# **Annex 3**

In Table 1 the analysis results of the secondary set of papers according to the complexity issues in FIS development are presented.

**Table 1.** Complexity issues found in the secondary set of papers (1 – strong description is found (i.e., an issue is described to the degree where a reader can understand and compare it to the same aspect in another paper), 0 – not found any description)

| Reference | Year | CC (1) | CFR (2) | CMF (3) | DC (4) |
| --- | --- | --- | --- | --- | --- |
| **[1, 2, 3]** | 2019 | 1 | 0 | 0 | 0 |
| **[4]** | 0 | 0 | 1 | 0 |
| **[5, 6]** | 1 | 1 | 0 | 0 |
| **[7, 8, 9]** | 0 | 1 | 1 | 0 |
| **[10]** | 0 | 1 | 0 | 0 |
| **[11]** | 1 | 1 | 1 | 0 |
| **[12, 13, 14, 15, 16, 17, 18]** | 0 | 1 | 0 | 0 |
| **[19]** | 2018 | 0 | 1 | 1 | 0 |
| **[20, 21, 22]** | 0 | 1 | 0 | 0 |
| **[23, 24]** | 1 | 1 | 0 | 0 |
| **[25]** | 2017 | 0 | 1 | 1 | 0 |
| **[26]** |  | 1 | 0 | 0 | 0 |
| **[27, 28]** |  | 0 | 0 | 0 | 1 |
| **[29, 30, 31, 32, 33]** | 0 | 1 | 0 | 0 |
| **[34, 35, 36, 37]** | 2016 | 1 | 0 | 0 | 0 |
| **[38]** | 0 | 0 | 1 | 0 |
| **[39, 40]** | 0 | 1 | 1 | 0 |
| **[41]** | 1 | 0 | 1 | 0 |
| **[42]** | 1 | 1 | 1 | 0 |
| **[43, 44]** | 0 | 1 | 0 | 0 |
| **[45, 46]** | 2015 | 0 | 0 | 1 | 0 |
| **[47, 48, 49, 50, 51]** | 0 | 1 | 0 | 0 |
| **[53, 54, 55]** | 1 | 0 | 0 | 0 |
| **[56, 57]** | 0 | 1 | 1 | 0 |
| **[58]** | 1 | 1 | 0 | 0 |
| **[59]** | 2014 | 0 | 0 | 1 | 0 |
| **[60, 61, 62, 63, 64, 65, 66]** | 0 | 1 | 0 | 0 |
| **[67, 68]** | 0 | 1 | 1 | 0 |
| **[69]** | 0 | 0 | 0 | 1 |
| **[70, 71]** | 1 | 0 | 0 | 0 |
| **[72]** | 1 | 1 | 0 | 0 |
| **[73, 74]** | 2013 | 1 | 0 | 0 | 0 |
| **[75, 76, 77, 78, 79, 80, 81, 82, 83, 84, 85]** | 0 | 1 | 0 | 0 |
| **[86]** | 1 | 1 | 1 | 0 |
| **[87]** | 0 | 0 | 1 | 0 |
| **[88]** | 1 | 1 | 0 | 0 |
| **[89]** | 2012 | 1 | 0 | 1 | 0 |
| **[90]** | 0 | 0 | 0 | 1 |
| **[91]** | 1 | 0 | 0 | 0 |
| **[92, 93, 94, 95, 96]** | 0 | 1 | 0 | 0 |
| **[97, 98, 99]** | 0 | 1 | 1 | 0 |
| **[100]** | 1 | 1 | 0 | 0 |
| **[101, 102, 103, 104]** | 2011 | 0 | 1 | 1 | 0 |
| **[105, 106]** | 0 | 0 | 1 | 0 |
| **[107, 108]** | 1 | 1 | 0 | 0 |
| **[109]** | 1 | 1 | 1 | 0 |
| **[110, 111, 112, 113, 114]** | 0 | 1 | 0 | 0 |
| **[115, 116, 117, 118]** | 2010 | 0 | 1 | 0 | 0 |
| **[119, 120]** | 0 | 1 | 1 | 0 |
| **[121]** | 0 | 0 | 1 | 0 |
| **[122, 123]** | 2009 | 0 | 1 | 0 | 0 |
| **[124, 125, 126]** | 1 | 0 | 0 | 0 |
| **[127, 128]** | 0 | 1 | 1 | 0 |
| **[129]** | 2008 | 0 | 1 | 1 | 0 |
| **[130, 131, 132]** | 0 | 1 | 0 | 0 |
| **[133]** | 1 | 1 | 0 | 0 |
| **[134]** | 2007 | 1 | 0 | 0 | 0 |
| **[135, 136, 137]** | 0 | 1 | 0 | 0 |
| **[138, 139]** | 0 | 1 | 1 | 0 |
| **[140, 141, 142]** | 2006 | 0 | 1 | 1 | 0 |
| **[143, 144]** | 0 | 1 | 0 | 0 |
| **[145, 146, 147]** | 2005 | 0 | 1 | 1 | 0 |
| **[148, 149]** | 0 | 1 | 0 | 0 |
| **[150] [126]** | 1 | 1 | 0 | 0 |
| **[151, 152, 153]** | 2004 | 0 | 1 | 0 | 0 |
| **[154]** | 0 | 1 | 1 | 0 |
| **[155]** | 1 | 1 | 1 | 0 |
| **[156]** | 2003 | 1 | 0 | 0 | 0 |
| **[157, 158, 159, 160, 161]** | 0 | 1 | 0 | 0 |
| **[162]** | 1 | 1 | 0 | 0 |
| **[163]** | 0 | 1 | 1 | 0 |
| **[164, 165, 166, 167, 168, 169]** | 2002 | 0 | 1 | 0 | 0 |
| **[170]** | 0 | 1 | 1 | 0 |
| **[171, 172]** | 2001 | 0 | 1 | 0 | 0 |
| **[173]** | 1 | 1 | 1 | 0 |
| **[174, 175, 176, 177]** | 0 | 1 | 1 | 0 |
| **[178]** | 0 | 0 | 1 | 0 |
| **[179, 180]** | 2000 | 0 | 1 | 1 | 0 |
| **[181, 182, 183, 184, 185]** | 0 | 1 | 0 | 0 |
| **[186]** | 1 | 1 | 0 | 0 |
| **[187]** | 0 | 0 | 1 | 0 |
| **[188]** | 1999 | 0 | 1 | 0 | 1 |
| **[189]** | 0 | 0 | 1 | 0 |
| **[190]** | 0 | 1 | 0 | 0 |
| **[191]** | 1998 | 0 | 1 | 0 | 0 |
| **[192, 193]** | 0 | 0 | 1 | 0 |
| **[194, 195, 196]** | 0 | 1 | 1 | 0 |
| **[197, 198]** | 1 | 1 | 0 | 0 |
| **[199]** | 1997 | 1 | 0 | 0 | 0 |
| **[200, 201, 202]** | 0 | 1 | 0 | 0 |
| **[203]** | 1996 | 1 | 0 | 0 | 0 |
| **[204, 205, 206]** | 0 | 1 | 0 | 0 |
| **[207]** | 0 | 1 | 0 | 1 |
| **[208, 209]** | 1995 | 0 | 1 | 0 | 0 |
| **[210, 211]** | 0 | 0 | 1 | 0 |
| **[212, 213]** | 1993 | 0 | 1 | 0 | 0 |
| **[214]** | 1 | 1 | 0 | 0 |

In Table 2, we have presented the mapping between References in this Annex 3 Table 1 and References in the main paper, since in the main paper only summarized results are presented and, therefore, not all papers from the secondary set are cited.

**Table 2. The mapping table between References in this Annex 3 and References in the main paper.**

|  |  |  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- | --- | --- |
| **No.** | **In Annex 3 Table 1** | **In the main paper** | **No.** | **In Annex 3 Table 1** | **In the main paper** | **No.** | **In Annex 3 Table 1** | **In the main paper** |
| 1. | **[1]** | **[4]** | 30. | **[71]** | **[95]** | 59. | **[144]** | **[122]** |
| 2. | **[2]** | **[69]** | 31. | **[75]** | **[12]** | 60. | **[145]** | **[123]** |
| 3. | **[3]** | **[70]** | 32. | **[76]** | **[96]** | 61. | **[146]** | **[124]** |
| 4. | **[5]** | **[71]** | 33. | **[79]** | **[97]** | 62. | **[147]** | **[125]** |
| 5. | **[6]** | **[72]** | 34. | **[81]** | **[98]** | 63. | **[150]** | **[126]** |
| 6. | **[7]** | **[73]** | 35. | **[83]** | **[99]** | 64. | **[158]** | **[127]** |
| 7. | **[8]** | **[74]** | 36. | **[84]** | **[100]** | 65. | **[161]** | **[128]** |
| 8. | **[9]** | **[75]** | 37. | **[86]** | **[101]** | 66. | **[175]** | **[129]** |
| 9. | **[10]** | **[76]** | 38. | **[87]** | **[102]** | 67. | **[178]** | **[130]** |
| 10. | **[11]** | **[77]** | 39. | **[90]** | **[103]** | 68. | **[180]** | **[131]** |
| 11. | **[13]** | **[78]** | 40. | **[94]** | **[104]** | 69. | **[185]** | **[132]** |
| 12. | **[16]** | **[79]** | 41. | **[95]** | **[105]** | 70. | **[187]** | **[133]** |
| 13. | **[19]** | **[80]** | 42. | **[96]** | **[106]** | 71. | **[188]** | **[134]** |
| 14. | **[20]** | **[87]** | 43. | **[97]** | **[107]** | 72. | **[189]** | **[135]** |
| 15. | **[25]** | **[2]** | 44. | **[99]** | **[108]** | 73. | **[190]** | **[136]** |
| 16. | **[26]** | **[8]** | 45. | **[100]** | **[109]** | 74. | **[191]** | **[137]** |
| 17. | **[27]** | **[82]** | 46. | **[101]** | **[27]** | 75. | **[193]** | **[138]** |
| 18. | **[38]** | **[84]** | 47. | **[106]** | **[110]** | 76. | **[198]** | **[139]** |
| 19. | **[39]** | **[106]** | 48. | **[108]** | **[111]** | 77. | **[201]** | **[140]** |
| 20. | **[45]** | **[10]** | 49. | **[109]** | **[112]** | 78. | **[206]** | **[141]** |
| 21. | **[49]** | **[86]** | 50. | **[117]** | **[113]** | 79. | **[207]** | **[142]** |
| 22. | **[52]** | **[87]** | 51. | **[120]** | **[114]** | 80. | **[214]** | **[143]** |
| 23. | **[54]** | **[88]** | 52. | **[122]** | **[6]** |  |  |  |
| 24. | **[57]** | **[89]** | 53. | **[123]** | **[115]** |  |  |  |
| 25. | **[58]** | **[90]** | 54. | **[126]** | **[116]** |  |  |  |
| 26. | **[65]** | **[91]** | 55. | **[127]** | **[117]** |  |  |  |
| 27. | **[68]** | **[92]** | 56. | **[138]** | **[119]** |  |  |  |
| 28. | **[69]** | **[93]** | 57. | **[139]** | **[120]** |  |  |  |
| 29. | **[70]** | **[94]** | 58. | **[141]** | **[121]** |  |  |  |

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