USER-MANUAL

GENERALIZED GROUND MOTION MODEL (GGMM) (SUBDUCTION)

This tool, named as Generalized Ground Motion Model for Subduction environment (GGMMSubd), uses a hybrid Recurrent Neural Network (RNN) framework to estimate a 35×1 cross -dependent vector (denoted as IM) of RotD50 Spectral Acceleration (RotD50 S_a) at 31 periods and geometric means of Arias Intensity ($I_{a_{geom}}$), Significant Duration ($D_{5-95_{geom}}$), Peak Ground Acceleration, (PGA_{geom}) and Peak Ground Velocity (PGV_{geom}) using a set of seismic source and site parameters as inputs. The source and site inputs to the RNN framework include a vector of 6 values including Subduction fault slab mechanism (F), magnitude (M_w), closest rupture distance (R_{rup}), Joyne-Boore distance (R_{JB}), soil shear-wave velocity (V_{s30}), and hypocentral depth (Z_{hyp}). The residuals of the RNN framework are used to construct between-event and within-event covariance matrices to account for the between-event and within-event variabilities of the ground motions. Hence, given the source and site parameters, this tool returns a median prediction of the IM and estimated correlated variance bands. The executable is developed by Jawad Fayaz (https://jfayaz.github.io/layouts/codeandsoft.html/) and collaborators (Miguel Medalla, Pablo Torres, and Carmine Galasso). For further details please read the article mentioned in the "Reference".

Download the application from the following Dropbox link

https://www.dropbox.com/scl/fo/lqcz9p7sk2mea3a1itlx5/AD6Z9OJWoQMyk5Y-svSxKH0?rlkey=l0wh0v5le3nu05mz8rtrjqu87&dl=0

1. GGMMSubd Inputs (in order)

i. Subduction Fault Mechanism (F)

Subduction Mechanism (F)	Value
Interslab	0
Intraslab	1

- ii. Magnitude (M_w): $3 \le M_w \le 9$
- iii. Closest Rupture Distance (R_{rup}) in kilometers (km): $0 \le R_{rup} \le 300$
- iv. Depth of Hypocenter (Z_{hyp}) in kilometers (km): $0 \le Z_{hyp}$

- v. Joyne-Boore Distance (R_{JB}) in kilometers (km): $0 \le R_{JB} \le 300$
- vi. Shear-Wave Velocity (V_{s30}) in meters per second (m/s): $0 \le V_{s30} \le 3000$
- vii. Conditional Period (T^*)

IM	Input for <i>T</i> *
PGV_{geom}	-3
$D_{5-95_{geom}}$	-3 -2
$I_{a_{geom}}$	-1
PGA_{geom}	0
T = 0.01	0.01
T = 0.05	0.05
T = 0.1	0.1
T = 0.15	0.15
T = 0.2	0.2
T = 0.25	0.25
T = 0.3	0.3
T = 0.4	0.4
T = 0.5	0.5
T = 0.6	0.6
T = 0.7	0.7
T = 0.8	0.8
T = 0.9	0.9
T = 1.0	1.0
T = 1.2	1.2
T = 1.4	1.4
T = 1.6	1.6
T = 1.8	1.8
T = 2.0	2.0
T = 2.25	2.25
T = 2.5	2.5
T = 2.75	2.75
T = 3.0	3.0
T = 3.25	3.25
T = 3.5	3.5
T = 3.75	3.75
T = 4.0	4.0
T = 4.25	4.25
T = 4.5	4.5
T = 4.75	4.75
T = 5.0	5.0

viii. Name of Output Folder (OutputFolderName)

2 Calling GGMMSubd

The tool package consists of the executable application "GGMMSubd.exe" which can be easily called from any command line or programming language/software. An example to run the GGMM program is given in Figure 1 where the inputs are in the same order as mentioned in above section "GGMM Inputs". The generalized syntax to run the executable is as follows:

GGMMSubd.exe F M_w R_{rup} Z_{hyp} R_{JB} V_{s30} T^* OutputFolderName

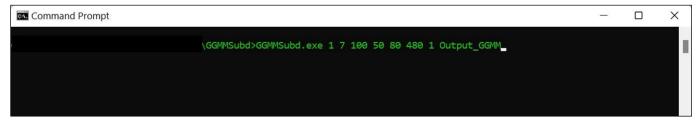


Figure 1: Calling "GGMMSubd.exe"

In case all the inputs are not properly provided the tool will throw an error as shown in Figure 2.

Figure 2: Error screen of "GGMMSubd.exe"

3 GGMMSubd Outputs

The tool creates a folder named as inputted by the user in the *OutputFolderName* (as described) within the current directory of the tool. The output screen of the framework is shown in Figure 3.

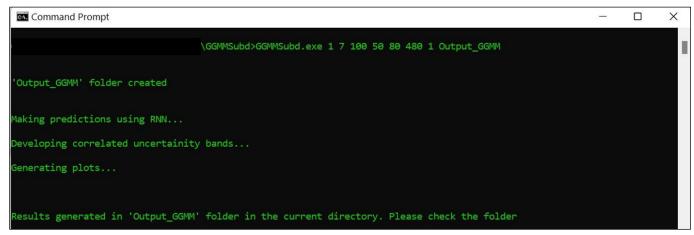


Figure 3: Output screen of "GGMMSubd.exe"

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The outputs consist of two files: 1) "GGMM.out" file containing the estimated median **IM** predictions and its conditional correlated variance bands (hence there is no variability at T^*) and 2) "GGMM.jpg" file showing the median and sigma bands of the estimated intensity measures in **IM** vector. The outputs are shown in Figures 4 and 5.

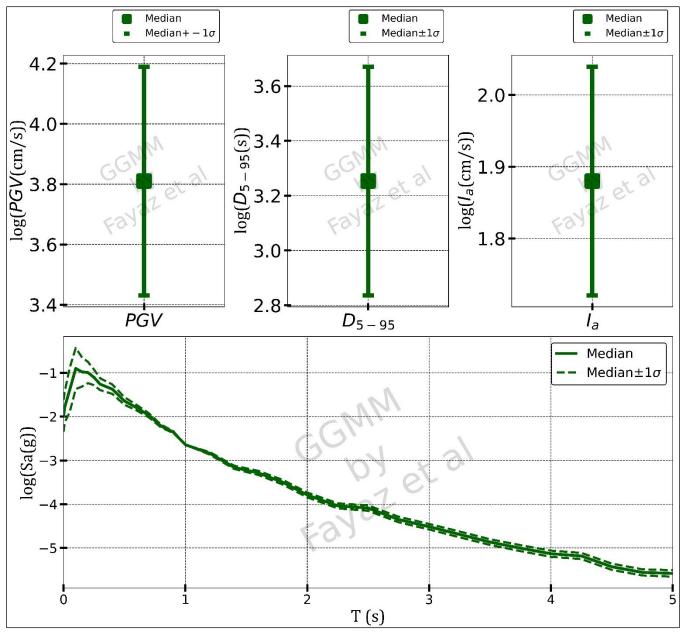


Figure 4: Estimated IM output

```
🚽 GGMM.out 🔣
    Т
        Median
                Median-sigma
                                Median+sigma
    PGV -0.37695977091789246
                                -0.37695977091789246
                                                        -0.37695977091789246
 3
    D595
            2.84094500541687
                                1.9415109758606557 3.7403790349730848
    Ia -0.12560594081878662
                                -0.6437176423300242 0.3925057606924509
                                                -3.433447468065836
    PGA -3.9913647174835205 -4.549281966901205
            -3.4784090518951416 -4.019641574367986 -2.937176529422297
    0.01
    0.05
            -3.1990227699279785 -3.7877270182943117 -2.6103185215616453
    0.1 - 2.7986066341400146 - 3.4092391452642703 - 2.187974123015759
    0.15
            -2.496680974960327
 9
                                -3.0927318929626475 -1.9006300569580068
    0.2 -2.737539529800415
10
                            -3.3293273380219186 -2.1457517215789115
    0.25
            -2.6430089473724365 -3.2195390012699114 -2.0664788934749616
    0.3 -2.865675449371338
12
                            -3.428094653340792
                                                -2.303256245401884
13
    0.4 - 3.1758267879486084 - 3.7518747284814182 - 2.5997788474157986
    0.5 -3.1301863193511963 -3.7030378114551072 -2.5573348272472853
                            -3.860323643904333
15
    0.6 -3.288729429244995
                                               -2.717135214585657
    0.7 - 3.7680866718292236 - 4.330125305414816 - 3.2060480382436314
    0.8 -3.741262912750244
                            -4.29845794290837
                                                -3.1840678825921183
    0.9 - 4.0367889404296875 - 4.588907961692457 - 3.4846699191669184
    1.0 -4.398294925689697 -4.94463090435241
                                                -3.851958947026984
    1.2 -4.965432643890381 -5.494882551893186 -4.435982735887576
    1.4 -5.36202335357666
                            -5.879500909413421 -4.844545797739899
    1.6 -5.661406517028809 -6.1707681936936
                                                -5.1520448403640176
    1.8 -6.044726371765137
                            -6.5506246416367935 -5.53882810189348
    2.0 -6.283880710601807
                            -6.787887088724079 -5.779874332479534
    2.25
            -6.401032447814941 \quad -6.9025169688225745 \quad -5.899547926807308
    2.5 -6.5673747062683105 -7.067221321513552
                                                -6.067528091023069
            -6.818793296813965
                               -7.3182220553598984 -6.319364538268031
    3.0 -6.978192329406738
                            -7.478020503324831
                                                -6.478364155488646
29
    3.25
            -7.162293910980225
                               -7.662787496390374 -6.661800325570075
    3.5 -7.231132984161377
                            -7.733609523677808 -6.728656444644946
            -7.508701801300049 -8.011055252541407 -7.0063483500586905
    3.75
    4.0 -7.671975135803223
                           -8.176972429713691
                                                -7.166977841892753
           -7.896575450897217 -8.404686456182528 -7.3884644456119055
34
    4.5 -7.711543560028076
                            -8.221055945399781
                                                -7.202031174656372
    4.75
            -7.842799663543701 -8.354660168020564
                                                    -7.33093915906684
                            -8.423843580153179
    5.0 -7.909265995025635
                                                -7.39468840989809
```

Figure 5: Estimated IM output in the "GGMM.out" file

Reference

Jawad Fayaz, Miguel Medalla, Pablo Torres, and Carmine Galasso (2023). "Recurrent Neural Networks based Generalized Ground Motion Model for Chilean Subduction Seismic Environment". *Structural Safety*. https://www.sciencedirect.com/science/article/pii/S0167473022000893?via%3Dihub.