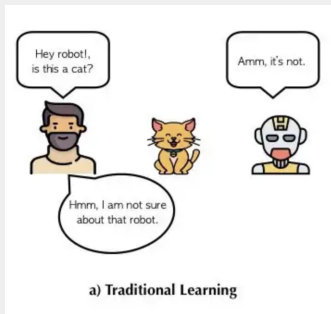


# QUALITY DIVERSITY OPTIMISATION FOR POST-HOC ENSEMBLING

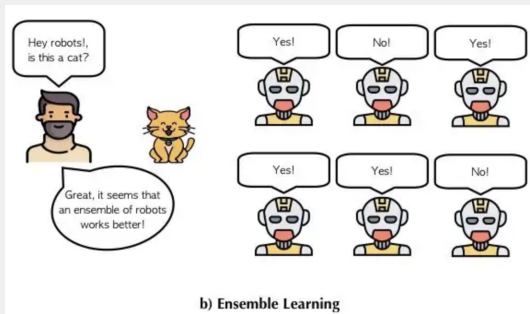
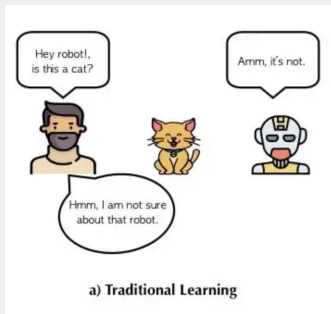
**M. ANASTACIO, L. PURUCKER, L. SCHNEIDER ET AL.**

DEC, 2022

# ENSEMBLING



# ENSEMBLING



source: <https://towardsdatascience.com>

# POST-HOC ENSEMBLING

- Ensemble learning
  - ▶ build ensemble and learn models simultaneously
  - ▶ ex: random forest

# POST-HOC ENSEMBLING

## ■ Ensemble learning

- ▶ build ensemble and learn models simultaneously
- ▶ ex: random forest

## ■ Post-hoc ensembling

- ▶ optimise the models
- ▶ build ensemble from bests
- ▶ ex: autosklearn, autogluon, ...

# CURRENT METHODS

- Single best model

# CURRENT METHODS

- Single best model
- Majority voting

# CURRENT METHODS

- Single best model
- Majority voting
- Ensemble voting



# CURRENT METHODS

- Single best model
- Majority voting
- Ensemble voting
- Greedy ensemble selection

# QUALITY DIVERSITY OPTIMISATION

## Quality

- Balanced accuracy
- ROC AUC

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- Balanced accuracy
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## Diversity

- Loss correlation
- Gower distance

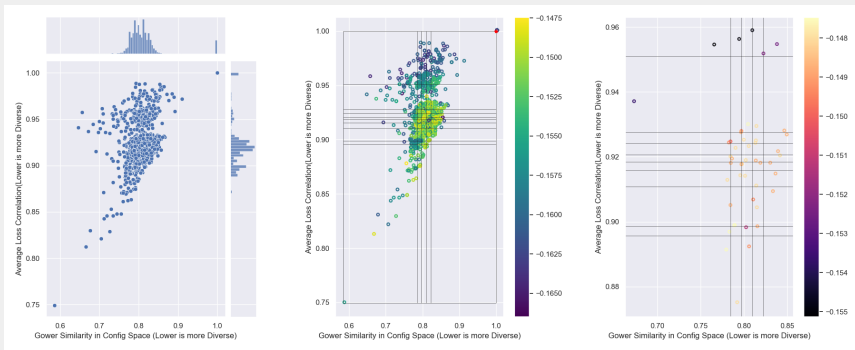
# QUALITY DIVERSITY OPTIMISATION

## Quality

- Balanced accuracy
- ROC AUC

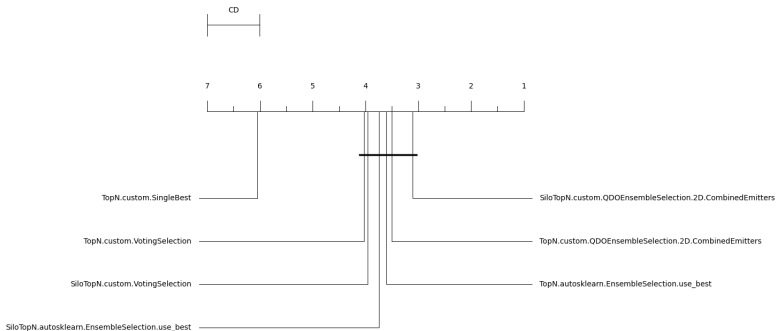
## Diversity

- Loss correlation
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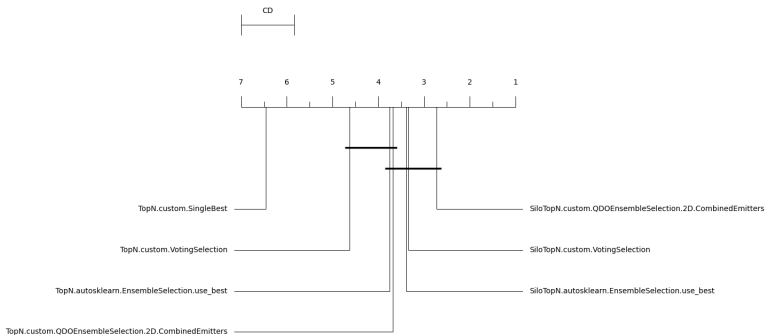
# QDO ENSEMBLING

Autorank Plot | CD Plot for Task Type Binary Classification; Results for all metrics concatenated



# QDO ENSEMBLING

Autorank Plot | CD Plot for Task Type Multi-class Classification; Results for all metrics concatenated



# FUTURE WORK

Promising first results

# FUTURE WORK

Promising first results but:

- Define the bins differently



# FUTURE WORK

Promising first results but:

- Define the bins differently
- Add mutations and crossover

# FUTURE WORK

Promising first results but:

- Define the bins differently
- Add mutations and crossover
- Define our initial population differently