The impact of class imbalance in classification performance metrics based on the binary confusion matrix.

Supplementary material

- 1. Derivation of $\mu = \mu(\lambda_{PP}, \lambda_{NN}, \pi_P, \pi_N)$.
- 1. Sensitivity (SNS).

$$SNS = TPR \equiv \frac{TP}{TP + FN} = \frac{m_{PP}}{m_P} = \frac{\lambda_{PP} m_P}{m_P} = \lambda_{PP}.$$
 (1)

2. Specificity (SPC).

$$SPC = TNR \equiv \frac{TN}{TN + FP} = \frac{m_{NN}}{m_N} = \frac{\lambda_{NN} m_N}{m_N} = \lambda_{NN}. \tag{2}$$

3. Precision (PRC).

$$PRC = PPV \equiv \frac{TP}{TP + FP} = \frac{m_{PP}}{e_P} = \frac{\lambda_{PP} m_P}{\lambda_{PP} m_P + \lambda_{NP} m_N} = \frac{\lambda_{PP}}{\lambda_{PP} + \lambda_{NP} \frac{m_N}{m_P}}.$$
(3)

$$PRC = \frac{\lambda_{PP}}{\lambda_{PP} + \lambda_{NP} \frac{\pi_N}{\pi_P}} = \frac{\lambda_{PP} \pi_P}{\lambda_{PP} \pi_P + \lambda_{NP} \pi_N} = \frac{\lambda_{PP} \pi_P}{\lambda_{PP} \pi_P + (1 - \lambda_{NN}) \pi_N}.$$
(4)

4. Negative Predictive Value (NPV).

$$NPV \equiv \frac{TN}{TN + FN} = \frac{m_{NN}}{e_N} = \frac{\lambda_{NN} m_N}{\lambda_{NN} m_N + \lambda_{PN} m_P} = \frac{\lambda_{NN}}{\lambda_{NN} + \lambda_{PN} \frac{m_P}{m_N}}.$$
 (5)

$$NPV = \frac{\lambda_{NN}}{\lambda_{NN} + \lambda_{PN} \frac{\pi_P}{\pi_N}} = \frac{\lambda_{NN} \pi_N}{\lambda_{NN} \pi_N + (1 - \lambda_{PP}) \pi_P}.$$
 (6)

5. Accuracy (ACC).

$$ACC \equiv \frac{TP + TN}{TP + FN + TN + FP} = \frac{m_{PP} + m_{NN}}{m}.$$
 (7)

$$ACC = \frac{\lambda_{PP} m_P + \lambda_{NN} m_N}{m} = \lambda_{PP} \pi_P + \lambda_{NN} \pi_N. \tag{8}$$

6. F_1 score.

$$F_{1} \equiv 2 \frac{PRC \cdot SNS}{PRC + SNS} = 2 \frac{\frac{\lambda_{PP} \pi_{P}}{\lambda_{PP} \pi_{P} + (1 - \lambda_{NN}) \pi_{N}} \cdot \lambda_{PP}}{\frac{\lambda_{PP} \pi_{P}}{\lambda_{PP} \pi_{P} + (1 - \lambda_{NN}) \pi_{N}} + \lambda_{PP}} = 2 \frac{\frac{\lambda_{PP} \pi_{P}}{\lambda_{PP} \pi_{P} + (1 - \lambda_{NN}) \pi_{N}}}{\frac{\pi_{P}}{\lambda_{PP} \pi_{P} + (1 - \lambda_{NN}) \pi_{N}} + 1}.$$
 (9)

$$F_{1} = 2 \frac{\frac{\lambda_{PP}\pi_{P}}{\lambda_{PP}\pi_{P} + (1 - \lambda_{NN})\pi_{N}}}{\frac{\pi_{P} + \lambda_{PP}\pi_{P} + (1 - \lambda_{NN})\pi_{N}}{\lambda_{PP}\pi_{P} + (1 - \lambda_{NN})\pi_{N}}} = 2 \frac{\lambda_{PP}\pi_{P}}{\pi_{P} + \lambda_{PP}\pi_{P} + (1 - \lambda_{NN})\pi_{N}}.$$
(10)

$$F_1 = 2 \frac{\lambda_{PP} \pi_P}{(1 + \lambda_{PP}) \pi_P + (1 - \lambda_{NN}) \pi_N}.$$
 (11)

7. Geometric Mean (GM).

$$GM \equiv \sqrt{SNS \cdot SPC} = \sqrt{\lambda_{PP} \cdot \lambda_{NN}}.$$
 (12)

8. Matthews Correlation Coefficient (*MCC*).

$$MCC = \frac{TP \cdot TN - FP \cdot FN}{\sqrt{(TP + FP)(TP + FN)(TN + FP)(TN + FN)}}.$$
(13)

$$MCC = \frac{\lambda_{PP} m_P \cdot \lambda_{NN} m_N - \lambda_{NP} m_N \cdot \lambda_{PN} m_P}{\sqrt{(\lambda_{PP} m_P + \lambda_{NP} m_N)(\lambda_{PP} m_P + \lambda_{PN} m_P)(\lambda_{NN} m_N + \lambda_{NP} m_N)(\lambda_{NN} m_N + \lambda_{PN} m_P)}}.$$
(14)

$$MCC = \frac{m_P m_N (\lambda_{PP} \lambda_{NN} - \lambda_{NP} \lambda_{PN})}{\sqrt{(\lambda_{PP} m_P + \lambda_{NP} m_N)(\lambda_{PP} m_P + \lambda_{PN} m_P)(\lambda_{NN} m_N + \lambda_{NP} m_N)(\lambda_{NN} m_N + \lambda_{PN} m_P)}}.$$
(15)

$$MCC = \frac{\lambda_{PP}\lambda_{NN} - \lambda_{NP}\lambda_{PN}}{\sqrt{\frac{(\lambda_{PP}m_P + \lambda_{NP}m_N)(\lambda_{PP}m_P + \lambda_{PN}m_P)(\lambda_{NN}m_N + \lambda_{NP}m_N)(\lambda_{NN}m_N + \lambda_{PN}m_P)}{m_Pm_N \cdot m_Pm_N}}.$$
(16)

$$MCC = \frac{\lambda_{PP}\lambda_{NN} - \lambda_{NP}\lambda_{PN}}{\sqrt{\frac{(\lambda_{PP}m_P + \lambda_{PN}m_P)}{m_P} \cdot \frac{(\lambda_{NN}m_N + \lambda_{NP}m_N)}{m_N} \cdot \frac{(\lambda_{PP}m_P + \lambda_{NP}m_N)}{m_P} \cdot \frac{(\lambda_{NN}m_N + \lambda_{PN}m_P)}{m_N}}}.$$
 (17)

$$MCC = \frac{\lambda_{PP}\lambda_{NN} - \lambda_{NP}\lambda_{PN}}{\sqrt{(\lambda_{PP} + \lambda_{PN})(\lambda_{NN} + \lambda_{NP})\left(\lambda_{PP} + \lambda_{NP}\frac{m_N}{m_P}\right)\left(\lambda_{NN} + \lambda_{PN}\frac{m_P}{m_N}\right)}}.$$
(18)

$$MCC = \frac{\lambda_{PP}\lambda_{NN} - (1 - \lambda_{NN})(1 - \lambda_{PP})}{\sqrt{1 \cdot 1 \cdot \left[\lambda_{PP} + (1 - \lambda_{NN})\frac{\pi_N}{\pi_P}\right] \left[\lambda_{NN} + (1 - \lambda_{PP})\frac{\pi_P}{\pi_N}\right]}}.$$
(19)

$$MCC = \frac{\lambda_{PP}\lambda_{NN} - (1 - \lambda_{PP} - \lambda_{NN} + \lambda_{PP}\lambda_{NN})}{\sqrt{\left[\lambda_{PP} + (1 - \lambda_{NN})\frac{\pi_N}{\pi_P}\right]\left[\lambda_{NN} + (1 - \lambda_{PP})\frac{\pi_P}{\pi_N}\right]}}$$
(20)

$$MCC = \frac{\lambda_{PP} + \lambda_{NN} - 1}{\sqrt{\left[\lambda_{PP} + (1 - \lambda_{NN})\frac{\pi_N}{\pi_P}\right] \left[\lambda_{NN} + (1 - \lambda_{PP})\frac{\pi_P}{\pi_N}\right]}}$$
(21)

9. Bookmaker Informedness (*BM*).

$$BM \equiv SNS + SPC - 1 = \lambda_{PP} + \lambda_{NN} - 1. \tag{22}$$

10. Markedness (*MK*).

$$MK \equiv PPV + NPV - 1 = \frac{\lambda_{PP}\pi_P}{\lambda_{PP}\pi_P + (1 - \lambda_{NN})\pi_N} + \frac{\lambda_{NN}\pi_N}{\lambda_{NN}\pi_N + (1 - \lambda_{PP})\pi_P} - 1.$$
 (23)

$$MK = \frac{\pi_P}{\pi_P + \frac{(1 - \lambda_{NN})}{\lambda_{PP}} \pi_N} + \frac{\pi_N}{\pi_N + \frac{(1 - \lambda_{PP})}{\lambda_{NN}} \pi_P} - 1.$$
 (24)

- 2. Derivation of $\mu = \mu(\lambda_{PP}, \lambda_{NN})$ when the classes are balanced.
- 1. Sensitivity (SNS).

$$SNS = \lambda_{PP}. \tag{25}$$

2. Specificity (SPC).

$$SPC = \lambda_{NN}.$$
 (26)

3. Precision (PRC).

$$PRC = \frac{\lambda_{PP}\pi_{P}}{\lambda_{PP}\pi_{P} + (1 - \lambda_{NN})\pi_{N}} = \frac{\lambda_{PP} \cdot \frac{1}{2}}{\lambda_{PP} \cdot \frac{1}{2} + (1 - \lambda_{NN}) \cdot \frac{1}{2}} = \frac{\lambda_{PP}}{\lambda_{PP} + (1 - \lambda_{NN})}.$$
 (27)

4. Negative Predictive Value (NPV).

$$NPV = \frac{\lambda_{NN}\pi_{N}}{\lambda_{NN}\pi_{N} + (1 - \lambda_{PP})\pi_{P}} = \frac{\lambda_{NN} \cdot \frac{1}{2}}{\lambda_{NN} \cdot \frac{1}{2} + (1 - \lambda_{PP}) \cdot \frac{1}{2}} = \frac{\lambda_{NN}}{\lambda_{NN} + (1 - \lambda_{PP})}.$$
 (28)

5. Accuracy (ACC).

$$ACC = \lambda_{PP}\pi_P + \lambda_{NN}\pi_N = \lambda_{PP} \cdot \frac{1}{2} + \lambda_{NN} \cdot \frac{1}{2} = \frac{1}{2}(\lambda_{PP} + \lambda_{NN}). \tag{29}$$

6. F_1 score.

$$F_1 = 2 \frac{\lambda_{PP} \cdot \frac{1}{2}}{(1 + \lambda_{PP}) \cdot \frac{1}{2} + (1 - \lambda_{NN}) \cdot \frac{1}{2}} = \frac{2\lambda_{PP}}{(1 + \lambda_{PP}) + (1 - \lambda_{NN})}.$$
 (30)

7. Geometric Mean (*GM*).

$$GM = \sqrt{\lambda_{PP} \cdot \lambda_{NN}}. (31)$$

8. Matthews Correlation Coefficient (*MCC*).

$$MCC = \frac{\lambda_{PP} + \lambda_{NN} - 1}{\sqrt{\left[\lambda_{PP} + (1 - \lambda_{NN})\frac{\pi_N}{\pi_P}\right]\left[\lambda_{NN} + (1 - \lambda_{PP})\frac{\pi_P}{\pi_N}\right]}}$$
(32)

$$MCC = \frac{\lambda_{PP} + \lambda_{NN} - 1}{\sqrt{\left[\lambda_{PP} + (1 - \lambda_{NN})\frac{0.5}{0.5}\right] \left[\lambda_{NN} + (1 - \lambda_{PP})\frac{0.5}{0.5}\right]}}.$$
(33)

$$MCC = \frac{\lambda_{PP} + \lambda_{NN} - 1}{\sqrt{[\lambda_{PP} + (1 - \lambda_{NN})][\lambda_{NN} + (1 - \lambda_{PP})]}}$$
(34)

9. Bookmaker Informedness (BM).

$$BM = \lambda_{PP} + \lambda_{NN} - 1. \tag{35}$$

10. Markedness (MK).

$$MK = \frac{\pi_P}{\pi_P + \frac{(1 - \lambda_{NN})}{\lambda_{PP}} \pi_N} + \frac{\pi_N}{\pi_N + \frac{(1 - \lambda_{PP})}{\lambda_{NN}} \pi_P} - 1.$$
(36)

$$MK = \frac{\frac{1}{2}}{\frac{1}{2} + \frac{(1 - \lambda_{NN})}{\lambda_{PP}} \cdot \frac{1}{2}} + \frac{\frac{1}{2}}{\frac{1}{2} + \frac{(1 - \lambda_{PP})}{\lambda_{NN}} \cdot \frac{1}{2}} - 1.$$
(37)

$$MK = \frac{\frac{1}{2}1}{1 + \frac{(1 - \lambda_{NN})}{\lambda_{PP}}} + \frac{1}{1 + \frac{(1 - \lambda_{PP})}{\lambda_{NN}}} - 1.$$
 (38)

3. Derivation of $\mu = \mu(\lambda_{PP}, \lambda_{NN}, \delta)$.

1. Sensitivity (SNS).

$$SNS = \lambda_{PP}. (39)$$

2. Specificity (SPC).

$$SPC = \lambda_{NN}. (40)$$

3. Precision (PRC).

$$PRC = \frac{\lambda_{PP} \pi_P}{\lambda_{PP} \pi_P + (1 - \lambda_{NN}) \pi_N} = \frac{\lambda_{PP} \frac{1 + \delta}{2}}{\lambda_{PP} \frac{1 + \delta}{2} + (1 - \lambda_{NN}) \frac{1 - \delta}{2}}.$$
 (41)

$$PRC = \frac{\lambda_{PP}(1+\delta)}{\lambda_{PP}(1+\delta) + (1-\lambda_{NN})(1-\delta)}.$$
 (42)

4. Negative Predictive Value (NPV).

$$NPV = \frac{\lambda_{NN}\pi_{N}}{\lambda_{NN}\pi_{N} + (1 - \lambda_{PP})\pi_{P}} = \frac{\lambda_{NN}\frac{1 - \delta}{2}}{\lambda_{NN}\frac{1 - \delta}{2} + (1 - \lambda_{PP})\frac{1 + \delta}{2}}.$$
(43)

$$NPV = \frac{\lambda_{NN}(1-\delta)}{\lambda_{NN}(1-\delta) + (1-\lambda_{PP})(1+\delta)}.$$
(44)

5. Accuracy (ACC).

$$ACC = \lambda_{PP} \frac{1+\delta}{2} + \lambda_{NN} \frac{1-\delta}{2} = \frac{1}{2} [\lambda_{PP} (1+\delta) + \lambda_{NN} (1-\delta)]. \tag{45}$$

6. F_1 score.

$$F_1 = 2 \frac{\lambda_{PP} \frac{1+\delta}{2}}{(1+\lambda_{PP}) \frac{1+\delta}{2} + (1-\lambda_{NN}) \frac{1-\delta}{2}} = \frac{2 \lambda_{PP} (1+\delta)}{(1+\lambda_{PP}) (1+\delta) + (1-\lambda_{NN}) (1-\delta)}.$$
 (46)

7. Geometric Mean (*GM*).

$$GM \equiv \sqrt{SNS \cdot SPC} = \sqrt{\lambda_{PP} \cdot \lambda_{NN}}.$$
 (47)

8. Matthews Correlation Coefficient (*MCC*)

$$MCC = \frac{\lambda_{PP} + \lambda_{NN} - 1}{\sqrt{\left[\lambda_{PP} + (1 - \lambda_{NN})\frac{\pi_N}{\pi_P}\right]\left[\lambda_{NN} + (1 - \lambda_{PP})\frac{\pi_P}{\pi_N}\right]}}$$
(48)

$$MCC = \frac{\lambda_{PP} + \lambda_{NN} - 1}{\left[\lambda_{PP} + (1 - \lambda_{NN}) \frac{\frac{1 - \delta}{2}}{\frac{1 + \delta}{2}}\right] \left[\lambda_{NN} + (1 - \lambda_{PP}) \frac{\frac{1 + \delta}{2}}{\frac{1 - \delta}{2}}\right]}$$
(49)

$$MCC = \frac{\lambda_{PP} + \lambda_{NN} - 1}{\sqrt{\left[\lambda_{PP} + (1 - \lambda_{NN})\frac{1 - \delta}{1 + \delta}\right]\left[\lambda_{NN} + (1 - \lambda_{PP})\frac{1 + \delta}{1 - \delta}\right]}}.$$
(50)

9. Bookmaker Informedness (BM).

$$BM = \lambda_{PP} + \lambda_{NN} - 1. \tag{51}$$

10. Markedness (MK).

$$MK = MK = \frac{\pi_P}{\pi_P + \frac{(1 - \lambda_{NN})}{\lambda_{PP}} \pi_N} + \frac{\pi_N}{\pi_N + \frac{(1 - \lambda_{PP})}{\lambda_{NN}} \pi_P} - 1.$$
 (52)

$$MK = MK = \frac{\frac{1+\delta}{2}}{\frac{1+\delta}{2} + \frac{(1-\lambda_{NN})}{\lambda_{PP}} \frac{1-\delta}{2}} + \frac{\frac{1-\delta}{2}}{\frac{1-\delta}{2} + \frac{(1-\lambda_{PP})}{\lambda_{NN}} \frac{1+\delta}{2}} - 1.$$
 (53)

$$MK = MK = \frac{1+\delta}{(1+\delta) + \frac{(1-\lambda_{NN})}{\lambda_{PP}}(1-\delta)} + \frac{1-\delta}{(1-\delta) + \frac{(1-\lambda_{PP})}{\lambda_{NN}}(1+\delta)} - 1.$$
 (54)

- 4. Derivation of $B_{\mu} = B_{\mu}(\lambda_{PP}, \lambda_{NN}, \delta)$.
- 1. Sensitivity (SNS).

$$B_{SNS}(\lambda_{PP}, \lambda_{NN}, \delta) = SNS(\lambda_{PP}, \lambda_{NN}, \delta) - SNS_b(\lambda_{PP}, \lambda_{NN}, \delta) = \lambda_{PP} - \lambda_{PP} = 0.$$
 (55)

2. Specificity (SPC).

$$B_{SPC}(\lambda_{PP}, \lambda_{NN}, \delta) = SPC(\lambda_{PP}, \lambda_{NN}, \delta) - SPC_b(\lambda_{PP}, \lambda_{NN}, \delta) = \lambda_{NN} - \lambda_{NN} = 0.$$
 (56)

3. Precision (PRC).

$$B_{PRC}(\lambda_{PP}, \lambda_{NN}, \delta) = PRC(\lambda_{PP}, \lambda_{NN}, \delta) - PRC_b(\lambda_{PP}, \lambda_{NN}, \delta).$$
 (57)

$$B_{PRC}(\lambda_{PP}, \lambda_{NN}, \delta) = \frac{\lambda_{PP}(1+\delta)}{\lambda_{PP}(1+\delta) + (1-\lambda_{NN})(1-\delta)} - \frac{\lambda_{PP}}{\lambda_{PP} + (1-\lambda_{NN})}.$$
 (58)

$$B_{PRC}(\lambda_{PP}, \lambda_{NN}, \delta) = \frac{1 + \delta}{(1 + \delta) + \frac{1 - \lambda_{NN}}{\lambda_{PP}} (1 - \delta)} - \frac{1}{1 + \frac{1 - \lambda_{NN}}{\lambda_{PP}}}.$$
(59)

4. Negative Predictive Value (NPV).

$$B_{NPV}(\lambda_{PP}, \lambda_{NN}, \delta) = NPV(\lambda_{PP}, \lambda_{NN}, \delta) - NPV_b(\lambda_{PP}, \lambda_{NN}, \delta). \tag{60}$$

$$B_{NPV}(\lambda_{PP}, \lambda_{NN}, \delta) = \frac{\lambda_{NN}(1 - \delta)}{\lambda_{NN}(1 - \delta) + (1 - \lambda_{PP})(1 + \delta)} - \frac{\lambda_{NN}}{\lambda_{NN} + (1 - \lambda_{PP})}.$$
 (61)

$$B_{NPV}(\lambda_{PP}, \lambda_{NN}, \delta) = \frac{1 - \delta}{(1 - \delta) + \frac{1 - \lambda_{PP}}{\lambda_{NN}} (1 + \delta)} - \frac{1}{1 + \frac{1 - \lambda_{PP}}{\lambda_{NN}}}.$$
(62)

5. Accuracy (ACC).

$$B_{ACC}(\lambda_{PP}, \lambda_{NN}, \delta) = ACC(\lambda_{PP}, \lambda_{NN}, \delta) - ACC_b(\lambda_{PP}, \lambda_{NN}, \delta). \tag{63}$$

$$B_{ACC}(\lambda_{PP}, \lambda_{NN}, \delta) = \lambda_{PP} \frac{1+\delta}{2} + \lambda_{NN} \frac{1-\delta}{2} - \frac{\lambda_{PP} + \lambda_{NN}}{2}.$$
 (64)

$$B_{ACC}(\lambda_{PP}, \lambda_{NN}, \delta) = \frac{\lambda_{PP}}{2} + \frac{\lambda_{PP}\delta}{2} + \frac{\lambda_{NN}}{2} - \frac{\lambda_{NN}\delta}{2} - \frac{\lambda_{PP}}{2} - \frac{\lambda_{NN}}{2} = \frac{\lambda_{PP}\delta}{2} - \frac{\lambda_{NN}\delta}{2}.$$
 (65)

$$B_{ACC}(\lambda_{PP}, \lambda_{NN}, \delta) = \frac{\delta}{2}(\lambda_{PP} - \lambda_{NN}). \tag{66}$$

6. F_1 score.

$$B_{F_1}(\lambda_{PP}, \lambda_{NN}, \delta) = F_1(\lambda_{PP}, \lambda_{NN}, \delta) - F_{1h}(\lambda_{PP}, \lambda_{NN}, \delta). \tag{67}$$

$$B_{F1}(\lambda_{PP}, \lambda_{NN}, \delta) = \frac{2\lambda_{PP}(1+\delta)}{(1+\lambda_{PP})(1+\delta) + (1-\lambda_{NN})(1-\delta)} - \frac{2\lambda_{PP}}{2+\lambda_{PP} - \lambda_{NN}}.$$
 (68)

7. Geometric Mean (GM).

$$B_{GM}(\lambda_{PP}, \lambda_{NN}, \delta) = GM(\lambda_{PP}, \lambda_{NN}, \delta) - GM_b(\lambda_{PP}, \lambda_{NN}, \delta) = \sqrt{\lambda_{PP} \cdot \lambda_{NN}} - \sqrt{\lambda_{PP} \cdot \lambda_{NN}} = 0.$$
 (69)

8. Matthews Correlation Coefficient (*MCC*).

$$B_{MCC}(\lambda_{PP}, \lambda_{NN}, \delta) = MCC(\lambda_{PP}, \lambda_{NN}, \delta) - MCC_b(\lambda_{PP}, \lambda_{NN}, \delta). \tag{70}$$

$$B_{MCC}(\lambda_{PP}, \lambda_{NN}, \delta) = \frac{\lambda_{PP} + \lambda_{NN} - 1}{\sqrt{\left[\lambda_{PP} + (1 - \lambda_{NN})\frac{1 - \delta}{1 + \delta}\right]\left[\lambda_{NN} + (1 - \lambda_{PP})\frac{1 + \delta}{1 - \delta}\right]}}{\frac{\lambda_{PP} + \lambda_{NN} - 1}{\sqrt{\left[\lambda_{PP} + (1 - \lambda_{NN})\right]\left[\lambda_{NN} + (1 - \lambda_{PP})\right]}}}$$
(71)

$$B_{MCCn}(\lambda_{PP}, \lambda_{NN}, \delta) = MCCn(\lambda_{PP}, \lambda_{NN}, \delta) - MCCn_b(\lambda_{PP}, \lambda_{NN}). \tag{72}$$

$$B_{MCCn}(\lambda_{PP}, \lambda_{NN}, \delta) = \frac{MCC(\lambda_{PP}, \lambda_{NN}, \delta) + 1}{2} - \frac{MCC_b(\lambda_{PP}, \lambda_{NN}) + 1}{2}.$$
 (73)

$$B_{MCCn}(\lambda_{PP}, \lambda_{NN}, \delta) = \frac{MCC(\lambda_{PP}, \lambda_{NN}, \delta) - MCC_b(\lambda_{PP}, \lambda_{NN})}{2}.$$
 (74)

$$B_{MCCn}(\lambda_{PP}, \lambda_{NN}, \delta) = \frac{B_{MCC}(\lambda_{PP}, \lambda_{NN}, \delta)}{2}.$$
 (75)

$$B_{MCCn}(\lambda_{PP}, \lambda_{NN}, \delta) = \frac{\lambda_{PP} + \lambda_{NN} - 1}{2\sqrt{\left[\lambda_{PP} + (1 - \lambda_{NN})\frac{1 - \delta}{1 + \delta}\right]\left[\lambda_{NN} + (1 - \lambda_{PP})\frac{1 + \delta}{1 - \delta}\right]}} - \frac{\lambda_{PP} + \lambda_{NN} - 1}{2\sqrt{\left[\lambda_{PP} + (1 - \lambda_{NN})\right]\left[\lambda_{NN} + (1 - \lambda_{PP})\right]}}$$
(76)

9. Bookmaker Informedness (*BM*).

$$B_{BM}(\lambda_{PP}, \lambda_{NN}, \delta) = BM(\lambda_{PP}, \lambda_{NN}, \delta) - BM_b(\lambda_{PP}, \lambda_{NN}, \delta). \tag{77}$$

$$B_{BM}(\lambda_{PP}, \lambda_{NN}, \delta) = (\lambda_{PP} + \lambda_{NN} - 1) - (\lambda_{PP} + \lambda_{NN} - 1) = 0. \tag{78}$$

10. Markedness (MK).

$$B_{MK}(\lambda_{PP}, \lambda_{NN}, \delta) = MK(\lambda_{PP}, \lambda_{NN}, \delta) - MK_b(\lambda_{PP}, \lambda_{NN}, \delta). \tag{79}$$

$$B_{MK}(\lambda_{PP}, \lambda_{NN}, \delta) = [PRC(\lambda_{PP}, \lambda_{NN}, \delta) + NPV(\lambda_{PP}, \lambda_{NN}, \delta) - 1] - [PRC_b(\lambda_{PP}, \lambda_{NN}) + NPV_b(\lambda_{PP}, \lambda_{NN}) - 1].$$
(80)

$$B_{MK}(\lambda_{PP}, \lambda_{NN}, \delta) = [PRC(\lambda_{PP}, \lambda_{NN}, \delta) - PRC_b(\lambda_{PP}, \lambda_{NN})] + [NPV(\lambda_{PP}, \lambda_{NN}, \delta) - NPV_b(\lambda_{PP}, \lambda_{NN})].$$
(81)

$$B_{MK}(\lambda_{PP}, \lambda_{NN}, \delta) = B_{PRC}(\lambda_{PP}, \lambda_{NN}, \delta) + B_{NPV}(\lambda_{PP}, \lambda_{NN}, \delta). \tag{82}$$

$$B_{MK}(\lambda_{PP}, \lambda_{NN}, \delta) = \frac{1 + \delta}{(1 + \delta) + \frac{1 - \lambda_{NN}}{\lambda_{PP}} (1 - \delta)} - \frac{1}{1 + \frac{1 - \lambda_{NN}}{\lambda_{PP}}} + \frac{1 - \delta}{(1 - \delta) + \frac{1 - \lambda_{PP}}{\lambda_{NN}} (1 + \delta)} - \frac{1}{1 + \frac{1 - \lambda_{PP}}{\lambda_{NN}}}.$$
(83)

$$B_{MKn}(\lambda_{PP}, \lambda_{NN}, \delta) = MKn(\lambda_{PP}, \lambda_{NN}, \delta) - MKn_b(\lambda_{PP}, \lambda_{NN}). \tag{84}$$

$$B_{MKn}(\lambda_{PP}, \lambda_{NN}, \delta) = \frac{MK(\lambda_{PP}, \lambda_{NN}, \delta) + 1}{2} - \frac{MK_b(\lambda_{PP}, \lambda_{NN}) + 1}{2}.$$
 (85)

$$B_{MKn}(\lambda_{PP}, \lambda_{NN}, \delta) = \frac{MK(\lambda_{PP}, \lambda_{NN}, \delta) - MK_b(\lambda_{PP}, \lambda_{NN})}{2} = \frac{B_{MK}(\lambda_{PP}, \lambda_{NN}, \delta)}{2}.$$
 (86)

$$B_{MKn}(\lambda_{PP}, \lambda_{NN}, \delta) = \frac{B_{PRC}(\lambda_{PP}, \lambda_{NN}, \delta) + B_{NPV}(\lambda_{PP}, \lambda_{NN}, \delta)}{2}.$$
(87)

$$B_{MKn}(\lambda_{PP}, \lambda_{NN}, \delta) = \frac{1}{2} \left(\frac{1+\delta}{(1+\delta) + \frac{1-\lambda_{NN}}{\lambda_{PP}} (1-\delta)} - \frac{1}{1 + \frac{1-\lambda_{NN}}{\lambda_{PP}}} + \frac{1-\delta}{(1-\delta) + \frac{1-\lambda_{PP}}{\lambda_{NN}} (1+\delta)} - \frac{1}{1 + \frac{1-\lambda_{PP}}{\lambda_{NN}}} \right).$$
(88)