

1 fctSPM: Factorial ANOVA and post-hoc tests for
2 Statistical Parametric Mapping in MATLAB

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6 **Summary**

7 Statistical Parametric Mapping (SPM) is a statistical method originally used in neuroimaging
8 developed in the early 90's in biomedical imaging, allowing to determine which brain zones
9 were solicited during a functional MRI (Friston et al., 1995). Originally developed for a three
10 dimensional analysis, the application of this method to the analysis of vectors or matrices was
11 made possible thanks to Pataky (2010) whom allows to perform statistical inference on curves
12 (vectors - 1D) or maps (matrices - 2D).

13 As in “classical” statistics on scalar values (0D), there is a parametric and a non-parametric
14 approach to the SPM method. While the parametric method is based on random gaussian
15 fields, the non-parametric method is based on label permutation tests (Nichols & Holmes,
16 2002), and thus, on re-sampling and randomness to make statistical inference. The main
17 advantage of the non-parametric approach is that a gaussian distribution of the data is not
18 required, making possible it to work with both curves and maps.

19 **Statement of need**

20 Most of physiological data measured during human movement are continuous and expressed
21 in function of time. However, researchers predominantly analyze extracted scalar values from
22 the continuous measurement, as the mean, the maximum, the amplitude, or the integrated
23 value over the time. Analyzing continuous values (i.e., time series) can provide more infor-
24 mation than extracted indicators, as the later discards one dimension of the data among the
25 magnitude and localization in time. In addition, oscillatory signals such as muscle vibrations
26 and electromyograms contain information in the temporal and frequency domains. However,
27 scalar analysis reduces the information at only one dimension by discarding two dimensions
28 among the magnitude and the localization in the time and/or frequency domain.

29 To analyze all the dimensions of a signal without losing information, the analysis of curves
30 or maps was proposed, coded, and put online by Pataky. However, the use of the proposed
31 functions does not allow the analysis of 2D data automatically. Moreover, a rather frequent
32 error is to consider only the significance of the last statistical test performed and not the
33 intersection between the post-hoc tests and the ANOVA. Indeed, a difference between two
34 samples can be significant if, and only if the ANOVA is significant in the same areas. This
35 package, integrated in the fctSPM repository at `./fctSPM/src/spm1d_Pataky` is [published](#)
36 [elsewhere](#), and thus is not part of this JOSS review.

37 fctSPM

38 The function we propose meets two objectives. 1/ to allow statistical inferences on curves
39 and maps with a standardized format and 2/ to simplify analyses by comparing means while
40 considering intersections with tests performed upstream (ANOVA and post-hoc of main ef-
41 fects).

42 The statistical tests are performed taking into account the independent and repeated measure
43 effects provided in the obligatory function inputs. ANOVA, up to three-way ANOVA with
44 three repeated measures, is performed if required, and followed by post-hoc tests as paired or
45 independent Student t-tests. By default, the ANOVA is performed with an alpha risk of 5%,
46 while post-hoc tests alpha risk is adjusted with Bonferroni correction. A number of 10/alpha
47 iterations (200 for a 5% risk) is defined for each test. Statistical parameters are customizable
48 via optional inputs, like `multiIT` which can be used to increase the number of iterations and
49 achieve numerical stability and reliable analysis (Nichols & Holmes, 2002). A Matlab (.mat)
50 file containing the number of permutations, the significant clusters, the statistical thresholds,
51 and the raw data used in the analysis is also generated for each test family.

52 To interpret the results, figures directly usable for presentations and/or articles are available.
53 In one dimension, the main figure contains the mean and standard deviations for each group
54 of the analyzed condition, and the results of the post-hoc tests corrected with the result of the
55 ANOVA. In two dimensions, the mean maps and the standard deviations are on two separate
56 figures, and the result of the statistical analysis is displayed on the map of differences between
57 two modalities. Therewith, other figures that display absolute and relative differences, effect
58 sizes, and the raw value of the statistical test and its threshold are available.

59 To personalize the figures, Matlab (.fig) files are implemented to perform a posteriori modifi-
60 cations, and optional inputs are available to a priori customize the figures. These parameters
61 are gathered in three categories: 1/ general plot parameters working identically for one and
62 two dimensions, acting on axis labels and limits, image resolution and size 2/ one dimension
63 parameters that act on the characteristics of the curves like color and transparency, or the
64 position of the statistical analysis relatively to curves 3/ two dimensions parameters acting
65 on the colormap and its limits, as well as the color of the statistical test displayed on map of
66 differences.

67 This function was already used in Trama et al. (2021) to compare soft-tissue and muscle vibra-
68 tions of the *vastus lateralis*. It is currently used to assess modifications of soft-tissue vibrations
69 caused by mountain ultra-marathons, the effect of the pedaling phase on *quadriceps* soft-tissue
70 vibrations, and differences in isokinetic torque after ACL operation and rehabilitation.

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