

Accurate Magnitude and Stress Drop using the Empirical Green's Function Method Applied to Distributed Acoustic Sensing

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Abstract

The reliable estimation of earthquake magnitude and stress drop are key in seismology. The novel technology of distributed acoustic sensing (DAS) holds great promise for source parameter inversion owing to the measurements' high spatial density. In this study, I demonstrate the robustness of DAS for magnitude and stress drop estimation using the empirical Green's function deconvolution method. This method is applied to 7 collocated earthquakes recorded in Israel following the 2023 Turkey earthquakes. Spectral ratios were calculated using amplitude spectra stacked along the fiber, and fitted with a relative Boatwright source spectral model. Excellent fits were obtained even for similar sized earthquakes. Stable seismic moments and stress drops were calculated assuming the moment of one earthquake is known. These robust source parameters suggest that stress drop increases with seismic moment. The results demonstrate the great potential of DAS for source studies.

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2 **Applied to Distributed Acoustic Sensing**

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6 **Key Points:**

- 7 • The Empirical Green's Function (EGF) approach was applied to 7 repeating DAS
8 recorded earthquakes in Israel
- 9 • Accurate relative magnitudes and stress drops were calculated and stress drop was found
10 to increase with seismic moment
- 11 • The results demonstrate the great potential of DAS for earthquake source studies

12 Abstract

13 The reliable estimation of earthquake magnitude and stress drop are key in seismology.
 14 The novel technology of distributed acoustic sensing (DAS) holds great promise for source
 15 parameter inversion owing to the measurements' high spatial density. In this study, I demonstrate
 16 the robustness of DAS for magnitude and stress drop estimation using the empirical Green's
 17 function deconvolution method. This method is applied to 7 colocated earthquakes recorded in
 18 Israel following the 2023 Turkey earthquakes. Spectral ratios were calculated using amplitude
 19 spectra stacked along the fiber, and fitted with a relative Boatwright source spectral model.
 20 Excellent fits were obtained even for similar sized earthquakes. Stable seismic moments and
 21 stress drops were calculated assuming the moment of one earthquake is known. These robust
 22 source parameters suggest that stress drop increases with seismic moment. The results
 23 demonstrate the great potential of DAS for source studies.

24 Plain Language Summary

25 Estimating earthquake magnitude and stress drop, the two most fundamental source
 26 parameters, is key for various seismological investigations, including earthquake self-similarity
 27 and hazard mitigation. The extraction of source parameters requires that source and path effects
 28 are reliably deconvolved from source contributions. One approach is to deconvolve a small
 29 earthquake from a colocated larger one in the frequency domain such that common path and site
 30 functions cancel and relative source function, between both earthquakes, remain. In this study, I
 31 apply the deconvolution approach to 7 earthquakes that followed the 2023 Turkey earthquakes,
 32 recorded using distributed acoustic sensing (DAS) applied to an optical fiber in Israel. DAS
 33 converts standard optical fibers into dense arrays, with seismic measurements every few meters
 34 along tens-of-kilometers long fibers. The deconvolution was calculated for different earthquake
 35 pairs and excellent fits were obtained with a relative source spectral model. Stable seismic
 36 moments and stress drops were calculated and their analysis suggests that stress drop increases
 37 with seismic moment. The results demonstrate the great potential of DAS for source studies.

38 1 Introduction

39 Estimating earthquake magnitudes and stress drops is fundamental for seismological
 40 research. Stress drop is key in studying earthquake self-similarity (e.g., Ide & Beroza, 2001) and
 41 strongly affects ground shaking intensity (e.g., Lior & Ziv, 2018). The estimation of these source
 42 parameters is model based, and is achieved using earthquake observations by isolating source
 43 contributions from path and site effects. Once the source term is extracted, it is fitted with a
 44 source model either in time (Al-Ismail et al., 2023; Lior & Ziv, 2018) or frequency (e.g., Huang
 45 et al., 2017; Lior & Ziv, 2017; Shearer et al., 2019) domains. The most commonly used model is
 46 the Omega-squared (Boatwright, 1980; Brune, 1970; Madariaga, 1976; Sato & Hirasawa, 1973),
 47 describing the far-field body-wave (P- or S-wave) radiation. For ground displacement it reads as:

$$\Omega(f) = \frac{\Omega_0 U_{\phi\theta}}{\left[1 + \left(\frac{f}{f_0}\right)^{2\gamma}\right]^{1/\gamma}}, \quad (1)$$

48 where γ equals 1 (Brune, 1970) or 2 (Boatwright, 1980), f_0 is the source corner frequency, Ω_0 is
 49 the low frequency displacement spectral plateau, and $U_{\phi\theta}$ is the radiation pattern. The latter is a
 50 function of the focal mechanism, azimuth and take-off angle from the source (Aki & Richards,
 51 2002; Kaneko & Shearer, 2015). Ω_0 is proportional to the seismic moment M_0 divided by the

52 hypocentral location R ($\Omega_0 \propto M_0/R$), and to parameters that describe the media (free surface
53 amplification factor and density and velocity at the source). Stress drop is commonly estimated
54 assuming a simple circular crack with uniform rupture velocity (Eshelby, 1957):

$$\Delta\tau = \frac{7}{16} M_0 \left(\frac{f_0}{kC_s} \right)^3, \quad (2)$$

55 where C_s is the shear-wave velocity and k is a phase dependent (P or S) model constant.
56 Equations (1) and (2) describe a simple source model with a smooth, isotropic slip distribution
57 on a single circular patch.

58 Station-specific corner frequency and stress drop estimates exhibit large variability for
59 the same earthquake for various reasons (e.g., Abercrombie et al., 2017; Kaneko & Shearer,
60 2015; Lior & Ziv, 2018). These variabilities are mainly attributed to the improper extraction of
61 source from path and site contributions, a hindrance that aggravates in the presence of low signal
62 to noise ratio (SNR) conditions, i.e., for small earthquakes. In addition, since stress drops are the
63 cube of corner frequencies (Equations 2), small errors in f_0 translate to larger errors in $\Delta\tau$.

64 A common method for the reliable extraction of source contributions from path and site
65 effects, is the empirical Green's function (EGF), or spectral ratios, method. This method uses
66 two collocated earthquakes recorded at the same location to deconvolve the smaller earthquake
67 (the empirical Green's function) from the larger one to remove common path, site and
68 instrumental effects. The deconvolved spectra is then fitted with a relative source model to
69 estimate the seismic moment ratio between both earthquakes, and the corner frequencies of each
70 earthquake (e.g., Abercrombie et al., 2017). This approach circumvents the modeling of path and
71 site effects, overcoming the main difficulty in estimating source parameters.

72 However, the EGF method suffers from several drawbacks that reduce its performance
73 and applicability. First, the magnitudes of both earthquakes need to be sufficiently different in
74 order to reliably fit both corner frequencies. Typically, a magnitude difference of at least 1 unit is
75 required (e.g., Abercrombie et al., 2017; Shearer et al., 2019). Second, while path and site effects
76 are assumed to be identical for both earthquakes, it is not necessarily the case for noise, and
77 spectral ratios need to be stacked to suppress random noise sources and facilitate reliable model
78 fitting (e.g., Wu et al., 2019). Third, EGF analysis is only feasible in a limited frequency band
79 where the spectral amplitudes of both earthquakes are above a predefined noise threshold,
80 typically set to be 3 (e.g., Viegas et al., 2010). Using a too small frequency range may bias the
81 fitted source parameters (Abercrombie, 2021). The combined requirement of magnitude
82 difference and SNR thresholds coupled with the need to stack multiple spectral ratios limits the
83 applicability of the EGF method.

84 It has recently been demonstrated that the dense measurements of Distributed Acoustic
85 Sensing (DAS) can be used to obtain spectral ratios (Chen, 2023). The use of DAS for source
86 parameter estimation via the EGF method can overcome the previously described limitations of
87 seismometer/accelerometer-based method. The application of the EGF method to DAS data is
88 described in Text S1, with the conclusion that DAS strain (rate) data can be directly used to
89 obtain spectral ratios (Equation S3):

$$\frac{\dot{\epsilon}_{1j}^e(f,x)}{\dot{\epsilon}_{2j}^e(f,x)} = \frac{M_{01} \left[1 + \left(\frac{f}{f_{02}} \right)^{2\gamma} \right]^{1/\gamma}}{M_{02} \left[1 + \left(\frac{f}{f_{01}} \right)^{2\gamma} \right]^{1/\gamma}}. \quad (3)$$

90 In this equation, $\dot{\epsilon}_{ij}^e(f, x)$ are strain rates of earthquake i ($i = 1, 2$) recorded at DAS channel j ,
91 and M_{0i} and f_{0i} are earthquake specific seismic moments and corner frequencies, respectively.

92 In this work, I demonstrate that high fidelity relative seismic moments, corner
93 frequencies and stress drops can be obtained using DAS via the EGF method. For each
94 earthquake, the entire fiber was reduced to a single spectrum by stacking move-out corrected
95 Fourier spectra along the full fiber length. These earthquake specific spectra were used to obtain
96 spectral ratios that were analyzed in the framework of Equation (3).

97 This manuscript is organized as follows. First, I describe the DAS data and processing.
98 Then, I demonstrate the method using 7 well-recorded co-located earthquakes that occurred in
99 central Israel following the devastating 2023 Kahramanmaraş Mw7.8 and Mw7.6 Eastern
100 Turkey earthquakes. Finally, I discuss implications for earthquake self-similarity.

101 **2 Data and Preprocessing**

102 Following the catastrophic 2023 Kahramanmaraş Mw7.8 and Mw7.6 earthquakes in
103 Eastern Turkey, Israel experienced a rapid increase in seismicity, including several earthquake
104 clusters that occurred off the main Dead-Sea-Transform fault. Here, I focus on one of these
105 clusters of co-located repeating earthquakes that were recorded by a nearby optical fiber between
106 February 7th and 18th. Earthquakes, fiber and seismometer location are shown in Figure 1a.
107 During this period, 10 earthquakes magnitudes 2.5 to 3.4 were recorded and cataloged by the
108 Geological Survey of Israel. Earthquake information is listed in Table S1.

109 DAS earthquake data were acquired using a Prisma Photonics interrogator unit. The
110 system is harnessing an innovative Hyper-Scan technology to produce high-quality quantitative
111 longitudinal strain-rate measurement. Measurements were made using a 34 km long fiber
112 installed in central Israel (Figure 1a) with a gauge length of 18.2 m, spatial sampling of 9.1 m
113 and sampling rate of 1500 Hz. Data were then resampled to 100 Hz. Additional seismometer data
114 from the Israeli seismic network (Kurzon et al., 2020) were used (Figure 1a).

115 In order to verify that the earthquakes qualify as EGF pairs, cross correlations (CC) were
116 calculated using the nearest seismic station. Because DAS data were acquired using a horizontal
117 fiber, with reduced P-wave sensitivity (e.g., Papp et al., 2017), the analysis focused only on S-
118 waves. Horizontal acceleration time-series were low pass filtered at 5 Hz and CC were calculated
119 in a 1 second window starting at the direct S-wave arrival at MEFR station (Figure 1a), located
120 10 km from the epicenter of the largest earthquake. Filtered time-series and CC values with
121 respect to the largest earthquake are shown in Figure 1b. All CC values for both horizontal
122 components were above 0.8, indicating significant similarity between the different earthquakes.
123 Thus, it is safe to assume that these repeating earthquakes are nearly co-located and have a
124 similar mechanism. In subsequent analysis, the catalog location of the largest earthquake
125 (Mw3.4, Table S1) is used for all events.

126 Waveform similarity is also evident in the DAS recordings shown in Figure 1c, where the
127 first 3 seconds of the S-wave are shown on a 2.2 km fiber segment. DAS CC were not calculated
128 because of the longer epicentral distance (~40 km) and because single channel DAS recordings
129 are typically characterized by lower SNR and less coherent wavefield compared to seismometers
130 (Lior, Sladen, Rivet, et al., 2021; van den Ende & Ampuero, 2021). Similar features can be
131 observed between the different earthquakes, for example, the coherent arrivals at the first
132 kilometer and first half a second.

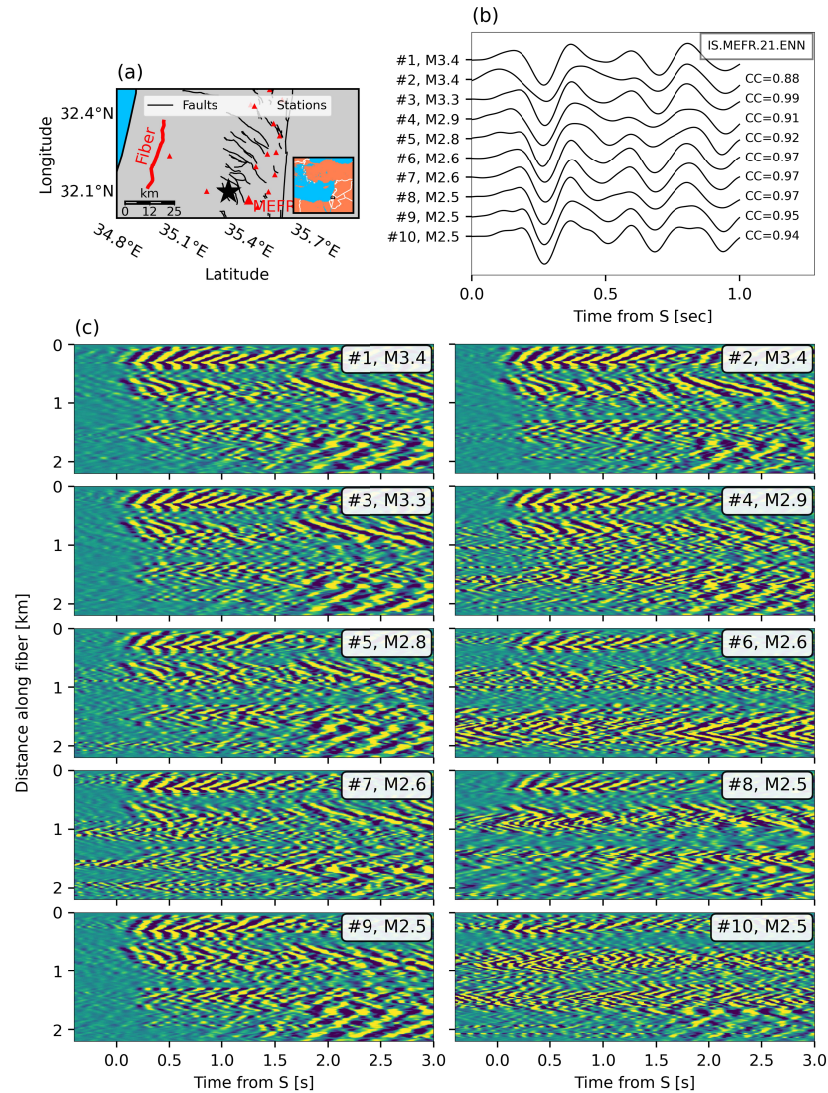


Figure 1. (a) Map of catalog earthquake location (black start), seismometer (red triangles), fiber (red curve) and known faults (black curves, from Sharon et al., 2020). The region is indicated by a black rectangle in the inset map. (b) The first second of the S-waves recorded by the north component of MEFR station are shown for all earthquakes. Waveforms are normalized and vertically shifted for visibility (catalog magnitudes indicated on the vertical axis; numbering as in Table S1). CC values for each waveform and the top one are indicated. (c) DAS strain-rate recordings of the first 3 seconds of the S-wave on a 2.2 km fiber segment for the different earthquakes. Each earthquake is normalized by its standard deviation. Earthquake numbering (corresponding to Table S1) and magnitudes are reported in each panel.

133 To obtain earthquake specific stacked Fourier spectra, few processing procedures were
 134 applied. First, theoretical arrival times from the earthquake source to all fiber channels
 135 were calculated and a move-out correction for the direct S-wave was applied. I use a uniform data
 136 interval of 20 seconds for all earthquakes, starting 1 second before the S-wave arrival. This data
 137 window is longer than the expected source durations of the analyzed earthquakes, and was
 138 chosen since it 1) produces reliable model fits (as described in the next section), 2) increases the

139 low frequency resolution that is useful to fit the lower frequency part of the spectra, and 3) yields
 140 a spatially averaged wavefield (Abercrombie, 2013; Baltay et al., 2010; Mayeda et al., 2007) that
 141 compensates for the single sensing direction (along the fiber) and the limited azimuthal
 142 observations of the earthquakes (fiber aperture of $\sim 45^\circ$). Move-out corrected strain-rates were
 143 then tapered using a 10% Tukey function and amplitude spectra were calculated and stacked by
 144 geometric averaging of channel-specific spectra.

145 Stacking amplitude spectra along the fiber has a pronounced smoothing effect. This
 146 behavior is visualized in Figure S1 for the M3.3 and M2.8 earthquakes, where seismometer
 147 spectra (panel a) and single DAS channel spectra (panel b) are compared to stacks of 1, 5, 19 km
 148 and the full fiber (panels c, d, e and f, respectively). When stacking over a longer spatial extent,
 149 local noises are suppressed and laterally coherent earthquake signals emerge, increasing the
 150 smoothness of the spectra. Furthermore, stacked spectra allow for clear separation between
 151 similar sized earthquakes which allows for reliable spectral ratio analysis, as further
 152 demonstrated.

153 The analysis is only reliable in a sufficiently high SNR frequency band that needs to be
 154 identified. Here, I used the same frequency band for all EGF pairs in order to reduce model
 155 fitting biases (Abercrombie, 2021). To determine the usable bandwidth, I calculated SNR using
 156 the spectra of the signal and pre-signal noise, both stacked over the entire fiber. The latter is
 157 calculated in an equivalent 20 second window. Stacked earthquake strain-rate amplitude spectra
 158 and stacked SNR (stacked signal divided by stacked noise) curves are plotted in Figure S2. To
 159 better determine the usable frequency band, SNR curves were resampled in equal log of
 160 frequency bins (e.g., Lior & Ziv, 2017) (panels b and d of Figure S2). Usually, EGF studies
 161 require an SNR threshold of at least 2 (e.g., Shearer et al., 2019; Viegas et al., 2010), however,
 162 when using stacked signals, this threshold can be relaxed, since stacked signal and noise spectra
 163 are smooth and exhibit less amplitude scatter, allowing for more precise differentiation of signal
 164 and noise. Thus, I only require that the stacked signals' spectral amplitudes are 10% above noise
 165 levels (SNR=1.1, horizontal black line in panels b and d of Figure S2). Even though this analysis
 166 facilitates a larger bandwidth, I set the usable bandwidth to be between 2 and 20 Hz. The validity
 167 of this frequency band will be later verified by inspecting model fits and parameter uncertainties.
 168 Two earthquakes did not meet the required SNR condition (#6 and #10) with low SNR between
 169 10 and 20 Hz (Figure S2d), and an additional earthquake (#8) had unexpectedly high amplitudes
 170 at low frequencies (~ 0.5 to ~ 4 Hz) that deviate from the omega squared model in Equation (1)
 171 (dashed black curve in Figure S2a). In the following, the remaining 7 earthquakes are analyzed.

172 **3 Spectral Ratios Application to DAS Data**

173 Earthquake-specific stacked spectra were used to construct spectral ratios for different
 174 earthquake pairs. Model fitting and parameter uncertainty estimation is similar to that proposed
 175 by Viegas et al. (2010). The stacked spectra of the larger earthquakes are divided by those of the
 176 smaller ones. This procedure is mathematically equivalent to calculating the spectral ratios at
 177 every DAS channel, followed by stacking, because stacking is done via geometric averaging.
 178 The spectral ratios are then resampled in equal log-of-frequency bins and the logarithm of the
 179 spectral ratios are fitted with the logarithm of Equation (3) to obtain the best fitting
 180 $\log(M_{01}/M_{02})$, f_{01} and f_{02} estimates. The inversion was done using a non-linear least square
 181 approach, implemented via Python's SciPy `curve_fit` tool. Parameter uncertainties were
 182 estimated by varying the value of one parameter and calculating the standard deviation of the fit

183 while treating the other two parameters as free fitting coefficients. The parameter-specific
 184 uncertainty range is determined by a standard deviation increase of 5%.

185 The fits to the spectral ratios improve when the latter are stacked over longer fiber
 186 segments as demonstrated in Figure 2 and S3. In these examples, spectral ratios are obtained for
 187 magnitude 3.3 and 2.8 earthquakes using MEFR accelerometer (panel a), a single DAS channel
 188 (panel b), stacks of 1, 5 and 19 km (panels c, d and e, respectively), and the full fiber (panel f).
 189 Fits are quantified using the standard deviation and R^2 (panel legends), and corner frequency
 190 uncertainties (error bars). The fits in Figures 2 and S3 are obtained using the Boatwright ($\gamma = 2$
 191 in Equation 3) and Brune ($\gamma = 1$ in Equation 3) models, respectively. The Boatwright model,
 192 with sharper corner frequencies, clearly outperforms the Brune model, producing better fits and
 193 lower parameter uncertainties. Thus, the Boatwright model is used in following analysis. While
 194 using a single accelerometer (panels a) produced slightly better fits than a single DAS channel
 195 (panels b), stacking over longer segments significantly reduces the scatter in the spectral ratios
 196 and dramatically improves the fits (panels c-f).

197 I calculate model fits to all EGF pairs using full fiber stacks, only considering pairs
 198 whose magnitudes differ by at least 0.4 units. The spectral ratios, fits and goodness of fit
 199 parameters are shown in Figure 3 for 12 earthquake pairs. Highly reliable fits are obtained, with
 200 consistent (within error bars) earthquake-specific corner frequencies among different EGF pairs.

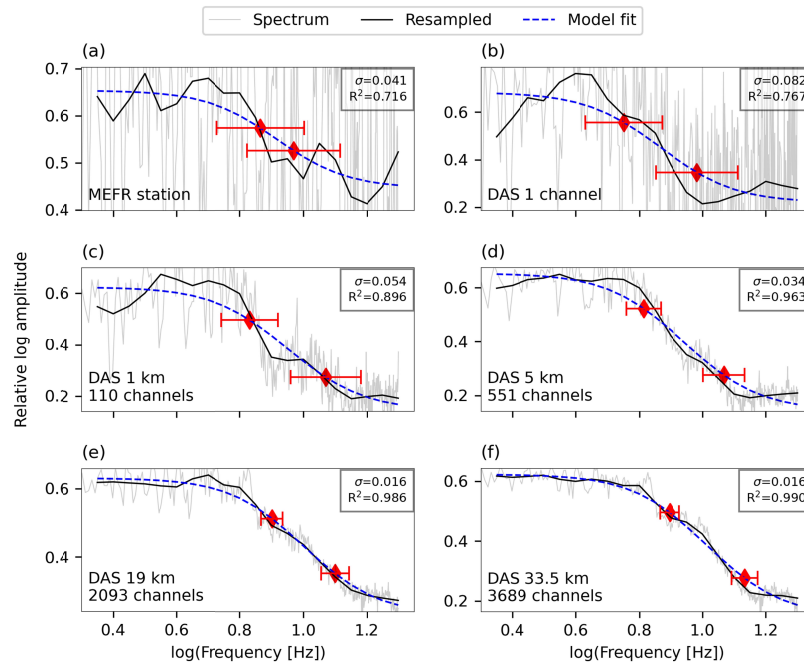


Figure 2. Spectral ratios and fits using the Boatwright ($\gamma = 2$ in Equation 3) model for magnitudes 3.3 and 2.8. Spectral ratios at (a) MEFR station (geometric mean of both horizontal components), (b) a single DAS channel, and stacks of (c) 1 km, (d) 5 km, (e) 19 km and (f) the full fiber. DAS stacks were centered around km 19 along the fiber. Spectral ratios or stacked ratios, resampled ratios, and model fits are indicated by thin grey curves, black curves, and blue curves, respectively. Standard deviations and R^2 are indicated in the panel legends, corner frequency uncertainties are indicated by red error bars.

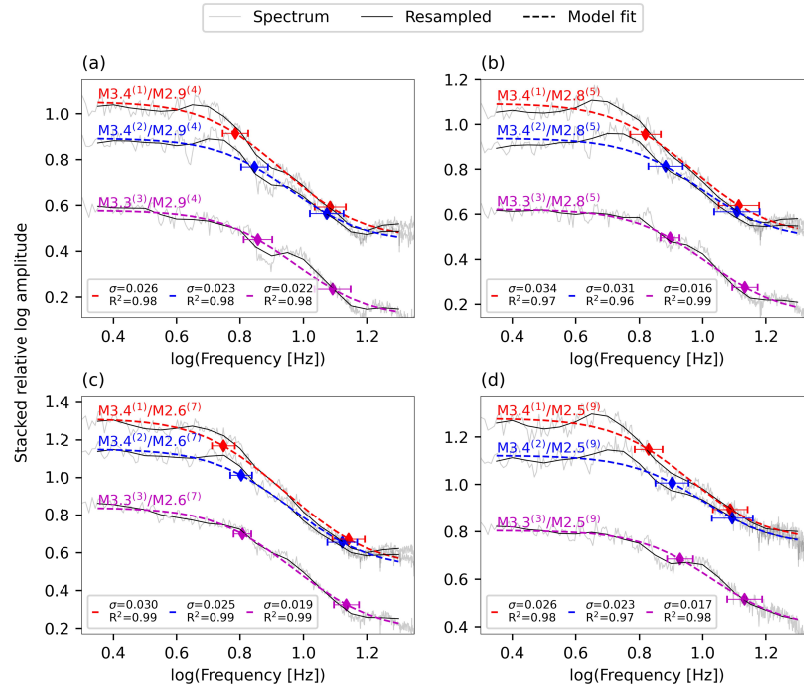


Figure 3. Spectral ratios fits using the Boatwright ($\gamma = 2$ in Equation 3) model for different earthquake pairs. Stacked spectral ratios, resampled ratios and model fits are indicated by thin grey curves, black curves, and colored curves, respectively. Standard deviations and R^2 are indicated in the panel legends, corner frequency uncertainties are indicated by horizontal error bars. Red, blue and purple curves correspond to different larger earthquake while each panel shows a specific smaller earthquake.

201 4 Implications for Source Parameter Studies

202 The spectral ratios method produces accurate corner frequencies, but only relative
 203 seismic moments; for accurate stress drops, reliable moment estimations are needed (Equation
 204 2). The obtained results are unique in that each of the 7 earthquakes is used for 3 or 4 different
 205 EGF pairs (Figure 3), resulting in several corner frequency and relative moment estimates per
 206 earthquake, such that all moments may be estimated if one is known. Here, seismic moments
 207 were determined using a linear least-squares inversion, detailed in Text S2, where the magnitude
 208 of the largest earthquake was fixed to its catalog moment magnitude of 4.3 (Table S1). The
 209 inversion produced stable earthquake specific seismic moments (Figure S4). Stress drops were
 210 then estimated via Equation 2 using the earthquake-specific seismic moments, the different
 211 corner frequencies, $k = 0.21$ (Madariaga, 1976) and $C_s = 3.2$ km/s. Event averaged source
 212 parameters are listed in Table S1. Figure 4 plots corner frequencies (panel a) and stress drops
 213 (panel b) as functions of earthquake-specific seismic moments. Fit-specific corner frequencies
 214 and stress drops are indicated by grey circles while event averages are indicated by red circles.

215 Corner frequency and stress drop within-event variabilities are larger for the 3 largest
 216 earthquakes compared to the 4 smallest ones (Figure 4), possibly since the larger corner of the
 217 EGF pairs, associated with the smallest earthquakes, is typically sharper than the smaller corner,
 218 associated with the largest earthquakes (Figure 3). This may be an artifact of the limited
 219 bandwidth, although corner frequencies are well resolved, present small uncertainties and are

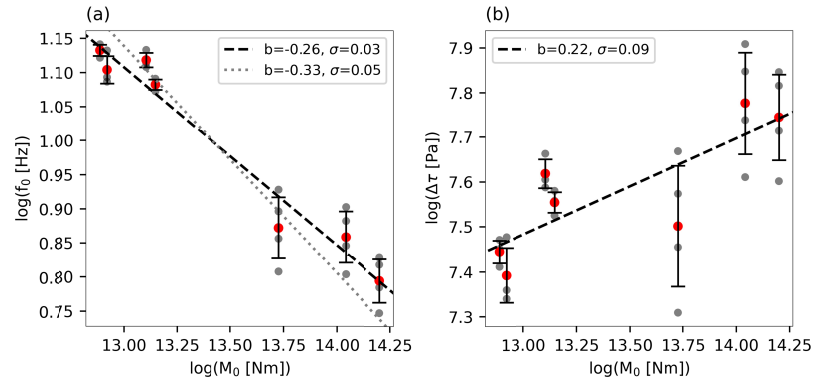


Figure 4. The logarithms of (a) corner frequencies and (b) stress drops as functions of the logarithm of seismic moments. Seismic moments were obtained using the least-squares inversion (Text S2) per earthquakes. Fit-specific and event averaged f_0 and $\Delta\tau$ are indicated by grey and red circles, respectively. Vertical error bars are the standard deviations to the different estimates. Linear fits to event averages are indicated by dashed black lines with slopes and standard deviations indicated in panel legends. The dotted curve in (a) is the fit assuming earthquake self-similarity (constant stress drop), i.e., a slope of $-1/3$.

220 sufficiently within the bandwidth limits (2 to 20 Hz). Alternatively, the observed scatter in the
 221 corner frequencies of the larger earthquakes may suggest complexities and variabilities among
 222 the noise conditions or source functions of the smaller earthquakes, that thus do not perform as
 223 ideal Green's functions.

224 Stress drop within-event variabilities are generally smaller than between-event
 225 variabilities (Figure 4b) demonstrating the robustness of these estimates and allowing for the
 226 analysis of stress drop variations with seismic moment. Stress drops span ~ 0.4 orders of
 227 magnitude while the available seismic moments span 1.3 orders of magnitude. In spite of the
 228 limited parameter ranges, it is clearly seen that the stress drops of the largest earthquakes are
 229 larger than those of the smaller ones by a difference of ~ 0.4 orders of magnitude (Figure 4 and
 230 Table S1). Linear fits to event averages in Figure 4 support this trend, indicating that $\log(\Delta\tau) \propto$
 231 $0.22 \log(M_0)$. However, given the small number of observations and limited magnitude range,
 232 this scaling relation should be treated with caution and cannot be reliably extrapolated. Further
 233 investigations into the temporal evolution of the studied earthquakes considering both local
 234 (seismicity patterns near the Dead-Sea Transform) and regional (the relation to the 2023 Turkey
 235 earthquakes) effects is required to better interpret the obtained source parameters.

236 5 Conclusions

237 In this work, I demonstrated the robustness of DAS for source parameter estimation via
 238 the spectral ratios method applied to 7 repeating earthquakes in central Israel. Stacking move-out
 239 corrected amplitude spectra along the fiber significantly reduced the effect of incoherent and
 240 locally-coherent noises, producing a smoother and more coherent earthquake spectra. These
 241 stacked spectra allow for the application of the EGF method even for similar sized earthquakes. I
 242 calculated spectral ratios for different EGF pairs, and obtained excellent fits to the model with
 243 robust and stable estimates of seismic moment ratios and both corner frequencies. Fits obtained
 244 using the Boatwright model (Figure 2) significantly outperform those obtained using the Brune
 245 model (Figure S3). Having prior knowledge of one earthquake moment magnitude, I performed a
 246 linear least-squares inversion to obtain earthquake specific seismic moments and stress drop.

247 The obtained source parameters are very reliable, as indicated by the excellent spectral
248 ratios model fits, allowing for in-depth investigations of source properties and stress drop scaling
249 with seismic moment. For these earthquakes, Source parameter analysis revealed that stress drop
250 increases with seismic moment. However, a more in-depth analysis of stress drop behavior
251 considering the temporal evolution of the analyzed sequence is required.

252 The results demonstrate the great potential of DAS for earthquake source studies: 1)
253 Owing to the abundance of DAS measurements, stacked along the fiber, SNR requirements were
254 reduced to $\text{SNR} > 1.1$, allowing for less restrictive application of the EGF method. This property
255 is especially beneficial for small earthquakes ($M < 3$) that typically suffer from low SNR and are
256 reliably recorded by fewer stations. Furthermore, having additional possible earthquake pairs for
257 the same dataset, improves the stability of source parameter inversion. 2) Spectral ratios stacked
258 along the fiber exhibit excellent fits to the model and circumvent the need to stack over many
259 seismic stations in a broad azimuthal range, potentially facilitating investigation of azimuthal
260 source parameter variations. This property may also allow for better separation of source from
261 path and site effects. 3) The EGF method applied to DAS circumvents the need to convert DAS
262 measurements to ground motions for source parameter inversion (Lior et al., 2023; Lior, Sladen,
263 Mercerat, et al., 2021). Finally, DAS can improve our knowledge of the relation between source
264 parameters and ground motion variabilities, as well as our understanding of source complexities
265 in various sized earthquakes.

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268 acquisition, handling and transfer.

269 **Open Research**

270 Seismometer data can be downloaded from the Geological Survey of Israel
271 (<https://eq.gsi.gov.il/en/earthquake/searchEQS.php>), and data to reproduce the results is found at
272 <https://osf.io/y8csp/>.

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