Data analysis in QSAR

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Also QSPR (Property)



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Optimise model on training set (2/3) Test model on test set (1/3) Need to use an estimate of the error of prediction (CV or bootstrap, but **not** resubstitution)

or

Build model on training set (2/3 of 2/3) Optimise on test set (1/3 of 2/3) Test model on validation set (1/3)

Using the error of prediction

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Using the error of prediction

Testing must be carried out on a hold-out set from the same distribution as the the set of molecules used for optimisation => don't optimise model on diverse set

Assessing Model Fit by Cross-Validation, JCICS, 2003, 43, 579. Beware of q², J.Mol.Graph.Model., 2002, 20, 269.

Feature selection



2ⁿ different subsets

Parameter optimisation

- Models have parameters which should be tuned
 - Support vector machines
 - gamma, cost, epsilon



Penalty of error: distance to hyperplane multiplied by error cost C.

QSAR: Feature selection and parameter optimisation

- **Aim**: To find a robust method of simultaneously optimising the parameters and performing feature selection
- Prediction of solubility for drug-like molecules
 - David Palmer
 - Support vector machines
 - gamma, cost, epsilon
 - 127 descriptors
- Large search space
 - stochastic algorithm required

Ant Colony Optimisation (ACO)

- Insipired by behaviour of ants foraging for food
- Ants lay down pheromones, which influence the path taken by other ants. Meanwhile pheromones are evaporating.
- Ants' trails converge to shortest path between nest and food



- Ant Colony Optimisation (ACO) Marco Dorigo, PhD Thesis, 1992.
- ACO for feature selection Shen et al, JCIM, 2005, 45, 1024.
- We have extended it to perform simultaneous parameter optimisation

Ant Colony Optimisation (ACO)

- population of ants (typically 50 to 100)
 - each ant represents a model i.e. a subset of descriptors and values for the parameters
 - each ant has a fitness score, e.g. 10-fold cross validation rmse
- it is more likely that an ant will choose a particular descriptor/parameter value in the next iteration if
 - many ants have chosen it in this iteration (local search), or
 - many ants have chosen it in their best models to date (global search)
- for descriptors/parameter values that are not chosen in the current or best models, the probability that they will be chosen decreases (evaporation)

Does it work?

