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The dark side of radiomics: on the paramount importance of publishing negative results

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Over the past few years, we have witnessed an exponential increase of the number of radiomic-related publications. In the PET literature, a PubMed search using “(radiomic OR radiomics OR texture OR textural) AND PET” search criterion yielded 37 publications in 2015 to 110 in 2018. Interestingly, an extensive survey of these publications demonstrates that 94% reported positive or “promising” results involving some sophisticated radiomic features, with large variations in the performance supporting that conclusion. As few as 6% of them clearly concluded at negative results, including the paper by Saadani et al (1) published in that issue of the Journal of Nuclear Medicine. These numbers highlight a publication bias well acknowledged in many different research fields, with an on-going debate about the actual false discovery rate in the medical literature (2-4). This publication bias has well understood roots. From the editors’ standpoint, positive results are often thought to be much more exciting and valuable than negative ones and are more likely to be cited hence favorably increase the journal influence metrics (eg, 2-year impact factor) (5). From the authors’ perspective, positive results are more rewarding than negative findings and might better contribute to boosting their careers. As a consequence, negative radiomic results, the dark side of radiomics, currently remain mostly unpublished.

Yet, publishing negative results is a must for ethical reasons. Criteria for publishing should be the quality of the study and its statistical power, whatever the outcome. A clinically or biologically relevant question and a methodologically well-designed study should warrant publication, no matter if the null hypothesis is rejected or not. In some instances, negative studies might be even more impactful than positive ones, as they may challenge existing paradigms and invite investigators to focus research efforts on different paths. The conclusion of Saadani et al paper (1) is that although BRAF mutation drives the MAPK pathway and glucose metabolism in some cancers, BRAFV600 status could not be successfully detected using radiomic features calculated from 18F-FDG PET/CT in melanoma patients. This observation should be an incentive to further explore the connection between the genetic mutations or pathway alterations and their macroscopic consequences detectable using our in vivo imaging devices. It is reasonable to expect that some biological dysfunctions resulting

from mutations will produce a cascade of events that might ultimately yield a signal detectable by our exquisite molecular imaging scanners. However, both the magnitude and the spatio-temporal extent of the biological effect will determine our ability to identify an abnormality from in vivo images using a given radiotracer. Investigations of the relationships between the triggering signal (here a mutation) and the net observable result (here change in tumor glucose metabolism) are absolutely needed for two reasons: first to establish realistic expectations regarding the potential power of radiomic features, second to use radiomic observations as a driver to formulate more precise biological assumptions regarding the underlying processes and subsequently test them. Advancing that field will require the publication of both positive and negative radiomic results.

One could argue that given the overwhelming number of radiomic-related publications reporting positive results, which might be the trees that hide the forest, the publication of negative results in that domain will be practically challenging. What we need are methodologically sound, properly powered and robust radiomic studies addressing a biologically-driven hypothesis and described in such a way that independent investigators can reproduce and confirm the findings. Indeed, replication has often more scientific value than original discovery in the radiomic field. To date, to the best of our knowledge, no PET radiomic model has ever been validated by an independent group since 2014. Even worse, the misleading interpretation of some of the most famous radiomic findings (6) has recently been demonstrated (7), confirming that some radiomic features initially interpreted as biomarkers of tumor heterogeneity were actually surrogates of the tumor volume, as shown as early as 2014 (8-9). These findings suggest that radiomic-related publications should be thoroughly designed with sufficient statistical power, described in such a way that they can be repeated, and that they should also include well-supported interpretation possibly based on dedicated experiments.

Among the methodological arsenal that can be used to chase spurious and confounding effects in radiomic studies, “sham” data and permutation/randomization tests could be used

more often. “Sham” data can be obtained from healthy regions or by reshuffling samples in artificial groups or randomly changing voxel values (7). Such sham data are useful to check that the findings are present in the real data and not in the sham data. This is thanks to such sham data that the erroneous interpretation of the radiomic features highlighted in (6) could be nicely demonstrated (7). Randomization/permutation tests also yield a precise estimate of the distribution of a test statistic under the null hypothesis in a non-parametric setting by calculating all possible values of the test statistic under rearrangements of the labels on the observed data points. They are frequently used in genomic data analysis to assign a statistical significance to the findings (10). Last, statistical reviewing should be systematic as it is effective in improving the quality of published articles.

The validity and soundness of radiomic studies also depend on how the “vibration of effects” is handled. Vibration of effects means that results can differ over a broad range depending on how the analysis is performed. This is observed when many variations of analyses are used, for instance selecting different variables, making statistical adjustments, testing different statistical models, sorting the data differently or using different inclusion and exclusion criteria. The variability of results as a function of the chosen analysis strategy gives an indication about their robustness. Reporting only the analyses that yield a particular result, called the chrysalis effect (11), can yield a bias, while consistent conclusions across various methodological approaches, as reported in Saadani et al paper (1), provide a persuasive indication of the correctness of the findings.

Radiomic results should be interpreted with great circumspection. Radiomic observations are easy to report given the wide availability of software enabling the calculation of radiomic features and of packages supporting sophisticated statistical analysis and data mining. Yet, radiomics should now go beyond reporting associations and designing predictive models. It should explain what the observed patterns and associations mean from a biological point of view, so as to enhance our understanding of the mechanisms. Dedicated experiments and multi-omics approaches are needed to fully take advantage of the radiomic approach, so that

the radiomic phenotype and associated models can be additional exploratory tools to untangle the complexity of the biological processes and of their macroscopic repercussions.

A more stringent selection of radiomic studies should thus be based on tougher editorial standards. More skepticism on the part of referees is needed to avoid the inflation of false positive findings or overzealous interpretation of radiomic results. In that respect, the five aspects discussed above might be systematically considered: biological rationale justifying the study, ability to independently replicate and reproduce the findings, comprehensive control of false discoveries validated by systematic statistical reviewing, investigation of the vibration of the effects, proofs supporting the interpretation of the results. Careful selection of publishable manuscripts based on these criteria will then leave room for publishing radiomic studies yielding negative results and confirmatory radiomic studies, which are both absolutely needed. Dedicated sections in top-tier journals could actually be devoted to important negative and confirmatory studies. Still, it will remain virtually impossible to publish all negative radiomic findings that would be useful to avoid unnecessary and costly repeats of experiments already done. Several practical solutions can be suggested. First, studies could be registered before they start in international registries (radiomic trial registries), so that it would be easier to determine whether a given question has already been or is being addressed by any investigator before investing in that research. Second, negative results could be made accessible in public repositories, such as arxiv.org, and referred to in review papers to facilitate their identification. Last, funding agencies should request the registration of the funded radiomic studies and the access to the results including the negative findings. All these actions together could considerably reduce the number of weekly contributive radiomic studies, clarify the state of the art, balance positive against negative radiomic results, and contribute to a proper identification of radiomic models that are useful either for patient care or for advancing our knowledge regarding how in vivo imaging can probe microscopic biological mechanisms. In addition, establishing and enforcing best practices should help us get continuous support from external stakeholders that provide funding to advance sustainable research in radiomics.

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The authors of this manuscript declare no relationships with any companies, whose products or services may be related to the subject matter of the article.

REFERENCES

1. Saadani H et al. Metabolic biomarker based BRAFV600 mutation association and prediction in melanoma, *J Nucl Med.* 2019, In press.
2. Ioannidis JPA. Why most published research findings are false. *PLoS Medicine.* 2005;2:e124.
3. Jagger LR, Leek JT. An estimate of the science-wise false discovery rate and application to the top medical literature. *Biostatistics.* 2014;15:1-12.
4. Chalkidou A, O'Doherty MJ, Marsden PK. False discovery rates in PET and CT studies with texture features: a systematic review. *PLoS One.* 2015;10:e0124165.
5. Duyx B, Urlings MJE, Swaen GHM, Bouter LM, Zeegers MP. Scientific citations favor positive results: a systematic review and meta-analysis. *J Clin Epidemiol.* 2017;88:92-101.
6. Aerts HJWL et al. Decoding tumour phenotype by noninvasive imaging using a quantitative radiomics approach. *Nat Commun.* 2014;5:4006.
7. Welch ML et al. Vulnerabilities of radiomic signature development: The need for safeguards. *Radiother Oncol.* 2019;130:2-9.
8. Orlhac F, Soussan M, Maisonneuve JA, Garcia CA, Vanderlinden B, Buvat I. Tumor texture analysis in 18F-FDG PET: relationships between texture parameters, histogram indices, standardized uptake values, metabolic volumes, and total lesion glycolysis. *J Nucl Med.* 2014;55:414-22.
9. Buvat I, Orlhac F, Soussan M. Tumor texture analysis in PET: where do we stand? *J Nucl Med.* 2015;56:1642-4.
10. Dudoit S, Popper Shaffer J, Boldrick JC. Multiple hypothesis testing in microarray experiments. *Statist Sci.* 2003;1:71-103.
11. O'Boyle EH, Banks GC, Gonzales-Mulé E. The chrysalis effect: how ugly initial results metamorphosize into beautiful articles. *J Management.* 2014;43:376-399.