Michael Biehl - Publications in reverse chronological order

[1] Alessandro Prete, Katharina Lang, David Pavlov, Yara Rhayem, Alice J Sitch, Anna S Franke, Lorna C Gilligan, Cedric HL Shackleton, Stefanie Hahner, Marcus Quinkler, et al. Urine steroid metabolomics as a diagnostic tool in primary aldosteronism. *The Journal of Steroid Biochemistry and Molecular Biology*, pages Article No. 106445,10 pages, 2023.

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[5] Thais P Rocha, Eka Melson, Roland J Veen, Lida Abdi, Tara McDonnell, Veronika Tandl, James Hawley, Laura Wittemans, Amarah Anthony, Lorna Gilligan, et al. Abstract or20-01 machine learning-based steroid metabolome analysis in women with polycystic ovary syndrome reveals three distinct androgen excess subtypes with different metabolic risk profiles. *Journal of the Endocrine Society*, 7(Supplement 1):bvad114–1653, 2023.

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with polycystic ovary syndrome-results from the prospective daisy-pcos study. In *Endocrine Abstracts*, volume 94. 2023.

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