Al-Assisted Tool to Improve the Quality and Assessment of PLGA Formulations

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RESULTS **SUMMARY** 21 0.0380.023 0.45 0.31 0.96 0.94 0.71 0.76 -0.8 0.81 The AI-based method provides high-throughput analysis of PLGA (poly 2 -0.21-0.047 0.11 0.31 0.14 -0.69 -0.77 -0.64 -0.6 0.62 -0.57 (lactic-co-glycolic acid))-based long-acting injectable (LAI) formulations to - 0.75 63-0.024 0.33 -0.59 -0.19 0.76 0.71 0.31 0.45 -0.85 0.66 -0.22 Surface area ratio - 0.84 60.17 0.029 0.49 -0.55 -0.45 -0.51 -0.52 0.2 -0.62 Manufacturing Method - -0.46 establish a correlation between material attributes, processing -0.54 - 0.50 011-0.22 0.24 0.27 0.22 0.25 -0.1 0.27 L:G - 0.21 -0.21 0.063 -0.54 1 -1 e-1-0.22 conditions, and product quality/performance. Solvent -0.0380.047-0.024 2e-16.4e-16 1 -0.0860.0486 3e-170.014 0.051 0.066 0.12 0.0061 0.1





- 0.00

- -0.25

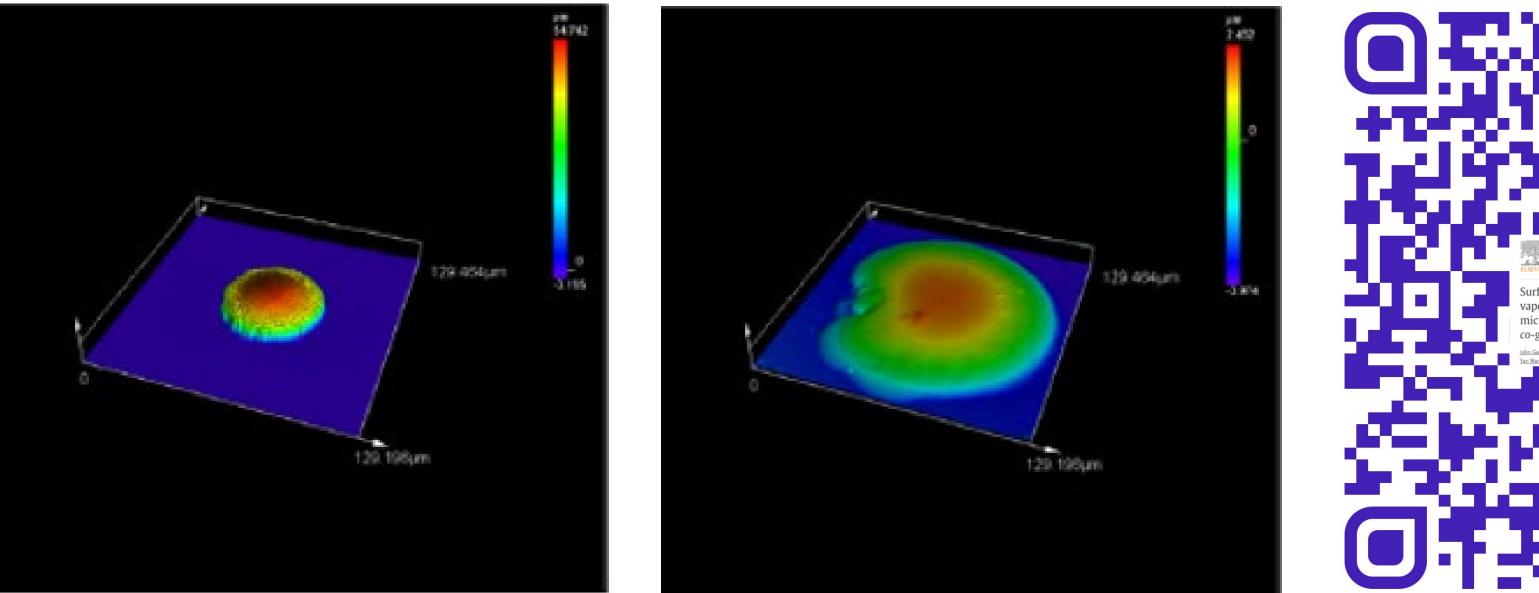
- -0.50

- -0.75

This AI method may serve as a tool in the future to evaluate the sameness of proposed generic products to reference listed drugs (RLD) by analyzing feature similarity across different formulations.

METHODS

- The study has compiled a comprehensive dataset of PLGA formulations from a previous FDA-funded research project [1].
- The dataset includes different formulation details, corresponding manufacturing data, detailed surface topographical characterization data due to solvent changes, and *in vitro* release testing data [1].





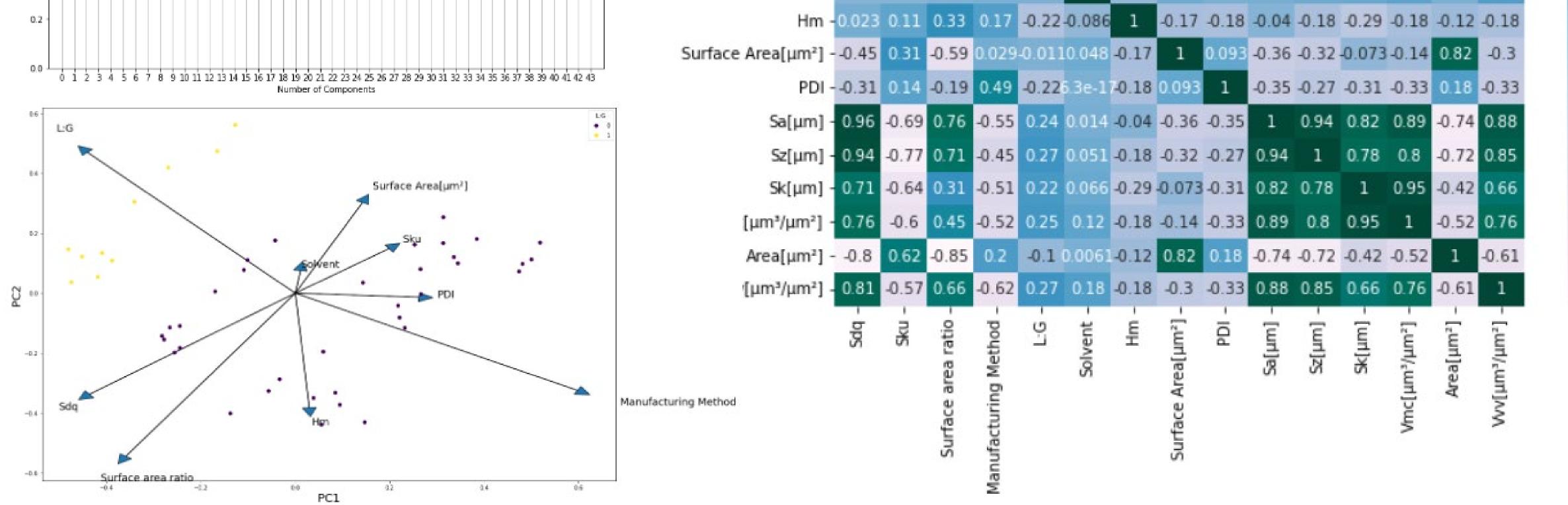


Fig 3: Principal component analysis (PCA) on the left and correlation matrix study on the right to find significant parameters

Accuracy (%)

99.5002

100

90

99.2778

(A) Prediction of Formulation Conditions

MAE

(B) Prediction of L:G Ratios from PLGA Formulations

Machine Learning MSE Techniques

Formulation

After exposure to semi-solventQR Code of Previousliquid to FormulationWork

Linear Regression0.0010.02Decision Tree00usRandom Forest0.34390.4Extra Trees
Regressor0.00420.0289

Fig 1: Microparticles of formulations in the dry state and after exposure to a semisolvent liquid at 0°C for 1 min and details of previous work in the QR code [1]

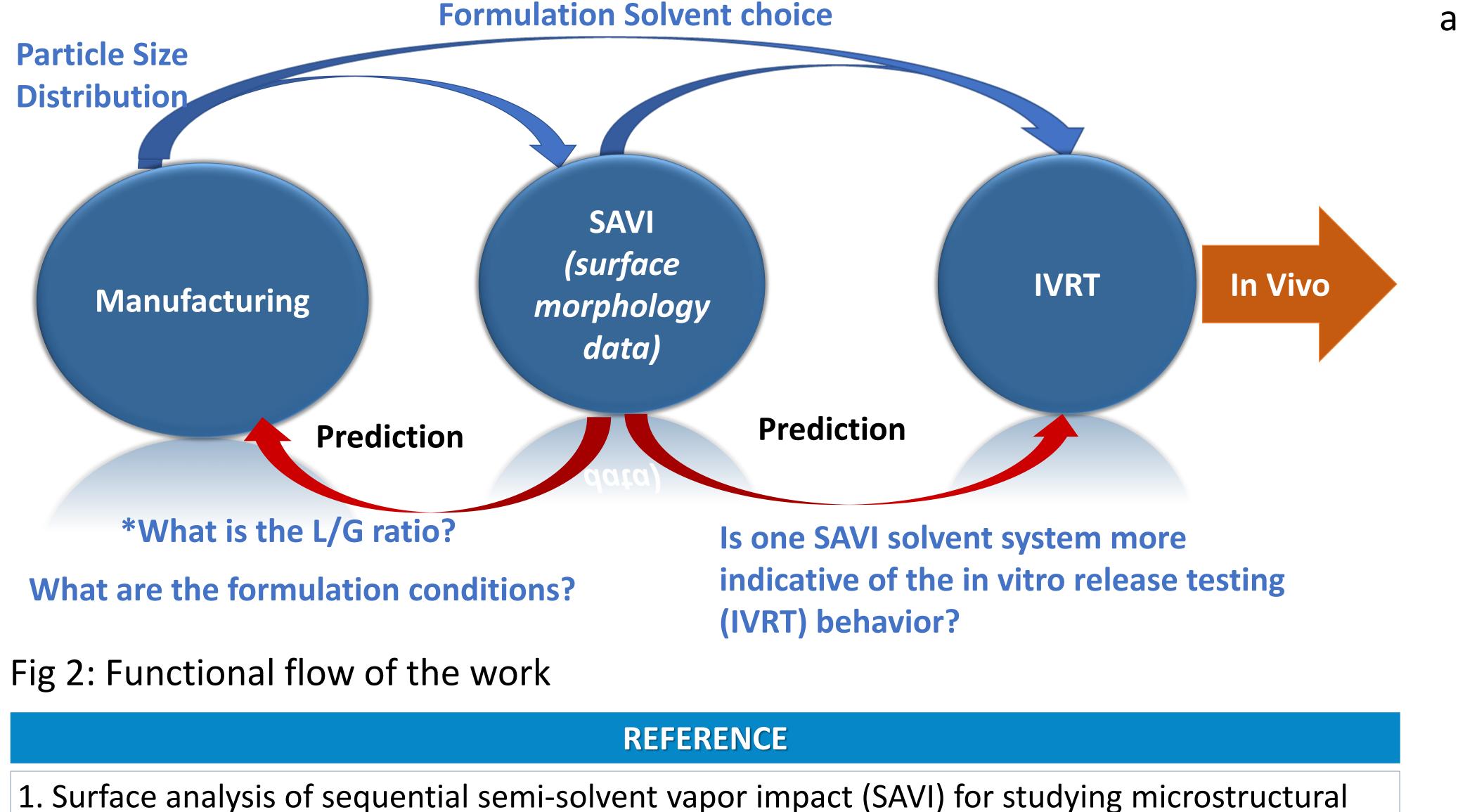
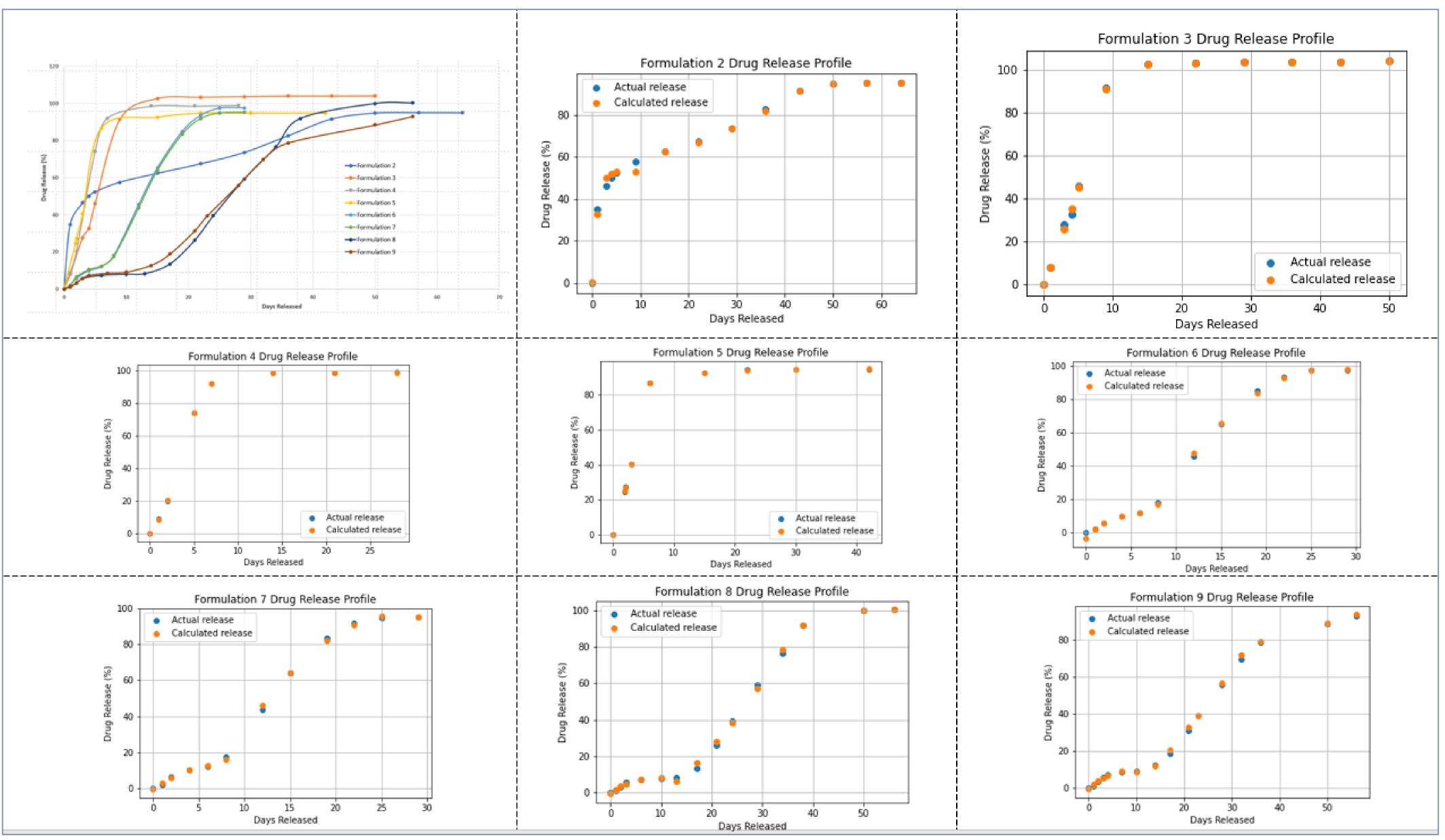


Table 1: Different machine learning (ML) algorithms were applied (A) to predict the formulation conditions and (B) to predict L:G ratios



arrangements of poly (lactide-co-glycolide) microparticles; John Garner, Sarah Skidmore, Justin Hadar, Haesum Park, Kinam Park, Bin Qin, Yan Wang; Journal of Controlled Release, 2022, 350, 600-612

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Fig 4: Prediction of IVRT profiles and compared with experimental release profiles

DISCLAIMER

The views expressed in this work reflect the views of authors and do not reflect the official policies of the U.S. Food and Drug Administration or the U.S. Department of Health and Human Services.