar constituents (M2, S2, N2, K2, K1, O1, P1, Q1 without 167 phase information, hereafter referred to as DFT (this if the default configuration of MOM5); 168 and a run equivalent to DFT, but with phase information and amplitude adopted from OSU 169 Tidal Inversion Software², hereafter referred to as T8.

170 The sea elevation adjustment is virtually instantaneous assuming the ocean is always in 171 equilibrium with tidal forces and disregarding the Darwin's correction when estimating the 172 equilibrium tide height in the presence of continents (Marchuk and Kagan, 1989). 173 Observations show that it is possible to simulate and predict the actual tide from equilibrium 174 form considering that it has been delayed and distorted slightly by the process of generation 175 and propagation (Schiller, 2004). The equilibrium tide in MOM5 is described considering the 176 tide-generating potential with corrections due to both the earth tide self-attraction and 177 loading (SAL). A scalar approximation to SAL is assumed to be equal to 0.948. The 178 equilibrium of tides described as a sum of harmonics, mainly diurnal and semidiurnal 179 constituents, is integrated into the momentum budget of the Boussinesq approximation 180 added to the transport equation, as shown in details in Griffies et al. (2004).

181 The tidal amplitude and phase adopted in the T8 experiment are based on astronomical 182 arguments used by OSU Tidal Prediction Software with initial condition dated 1st January 183 1992 00:00 Greenwich time, shown in Table 1. No update is made to the astronomical 184 argument of the partial tide that is known slightly time-dependent (Schwiderski, 1980).

185 4. Application of STDs

As described in section 2, the traditional approach to model assessment is to compare individual statistics separately. An example of such an assessment is presented in Figure 4. This figure depicts the averaged amplitude of the semidiurnal (yellow) and diurnal (green) tides for 29 tide gauge stations around the world. This includes estimates at the end of the 10th (coloured circle) and 20th (black circle) year of the DFT and T8 runs; and estimates from

¹⁹¹ ² http://volkov.oce.orst.edu/tides/otis.html





192 observations (grey circle). This comparison shows a mix of results. In some cases, the 193 modelled and observed tidal amplitudes are similar (e.g., across the Pacific Ocean), while 194 other cases there are large discrepancies between the modelled and observed tidal amplitudes (e.g., off North-Western Australia). These results are only one element of the 195 196 comparisons needed to assess the model's reproduction of the tidal signals. Arguably, a 197 better way of assessing the model's reproduction of the observed tidal signals is presented 198 in Figure 5. The values in each STD are normalized by the reference standard deviation of 199 power, enabling the inter-comparison of the change in model performance over different 200 periods of the simulation. 201 In the Atlantic, the T8 experiment shows a noticeable improvement for diurnal constituents 202 (Fig.5A) at the selected Ilha Fiscal tide gauge, which is not evident in Figure 4. The 203 coherence increased from 0.59 to 0.72 and the centered RMS error reduced from 1.03 to 204 0.95. However, T8 overestimates the energy content from 1.22 to 1.36. DFT has a better 205 response for semidiurnal constituents (coherence increases from 0.66 to 0.73) while T8 206 keeps almost unchanged. At the Gan tide gauge station it is also difficult to differentiate between the experiments in Figure 4, but it is well stated in Figure 5B that T8 improves both 207 208 semidiurnal and diurnal constituents. For both experiments, all tidal constituents are 209 underestimated at the Gan tide gauge station. In the Pacific, at Townsville tide gauge station, 210 T8 has improved the coherence for semidiurnal from 0.56 to 0.68 and for diurnal constituents 211 from 0.54 to 0.72. However, DFT has a better estimate for diurnal constituents, reducing the

centered RMS error from 0.82 to 0.65, enhancing the coherence from 0.59 to 0.76, and increased power from 0.69 to 0.83 as shown in Figure 5C. The semidiurnal constituents started with higher coherence in DFT (Figure 5C) in agreement with Figure 4, but decreased throughout the run. The higher coherence shown by the semidiurnal constituents for the vast majority of tide gauges is due to a severe underestimation of power and therefore amplitude in the model's simulations.

218 A comprehensive assessment of the model for all stations focusing on one frequency and a 219 single (Figure 6) or multiple bands of frequencies (Figure 7) is only possible by using STD. 220 A selection of the 17 tidal stations with significant data for evaluate long term frequencies is 221 done for Figure 6 and 7. It's possible to certify that the model superestimates semidiurnal 222 and diurnal bands in the Bering Sea (tide station number 12) while underestimates long term 223 and semidiurnal bands for most of tide stations. As expected, M2 is also underestimated in 224 the model expect for region close to north-eastern coast of Queensland (tide station number 225 6).





226 Separated in basins style, Figure 7 shows that is possible to evaluate multiple bands in 227 multiple regions. The model has better response to diurnal band, than in other frequencies. 228 In the Atlantic, the better response was shown close to Palmeira, Halifax and Ilha Fiscal tide stations, as shown in Figure 7. The best response of diurnal band it placed in the Indian 229 230 Ocean, where the best fit is shown in Gan tide station (shown in APPENDIX A). Australia 231 diurnal band is well represented by the model in both Indian and Pacific Oceans. The M2, 232 semidiurnal and long-term bands are underestimated in the model except for regions close 233 to Townsville (better response for M2) and Fremantle (overestimated semidiurnal band) 234 (shown in APPENDIX A).

235 5. Conclusions and Further applications

236 A new tool designed to help in the assessment and inter-comparison of model results in the 237 frequency domain is presented. The STD arguably provide a better summary of each 238 comparison - better highlighting the positive and negative aspects of each comparison. 239 STDs may benefit other studies that seek to assess models - or inter-compare models - for 240 specific frequencies, or for specific frequency bands, that might correspond to a particular 241 process of interest. For the examples used to showcase this new analysis tool in this study, 242 we showed a series of comparisons between a global model with explicit tidal forcing. 243 In contrast to Taylor Diagram, the spectral version enables a multiple band of frequencies 244 preview without using filtering techniques. Multi regions comparison it's also possible using a 245 normalized standard deviation of power strategy, that can be also useful to track model's

skills over different periods of the simulation. The versatility of STD is based on detection ofanomalous patterns in a phenomena analysis.

Although the STD has been designed for tidal analysis purpose, it is a powerful tool to detect co-oscillating patterns in multi scale analysis, and may provide a guidance in devising skill scores for inter-compare models.

251 Code and data availability

252 Spectral Taylor Diagram is an open source script available at https://github.com/mabelcalim,

²⁵³ as well as the figures plots created with IPython Notebook. The harmonic analysis based on

254 pytides (available in https://github.com/sam-cox/pytides), an open source script in python

²⁵⁵ made by Sam Cox, was parallelized in the Brazilian Supercomputer Tupã.





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338 Table 1: Global constants of tidal frequency, amplitude, and phase applied for experiments

Tidal Mode		frequency (Hz)	Love numbers	DFT	Т8				
				amplitude (m)	amplitude (m)	Phase (rad)			
	semidiurnal								
M2	principal Iunar	1.40519 10 ⁴	0.693	0.242334	0.244102	1.731557546			
S2	principal solar	1.45444 10 ⁴	0.693	0.112743	0.113568	0.000000000			
N2	elliptical lunar	1.37880 10 ⁴	0.693	0.046397	0.046735	6.050721243			
K2	declination luni-solar	1.45842 10 ⁴	0.693	0.030684	0.030879	3.487600001			
	diurnal								
K1	declination luni-solar	0.72921 10 ⁴	1.0+0.256-0.520	0.141565	0.142435	0.173003674			
01	principal Iunar	0.67598 10 ⁴	1.0+0.298-0.603	0.100661	0.101270	1.558553872			
P1	principal solar	0.72523 10 ⁴	1.0+0.287-0.603	0.046848	0.047129	6.110181633			
Q1	elliptical lunar	0.64959 10 ⁴	1.0+0.298-0.603	0.019273	0.019387	5.877717569			

339 DFT and T8. Love numbers are frequency dependent and generally close to 0.7.







Figure 1: Example of a standard Taylor Diagram (left; adapted from Oke et al. 2012); and a
Taylor Diagram with a skill-score (right; adapted from Divakaran et al. 2012).







Figure 2: Upper painel: Geometric relationship between correlation coefficient (*R*), the standard deviations of the test (σ_{-} f) and the reference (σ_{-} r) fields, and the centered root-mean-square difference (*E*^) in a TD (Taylor, 2001). Lower painel: artificial series demonstrating the validity of this relationship, root-mean-square difference (in red) equals to second term of the equation (in blue).







428 Figure 3: Spectral Taylor Diagram for displaying patterns in frequency domain shown in (C). 429 The relationship between three statistical quantities: coherence, standard deviation of power, 430 and centered RMS difference are shown. The reference point is marked with a black star, 431 plotted along the abscissa, and all the tests are normalized by the reference standard 432 deviation of power. Artificial series created for test changes in: (A) amplitude and (B) phase. 433 The frequency is fixed for all time series, where obs is the reference, amplitude is changing 434 from 0.33 to 1.33, and the phase is changing from 180° to 45°. The standard deviations are 435 proportional to radial distances, the azimuthal position gives the coherence while the 436 concentric labeled lines indicate the centered RMS difference. The Spectral Taylor Diagram 437 better expresses both the changes in amplitude and phase not captured by the power 438 spectrum analysis, shown in (D).







439 Figure 4: Amplitudes of semidiurnal (yellow) and diurnal (green) averaged tidal constituents

estimated after 10th and 20th year of simulation for T8 (top panel) and DFT (lower panel)

441 compared to tidal gauges (grey) from GLOSS.











- 442 Figure 5: STDs comparing sea-level from T8 and DFT with tidal gauge observations during 443 DJF for an example in the A) Atlantic Ocean (Ilha Fiscal); B) Indian Ocean (Gan); and C) Pacific Ocean (Townsville) - showing the semidiurnal (yellow) and diurnal (green) averaged 444 constituents for DFT (squares) and T8 (stars) at the end of the 10th (filled markers) and 20th 445 (unfilled markers) year of each simulation. The 10th year is connected with 20th year of 446 447 simulation by a line showing the evolution of the skill of the model's in relation of tide gauge 448 data. The standard deviations of power have been normalized by the observed standard 449 deviation of power. The reference (black star) is also divided into two time periods: 10 years
- 450 and 20 years after initial condition of 01/01/1992.







451 Figure 6 - STDs overall tide gauges stations with significant data (17) separated by 452 frequencies. Upper left: long-term band; Upper right: diurnal band; Lower left: semidiurnal 453 band; and Lower right: M2 frequency. The model does a better job close to Townsville tide 454 station for M2 frequency, diurnal and long-term bands.







⁴⁵⁵ Figure 7. STD overall tide gauges stations with significant data (5) in Atlantic Ocean.





456 APPENDIX A





⁴⁵⁷ Figure A1. STD overall tide gauges stations with significant data (6) in Indian Ocean.



⁴⁵⁸ Figure A2. STD overall tide gauges stations with significant data (5) in Pacific Ocean.