KNN → K - Nearest Neighbours

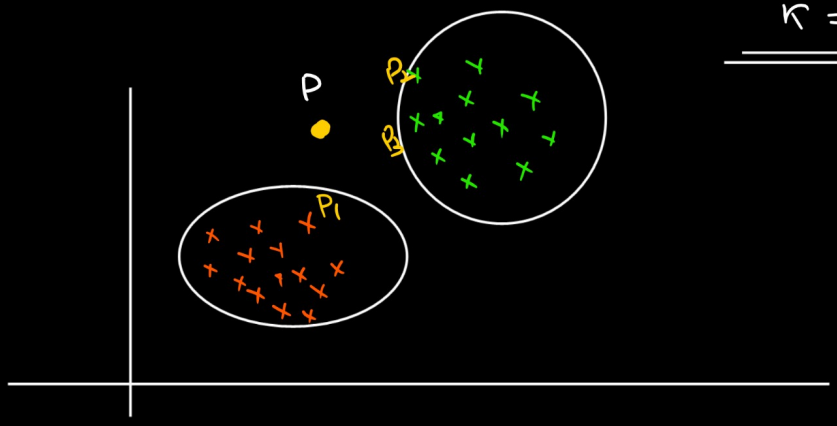
↓

Supervised ML — classification

Regression

K → # neighbours

K = 3



Distance metric → Euclidean distance, Manhattan distance, Hamming distance

Continuous values

P_1 — P = 4

P_2 — P = 3

P_3 — P = 1

P_4 — P = 7

P_5 — P = 10

P_6 — P = 15

P_7 — P = 12

classification

Majority voting → P → green points

Regression → average / mean

Task → KNN → K → user → 5 / 3

1. Heap

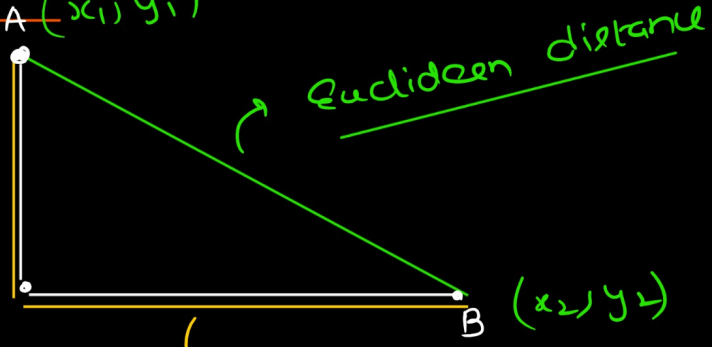
2. ① distance
   ② sort
   ③ first k values → Points → Predict

Euclidean Distance → $\sqrt{(x_1 - x)^2 + (y_1 - y)^2}$ → More sensitive to outliers

L2 Norm

Manhattan Distance → $|x_1 - x| + |y_1 - y|$

L1 Norm



Hamming distance → string

string → A T C G (how many positions differ)

string → A T G G

1 → 1

Quantities genetic info binary vector

1 0 1 1 1 0 0

1 0 0 1 0 0 0

1 + 1 → 2

KNN → Lazy learner

↳ limitation → huge amount of data

$y = mx + c$

↳ optimised value of m & c

- Train → it does nothing
- Prediction → distance, sort & get K nearest neighbours

Real time usage →

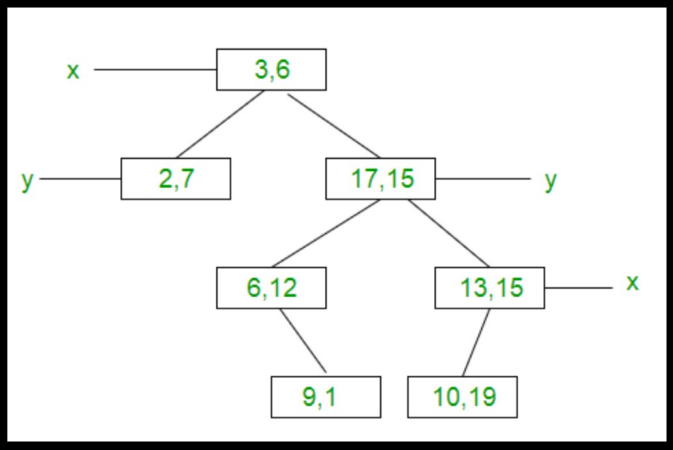
↳ Prediction

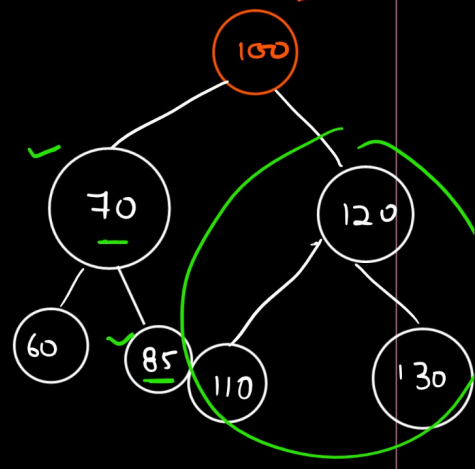
Static → imputation

↳ SMOTE — Task (2)

Optimisation in Tree

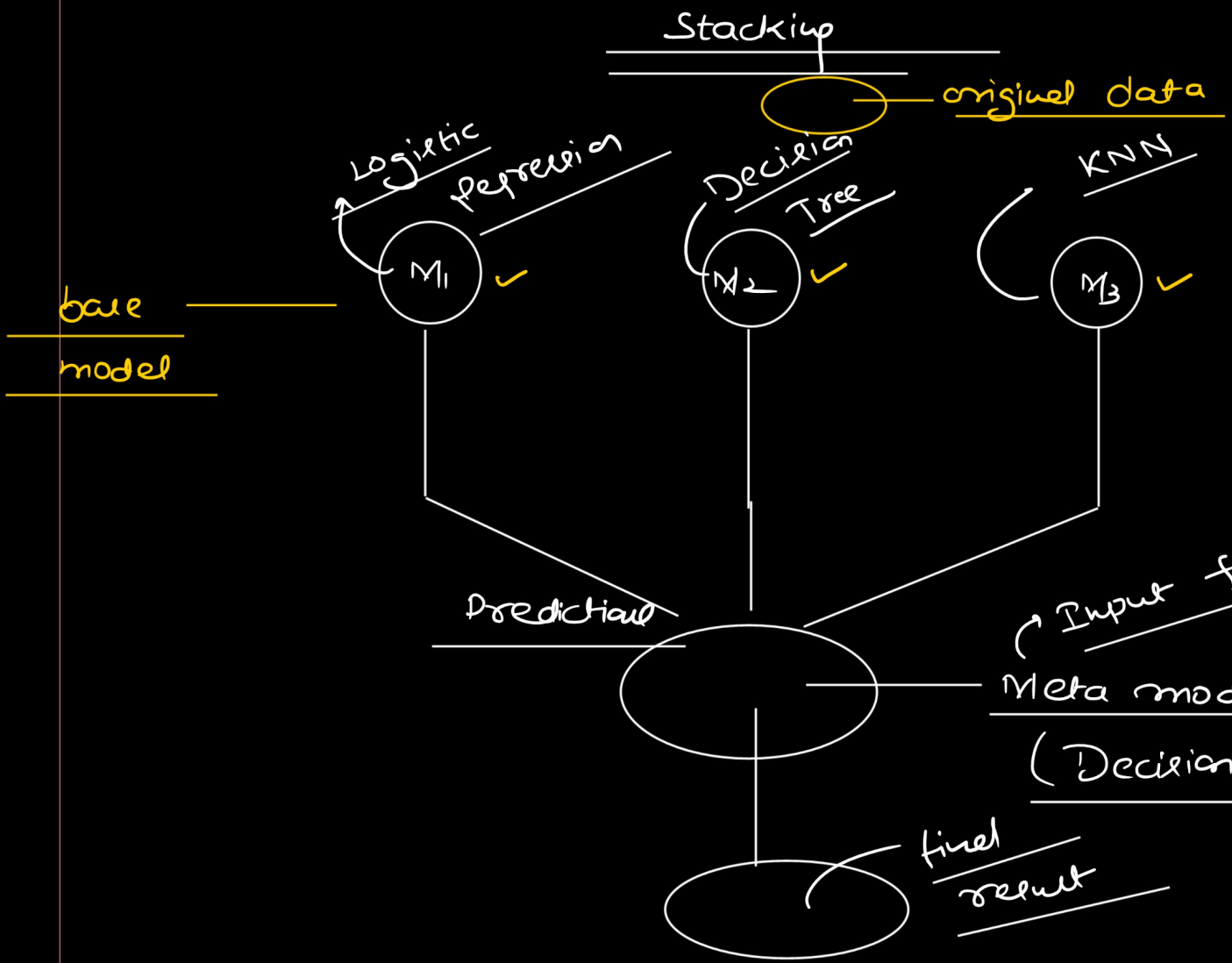
- BST (Binary Search Tree)
  - AVL Tree
  - Red Black Tree
  - Kd Tree
  - ball_tree





xx

Stacking



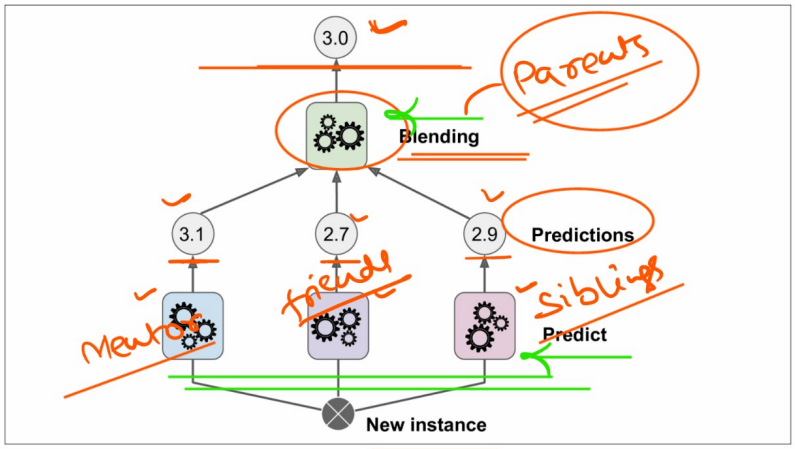


Figure 7-12. Aggregating predictions using a blending predictor