UNIVERSIDAD DE SANTIAGO DE COMPOSTELA

MASTER'S THESIS

Predictive model for drug response based on multi-omics patterns

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Abstract

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Menyhárt and Győrffy (2021), Eraslan et al. (2019), Chaudhary et al. (2018), Chakraborty et al. (2018), Vasaikar et al. (2018), Hasin et al. (2017), Kourou et al. (2015), Libbrecht and Noble (2015), Meng et al. (2016), Misra et al. (2019), Cai et al. (2022), Sammut et al. (2022), Topol (2019). The page is kept centered vertically so can expand into the blank space above the title too...

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