

1 A generalised ecohydrological landscape classification for 2 assessing ecosystem risk in Australia due to an altering water 3 regime

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10 Abstract

11 Describing and classifying a landscape for environmental impact and risk assessment purposes is a
12 non-trivial challenge because this requires region specific landscape classifications that cater for
13 region specific impacts. Assessing impacts on ecosystems from the extraction of water resources
14 across large regions requires linking landscape features to their water requirements. We present the
15 rationale and implementation of an ecohydrological classification for regions where coal mine and
16 coal seam gas developments may impact on water. Our classification provides the essential
17 framework for modelling the potential impact of hydrological changes from future coal resource
18 developments at the landscape level.

19 We develop an attribute-based system that provides representations of the ecohydrological entities
20 and their connection to landscape features and make use of existing broad-level, classification
21 schemes into an attribute-based system. We incorporate a rule-set with prioritisation, which
22 underpin risk modelling and make the scheme resource efficient, where spatial landscape or
23 ecosystem classification schemes, developed for other purposes, already exist.

24 A consistent rule-set and conceptualised landscape processes and functions allow combining diverse
25 data with existing classification schemes. This makes the classification transparent, repeatable, and
26 adjustable, should new data become available. We apply the approach in three geographically
27 different regions, with widely disparate information sources, for the classification, and provide a
28 detailed example of its application. We propose that it is widely applicable around the world for
29 linking ecohydrology to environmental impacts.

30 **Keywords:** Typology, ecology, hydrology, causal pathway, bioregional assessments, environmental
31 impact, risk analysis

32 1 Introduction

33 The categorisation of the Earth's surface into geo-ecological landscape classes provides a way to
34 simplify the complexity of the form and function of the landscape and provides vital contextual
35 information to support land and water management, and policy initiatives. This includes identifying
36 geographical regions within which landscape-scale attributes, such as climate, topography, geology,
37 and land cover, that are homogeneous and distinctive compared to other regions. It involves

38 identifying broad-scale, general patterns, processes, and functions. Landscape class units are
39 'ecologically equivalent', having the same dominant processes that sustain a similar suite of species,
40 and are likely to respond in similar ways to management initiatives or environmental changes. This
41 ecological equivalence enables the selection of assessment locations for monitoring, measurement
42 or experimentation, and it enables the extrapolation of results to all areas within the same
43 ecological class (Hawkins and Norris, 2000; MacMillan et al., 2003; Cullum et al., 2016a; Cullum et
44 al., 2016b).

45 Such a landscape classification also explains variation in ecological characteristics (e.g. assemblage
46 structure) and is predictive of the ecological attributes of those areas. This predictive quality is useful
47 for defining ecological criteria, identifying reference and degraded sites, defining conservation goals,
48 including the assessment of biodiversity, and the setting of restoration objectives (Hawkins et al.,
49 2000; McMahon et al., 2001; Snelder et al., 2004).

50 In summary, landscape classification is a way of dividing a landscape into components where the
51 characteristics within the components are more similar than the characteristics between the
52 components. That is, the components have their own distinct features that separate them from the
53 other components.

54 However, describing and classifying a landscape for environmental impact and risk assessment
55 purposes is a non-trivial challenge, where hydrological records are limited (see e.g. Wolfe et al.,
56 2019). This is the case for many regions in Australia, where low population densities, high
57 urbanisation and limits in (water) resource management information exist. For our purpose, which
58 was the assessment of risk to ecosystems within the regions of the Bioregional Assessments
59 Programme (Bioregional Assessments, 2018), we needed a landscape classification that reflected the
60 hydrological connectivity of surface and groundwater with ecosystems in the landscape. The
61 Bioregional Assessment Programme, an Australian regional scale impact assessment, investigated
62 the impacts and risks of coal seam gas (CSG) and large coal mining developments on water resources
63 and water-dependent assets via a water pathway (Bioregional Assessments, 2018). This investigation
64 focussed on the landscape level, that is on areas within the regions where the landscape is made up
65 of different interacting land-uses and ecosystems.

66 In our case, the broad scale assessments of impacts from resource developments on ecosystems
67 required an understanding of landscape composition and structure, and how these relate to the
68 ecosystems embedded in the landscape. The type and composition of the landscape components
69 are dependent on the focus of the assessment and therefore require careful consideration of the
70 questions the assessment seeks to answer (Wiens and Milne, 1989; Eigenbrot, 2016). For Australia,
71 there are several landscape level classifications available (see e.g. Thackway and Cresswell, 1995;
72 Pain et al., 2011; Hall et al., 2015; NVIS Technological Working Group, 2017; Gharari et al., 2011).
73 Unfortunately, these available classifications are not directly applicable for our assessment regions
74 because there is no alignment between the regions and existing classification boundaries, or the
75 classifications, even if they include ecohydrological elements, are limited to their locations or
76 domain of interest.

77 Identifying the water dependency of landscape components is a prerequisite when analysing the
78 potential impacts of proposed coal and gas resource developments on water resources at a regional
79 scale. For example, coal resource developments generally need to manage both groundwater and
80 surface water as part of their operations. With multiple developments within the one region,

81 impacts are likely to go beyond the local scale and affect ecosystems at the landscape level (see for
82 example Bioregional Assessments, 2018, 2019). In this context, there is a need for an ecological
83 classification of the landscape that identifies and causally connects the water dependency of its
84 components to activities of resource extraction, in a spatially explicit manner. Further, there is a
85 need to identify impact pathways between resource extraction sites and the ecosystems that show
86 causal connectivity between extraction activities and ecosystem impacts.

87 Land classification systems reveal patterns and underlying drivers of ecosystem structure and
88 function, or produce a tractable unit of assessment for evaluating environmental change (Hobbs and
89 McIntyre, 2005; Poff et al., 2010). Many different classification approaches and methodologies
90 currently exist to represent ecosystems in a landscape. This includes the interim bioregional
91 classification for Australia (IBRA), which provides the basis for defining and managing the national
92 reserve system and; the national vegetation information systems (NVIS), that describes the extent
93 and distribution of vegetation ecosystems for the Australian continent (Thackway and Cresswell,
94 1995; Department of Agriculture, 2021). Classifications addressing hydrology in Australia incorporate
95 a framework for river management that delineate boundaries between homogenous landscape
96 components, based on either their dependency on surface water or groundwater regimes (Poff et
97 al., 2010; Aquatic Ecosystems Task Group, 2012; Olden et al., 2012). However, none of these
98 classifications describe ecohydrological connections between waters and the wider landscape. For
99 example, IBRA and NVIS are based purely on vegetation classifications and so do not contain any
100 hydrological details, while the available hydrological classifications focus purely on the streams and
101 waterbodies within the landscape, as their focus is on aquatic organisms and environmental flows.
102 While both these elements are part of the immediate landscape surrounding water bodies, they do
103 not in themselves provide conceptual and direct linkages between changes in water and ecosystem
104 responses in the wider landscape. Therefore, a standardised approach to formulating classifications
105 that combine these two aspects, ecosystems and their water sources, is lacking.

106 This conundrum exists because different analysis contexts require classifications for different
107 purposes, ranging from conservation planning, habitat mapping, resource assessment and
108 vegetation modelling, and because there is contention between the generality of broad
109 classifications and their applicability at the local scale (Leathwick et al., 2003; Abella et al., 2003;
110 Poulter et al., 2011; Cullum et al., 2016b; Pyne et al., 2017). Hence, we needed a new classification
111 system, when evaluating water dependency in the context of regional scale for multiple coal and
112 coal seam gas resource developments. This new system must incorporate surface water and
113 groundwater regimes into a spatial demarcation of ecosystem boundaries in the landscape.
114 Including surface water and groundwater regimes will provide the establishing of conceptual
115 connection between impacts from developments on surface water and groundwater within the
116 classification, and the classification must be spatially explicit, to enable a landscape wide analysis of
117 those impacts so that one can link changes in water at one part of the landscape to ecological
118 responses at another part of the landscape.

119 With this context in mind, the objectives for this paper are to:

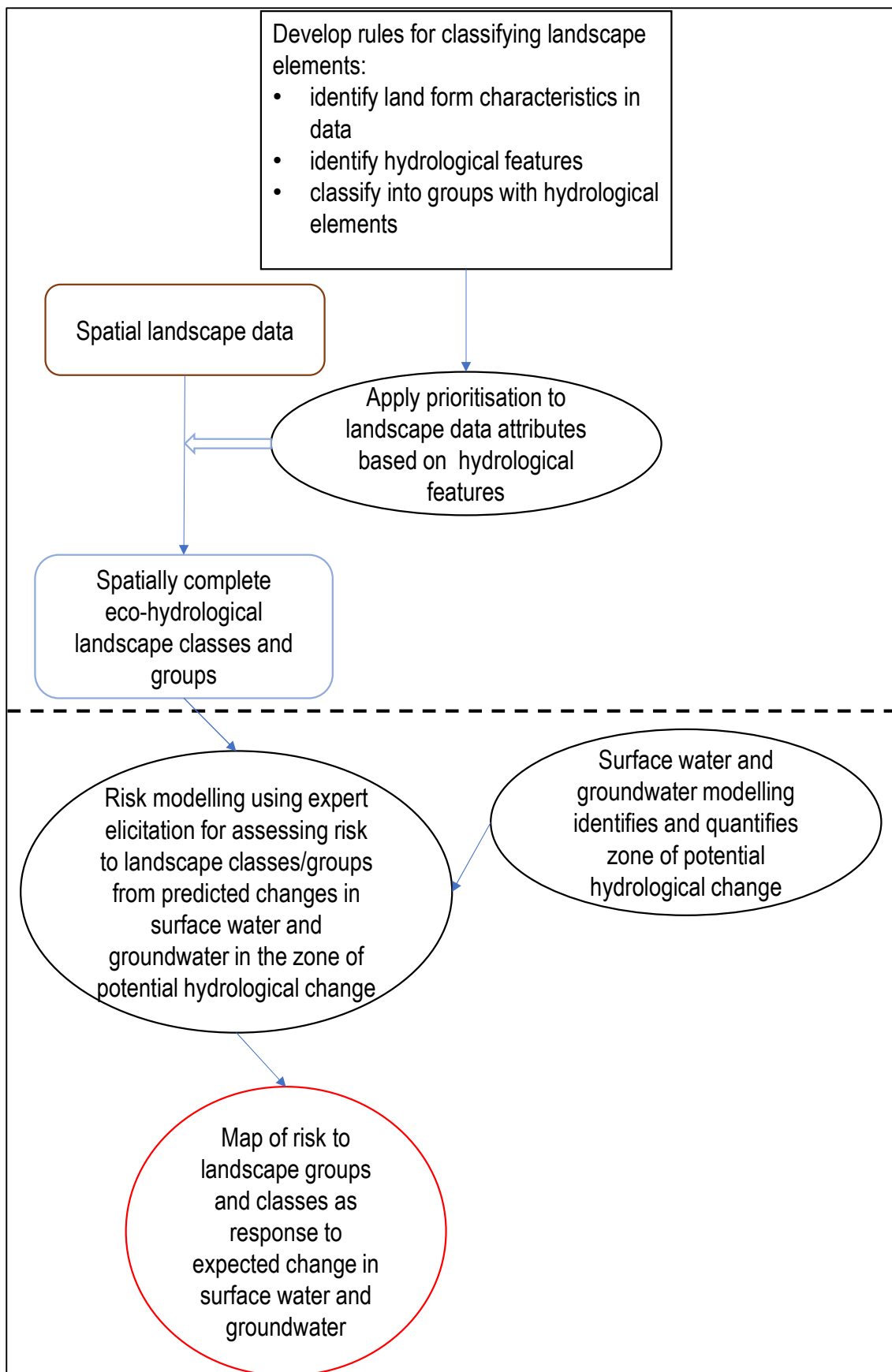
- 120 1. characterise a regional level landscape based on patterns in land use, ecology, geomorphology
121 and hydrology,
- 122 2. develop landscape classes of water-dependent, remnant and human-modified features, and
- 123 3. ensure landscape classes sit within a common framework that aids in formulating
124 conceptual models and patterns of water dependency across the landscape.

125 Here, we present the rationale, formulation and implementation of an ecohydrological landscape
126 classification. Based on a generalised conceptual model of the typical hydrological connectivity
127 within landscape features in a region, the classification integrates pre-existing, broad-level
128 classification schemes into an attribute-based schema applied at the regional scale. It places the
129 landscape classification within a common framework (i.e. a framework that is common to all
130 landscape elements in the region) that aids in formulating conceptual models and patterns in water
131 dependency across the landscape. This makes our approach generally applicable for assessments
132 aimed at regional hydrological impacts on, and risks to, ecosystems. Importantly, the classification
133 also provides the ability to develop a conceptual understanding of, and causally connect,
134 hydrological changes at the landscape level, with impacts on ecological entities within the landscape.
135 These causal pathways are the basis for spatially identifying the impacted areas, and for developing
136 an appropriate mitigation response, including for extractive resource developments and water
137 extraction.

138 We have applied this approach to several regions across eastern Australia with coal and CSG
139 resource developments. Here we will focus on its application in three regions; Namoi, Maranoa–
140 Balonne–Condamine and Galilee, and subsequently discuss why the approach is transferable to
141 other regional developments that may carry a hydrological based risk to ecosystems, even those in a
142 different contextual setting with regards to data sources and existing landscape classifications.

143 The remainder of the paper is structured as follows: in the Methods section we describe the general
144 approach for achieving the classification, including descriptive examples of existing data sources. It
145 also provides a description of the three study regions in which we applied and tested the
146 classification. The Results section provides evidence of the general applicability of our approach in
147 that it shows the detailed ecological landscape classification for the three distinctively different
148 region in terms of location, topography, and climate. In the Discussion section we provide an
149 example on the use of the landscape classification. Here we describe an impact assessment in the
150 Namoi region using modelling that includes a Bayesian expert assessment approach. We also discuss
151 limitations and provide our conclusions.

152 Figure 1 provides a visual outline of the paper, giving an overview of the workflow we applied. In this
153 context the figure incorporates Methods and Results above the dashed line. Below the dashed line
154 are the Discussion parts, which include applying our classification using quantitative and qualitative
155 risk modelling in combination with surface water and groundwater modelling. Surface water and
156 groundwater modelling establish a zone of hydrological change in which impacts are likely.



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Figure 1: Visualisation of workflow for developing our ecological landscape classification (above the dashed line) and its application to develop an ecological risk assessment (below the dashed line).

160 **2 Methods**

161 In the following section, we show the development of a dataset-agnostic method to develop a
162 regional-level landscape classification that is flexible in incorporating data sources at different scales,
163 including region-specific datasets. Ecological systems are complex and work at a range of scales
164 within regions/landscapes, and they exhibit interactions and feedbacks that work across scales.
165 Consequently, there is no one scale appropriate for a subsequent analysis of ecological impacts.
166 Here we use a variable scale range that is relevant for ecological impacts of water changes from coal
167 resource developments when using an expert assessment approach. Our classification focuses on a
168 scale range (36,000 km² to 600,000 km²) that is associated with eco-hydrological linkages (and
169 associated causality) between the response of ecological components to predicted hydrological
170 changes. This scale range is what most hydrologists would consider the “regional” scale range
171 (Gleeson and Paszkowski, 2014). It provides the basis and flexibility for experts to build their
172 conceptual understanding of causal pathways and use these to assess ecological impacts with the
173 landscape classes (see also Figure 1).

174 **2.1 Study areas**

175 Our three study areas are the Namoi, the Maranoa–Balonne–Condamine and the Galilee regions in
176 eastern Australia. Each of these regions have coal resource developments within them and have
177 distinctly different landscape characteristics. They cover different state jurisdictions, or even cross
178 state jurisdictions, and range from approximately 36,000 km² to 600,000 km² in size. Consequently,
179 the classification is based on different state-based datasets. Each region’s classification relies on the
180 extent of surface water and groundwater systems that existing and potential future coal resource
181 developments in the region may impact.

182 **2.1.1 Namoi region**

183 The Namoi region covers approximately 35,700 km² in eastern Australia, is located within New South
184 Wales and forms one catchment of the Murray-Darling Basin. The long-term mean annual rainfall
185 varies from 600 to 1100 mm and potential evapotranspiration (PET) varies from 1200 to 1400 mm. It
186 contains six operational coal mines (one underground mine and five open-cut mines), nine potential
187 future coal mines and one potential CSG development. The nine potential future coal mines consist
188 of two underground, one combined open cut and underground, and seven open cut mines. The
189 region covers most of the Namoi River catchment, with the Namoi River being the main river within
190 the region. It also contains two major aquifer systems – the Namoi Alluvial aquifer and the Pilliga
191 Sandstone aquifer (Figure 2a).

192 The main land use within the region is agriculture, both dryland and irrigated cropping, and livestock
193 grazing, as well as forestry. There is also a diverse range of landscapes and ecosystems within the
194 region, including the Liverpool and Kaputar ranges, the Liverpool Plains floodplains, and Darling
195 Riverine plains in the west of the region, open box woodlands on the slopes, and temperate and
196 sub-alpine forests in the east of the region. A range of aquatic habitats occur downstream of
197 Narrabri, with large areas of anabranches and billabong wetlands. The Pilliga Nature Reserve in the
198 upper catchment of Bohena Creek, together with The Pilliga State Forest, form the largest remaining
199 area of dry sclerophyll forest west of the Great Diving Range in New South Wales (Welsh et al.,
200 2014).

201 **2.1.2 Galilee region**

202 The Galilee region covers approximately 612,300 km² and is located mostly within Queensland,
203 Australia. PET far exceeds rainfall, particularly in the summer months. Yearly rainfall ranges from 300
204 to 700 mm and PET from 2200 to 2900 mm. There are 17 proposed coal resource developments in
205 the Galilee region. These include three open-cut coal mines, two underground coal mines, five
206 combined open-cut and underground coal mines, four coal mines of currently unknown type, and
207 three CSG projects (Figure 2b).

208 The Galilee region includes the headwaters of seven major drainage catchments. These catchments
209 are Bulloo, Burdekin, Cooper Creek, Diamantina, Flinders, Paroo and Warrego. The largest of these
210 catchments within the region are the Cooper Creek and Diamantina. Groundwater within the region
211 is a very important resource, as most of the streams are ephemeral. Groundwater is used for town
212 water, agriculture and industry. Most groundwater in the region is extracted from the Great Artesian
213 Basin (Figure 2b).

214 The region covers a range of environments, including mountains of the Great Dividing Range in the
215 east, through to semi-arid and arid areas in the central and western parts of the region. The main
216 land use in the region is livestock grazing on native vegetation. There is no intensive agriculture in
217 the region, and a low human population density, largely due to the low and unpredictable rainfall
218 (Evans et al., 2014).

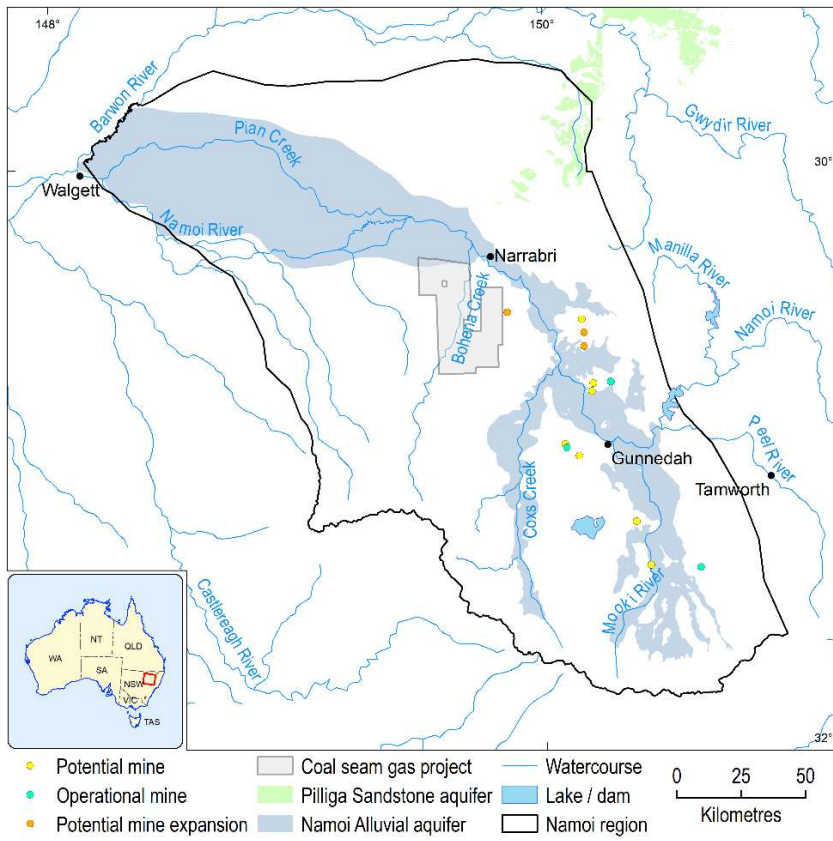
219 **2.1.3 Maranoa–Balonne–Condamine region**

220 The Maranoa–Balonne–Condamine region covers approximately 130,000 km² and is located mostly
221 within south-east Queensland with about half the area within the Murray-Darling Basin. From east
222 to west, average annual rainfall decreases from 800 mm to 420 mm, as PET increases from 1500 to
223 2370 mm. The region overlies the Surat Basin and has five open-cut coal mines and five CSG projects,
224 as well as two proposed open-cut coal mines (Figure 2c).

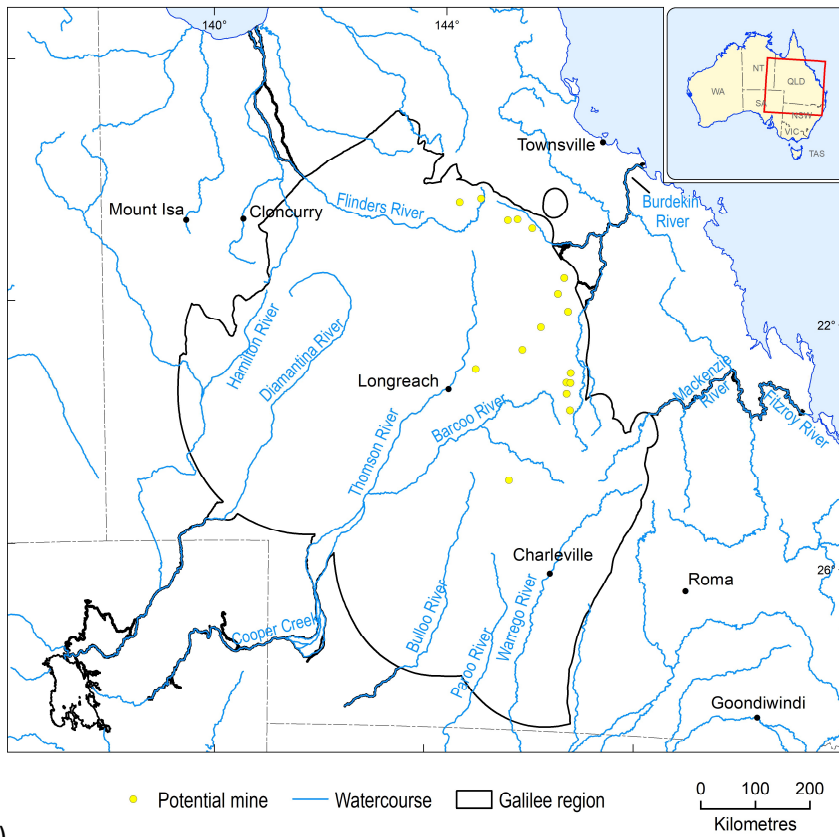
225 The region contains the headwaters of the Condamine-Balonne, Moonie, Weir, Maranoa and
226 Dawson rivers. Most of the rivers within the region are ephemeral. Groundwater is therefore an
227 important water source and is used for stock and domestic purposes, and in some cases, town water
228 supply. The Great Artesian Basin is the main source of groundwater used within the region (Welsh et
229 al., 2015).

230 The main land use within the region is grazing on natural vegetation, with dryland cropping and
231 production forestry also major land uses. The main vegetation type within the region is grassy
232 woodlands, with river red gums, coolabah and river oak common riparian species. There are also six
233 wetlands of national significance within the region: Balonne River Floodplain, Boggomoss Springs,
234 Dalrymple and Blackfellow Creeks, Lake Broadwater, Palm Tree and Robinson Creeks, and The Gums
235 Lagoon (Welsh et al., 2015).

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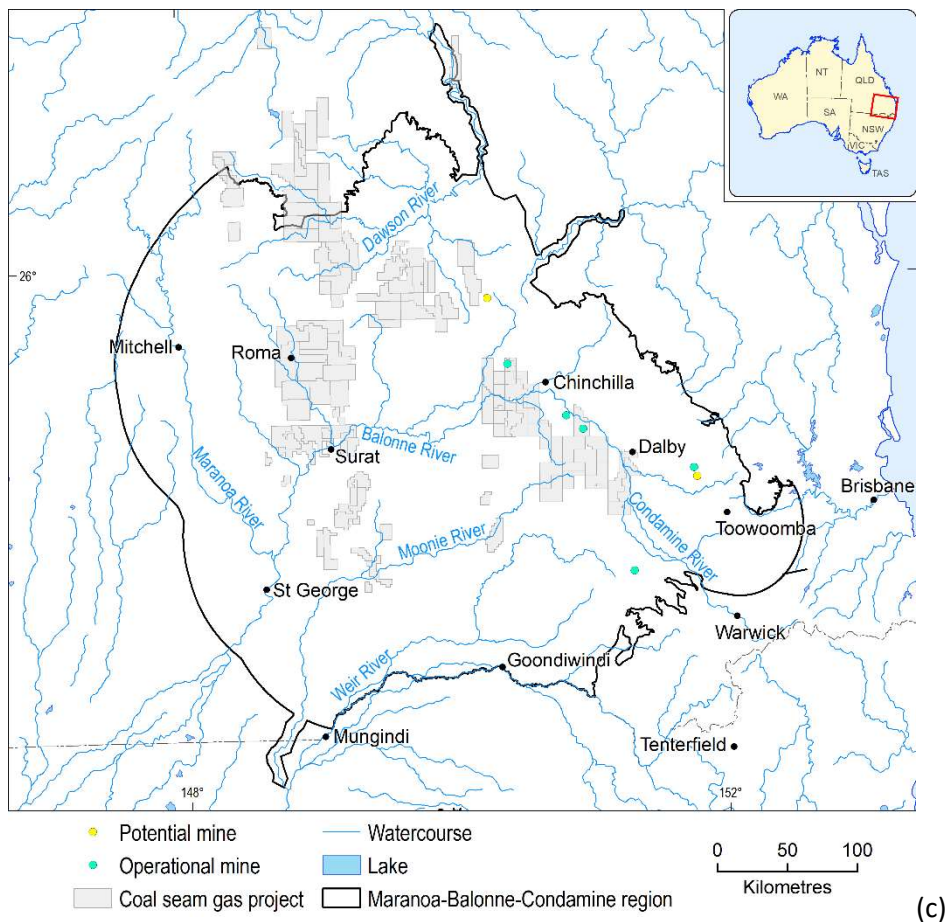


237 (a)



238 (b)

239 **Figure 2. Study areas for (a) the Namoi region (b) the Galilee region and (c) the Maranoa–Balonne–Condamine region,**
 240 **showing the potential coal resource development sites**



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Figure 2 (cont). Study areas for (c) the Maranoa–Balonne–Condamine region, showing the potential coal resource development sites

244 2.2 Landscape classification development – overview and rationale

245 The purpose of this ecohydrological landscape classification is to characterise the landscape based
 246 on patterns in land use, ecology, geomorphology and hydrology, and from these, develop landscape
 247 classes of water-dependent, remnant and human-modified features. We chose these features
 248 because these three types represent a generally applicable delineation used in our spatial dataset.
 249 For example, in Australia the word remnant vegetation (our remnant features) describes all
 250 vegetation where there was no clearing or regrowth¹ of (semi-) native vegetation, resulting in a
 251 vegetation community that resembles its predecessor’s structure. It represents areas with low to
 252 very minimal human interference. This is opposed to human-modified, where human activities are
 253 the defining features of the area, such as urban areas or other infrastructure. Water dependency is
 254 essential for establishing a conceptual linkage of water across landscape elements. Our classification
 255 employs a geographical information system to overlay existing spatial data for each region. The
 256 spatial data are the basis for categorising the landscape features using a rule-set to prioritise the
 257 spatial data based on their attribute features.

258 The datasets have a regional, state or national coverage. Using a feature-based classification helps to
 259 place the landscape classes within a common biophysical system that aids in formulating
 260 conceptual models and patterns in water dependency across the landscape of each region. This
 261 provides a classification that is aligned with the idiosyncrasies of each region. Maintaining regionality
 262 is essential when developing conceptual models and quantitative models for assessing the risk to

263 ecological components from hydrological changes. For example, arid and semi-arid regions have very
264 different ecological environments, functions and processes than subtropical or temperate
265 woodlands.

266 Our approach uses a defined rule-set and priorities, which we apply to regionally available datasets
267 to achieve a landscape classification for each of our regions. Tables 1 to 3 provide a list of citations
268 for example datasets used in this process. This is different to most other landscape classifications
269 that may use climate, topography, hydrological assessment units and remote sensing data, and apply
270 statistical dimensionality reduction and classifications such as proximity analysis (see e.g. Gharari et
271 al., 2011; Leibowitz et al., 2014; Sawicz et al., 2014; Zhang et al., 2016; Addicott et al., 2021; Carlier
272 et al., 2021; Jones et al., 2021).

273 When considering the characteristics of our regions, the following features form part of the rule-set
274 for defining landscape classes:

- 275 • broad habitat/land use type (remnant/human-modified).
276
- 277 • wetland (wetland/non-wetland)
- 278 • topography (upland/lowland, floodplain/non-floodplain)
- 279 • groundwater (groundwater dependent/non-groundwater dependent, Great Artesian Basin
280 (GAB)/non-GAB)
281 Note: identifies groundwater dependency and classifies this with the presence/absence of
282 Great Artesian basin groundwaters.
- 283 • vegetation type (riparian/woodland floodplain/grassy woodland/rainforest)
- 284 • water regime (permanent/ephemeral/null) of surface water

285 These features identify groups of landforms and use, streams and springs.

286

287 The hydrological connectivity is the main reason for developing a new classification as this allows us
288 to assess the potential impact of coal resource developments on the landscape via a water pathway.
289 Therefore the most important characteristics are the hydrological features. Describing the
290 conceptual understanding of how water connects the landscape elements allows us to identify
291 where in the landscape impacts are likely to occur. Therefore, we developed a hierarchical approach,
292 where hydrological features have priority over other landscape characteristics. This resulted in a
293 spatially complete landscape classification, where there are no gaps in the mapping data. The
294 method of prioritisation depended on region-specific characteristics and the data availability. This
295 resulted in a classification where the landscape classes have their origin in the spatial datasets, and
296 included the water dependency, which was a pre-requisite of the prioritisation. An example
297 prioritisation assigned in order of highest to lowest is:

- 298 • aquatic ecosystems (e.g. wetlands, streams and lakes)
- 299 • remnant vegetation

- other landscape components that are ‘non-remnant vegetation’ and are typically ‘human-modified’.

Subsequent use of the landscape classification for risk identification with expert input also required combining landscape classes into broader landscape groups. Landscape groups are sets of landscape classes that share hydrological properties. These landscape groups provided efficiencies in the expert elicitation process of the risk modelling, as they combined similar ecological system components based on our landscape classes while also accounting for region specific differences. For example, in the Namoi region there are two landscape groups where we do not expect any impact from coal resource developments. Firstly, the ‘Dryland remnant vegetation’ landscape group is ruled out from potential impacts because it comprises vegetation communities that are reliant on incident rainfall and local runoff and do not include features in the landscape that have potential hydrological connectivity to surface water or groundwater features. Secondly, the ‘Human-modified’ landscape group is excluded from the ecological impact assessment because it primarily comprises agricultural and urban landscapes that are highly modified by human activity. Here the impact assessment focus is on economic assets such as groundwater bores, and therefore beyond the scope of this publication.

2.2.1 Landform classification

Landform classification relied on the dominant land type of either habitat or land use to determine landscapes that are relatively natural and those that have been ‘human-modified’. Relatively intact areas are more likely to contain ecological assets, such as species and ecological communities, than highly modified areas. Location within the region (topography–upland/lowland, floodplain/non–floodplain), groundwater dependency and water regime, were part of classifying the landscape. Determining areas that are subjected to flooding, or that have persistent water, assists in identifying landscapes that support water-dependent habitat and vegetation, and aquatic ecosystems (Table 1).

2.2.2 Stream classification

Stream classification in each of the study regions was based on stream position within the catchment (e.g. upland/lowland), water regime (perennial/near permanent or ephemeral/temporary) and dependence and source of groundwater (Table 2). Catchment position is a potential indicator of stream morphology and flow patterns, while water regime is important when considering habitat suitability and physical processes within the channel and riparian zone. Streams can also gain and lose water to local and regional groundwater systems, interacting with groundwater-dependent ecosystems (Table 2).

2.2.3 Spring classification

The water source is the basis of spring classification. The source of groundwater is important when considering regional scale landscape classifications, due to the hydrological connectivity of aquifers and potential coal resource developments (Table 3).

337 **Table 1. Landform classification criteria and example datasets**

Characteristic	Classification	Example datasets
Habitat/land use	<ul style="list-style-type: none"> • Non-remnant • Remnant (and stream, Wetland) 	<p>Australian land use mapping (Australian Bureau of Agricultural and Resource Economics and Sciences, 2014) (national)</p> <p>NSW regional vegetation (NSW Office of Environment and Heritage, 2015) (regional)</p>
Topography	<ul style="list-style-type: none"> • Floodplain • Non-floodplain 	<p>NSW regional vegetation (NSW Office of Environment and Heritage, 2015) (regional)</p> <p>Namoi Valley Flood Plain Atlas 1979 (NSW Office of Environment and Heritage, 1979) (regional)</p> <p>Murray-Darling Basin aquatic ecosystem classification (Department of Sustainability, Environment, Water, Population and Communities, 2014) (regional)</p> <p>GEODATA TOPO 250K Series 3 (Geoscience Australia, 2006) (national)</p>
Groundwater	<ul style="list-style-type: none"> • Groundwater dependent (source) • Non-groundwater dependent 	<p>Namoi groundwater dependent ecosystems (NSW Office of Water, 2015) (regional)</p> <p>Queensland Groundwater Dependent Ecosystems and Shallowest Watertable Aquifer (Queensland Herbarium, Department of Science, Information Technology, Innovation and the Arts, 2015) (state)</p> <p>Queensland Groundwater Dependent Ecosystems (Queensland Department of Science, Information Technology, Innovation and the Arts, 2013) (state)</p>
Water regime	<ul style="list-style-type: none"> • Temporary • Near-permanent • Fresh • Saline 	<p>Queensland wetland data version 3 - wetland areas (Queensland Department of Science, Information Technology, Innovation and the Arts, 2012) (state)</p> <p>Queensland wetland data version 3 - wetland areas (Queensland Department of Science, Information Technology, Innovation and the Arts, 2012) (state)</p> <p>Queensland Groundwater Dependent Ecosystems and Shallowest Watertable Aquifer (Queensland Herbarium, Department of Science, Information Technology, Innovation and the Arts, 2015) (state)</p> <p>South Australian Wetlands – Groundwater Dependent Ecosystems (GDE) Classification (Sa Department for Water, 2010) (state)</p>
Vegetation	<ul style="list-style-type: none"> • Broad vegetation type 	<p>NSW regional vegetation (NSW Office of Environment and Heritage, 2015) (regional)</p>

338 NSW = New South Wales

339 **Table 2. Stream classification criteria and example datasets**

Characteristic	Classification	Example datasets
Topography	• Upland	Murray-Darling Basin aquatic ecosystem classification (Department of Sustainability, Environment, Water, Population and Communities, 2014) (regional)
	• Lowland	NSW regional vegetation (NSW Office of Environment and Heritage, 2015) (regional)
		MrVBF (CSIRO, 2000) (national)
Groundwater	• Groundwater dependent (source)	Asset database for the Namoi subregion (Bioregional Assessment Programme, 2016) (regional)
	• Non-groundwater dependent	Queensland Groundwater Dependent Ecosystems and Shallowest Watertable Aquifer (Queensland Herbarium, Department of Science, Information Technology, Innovation and the Arts, 2015) (state)
		Queensland groundwater dependent ecosystems (Queensland Department of Science, Information Technology, Innovation and the Arts, 2013) (state)
Water regime	• Perennial	Murray-Darling Basin aquatic ecosystem classification (Department of Sustainability, Environment, Water, Population and Communities, 2014) (regional)
	• Ephemeral	Geofabric Surface Cartography (Bureau of Meteorology, 2012) (national)

340

341 **Table 3. Spring classification criteria and example datasets**

Characteristic	Classification	Example datasets
Groundwater	• Groundwater dependent (source)	Asset database for the Namoi subregion (Bioregional Assessment Programme, 2016) (regional)
	• Non-groundwater dependent	Asset database for the Maranoa–Balonne–Condamine subregion (Bioregional Assessment Programme, 2015) (regional)
		Spring vents assessed for the Surat Underground Water Impact Report 2012 (Office of Groundwater Impact Assessment, 2015) (regional)

342

343 **3 Results**

344 Below we present the resulting landscape classes for the three regions. For each region, we also
 345 combined the landscape classes into landscape groups, which were specific to each region and were
 346 based on distinctions in topography, water dependency and association with GAB or non-GAB GDEs,
 347 floodplain/non-floodplain or upland/lowland environments and remnant/human-modified habitat
 348 types. The purpose of the landscape groups was to combine non-water dependent landscape classes
 349 and relate water dependent landscape classes to region specific aspects of their water dependency.
 350 This enabled experts to develop a conceptualisation of the landscape for developing their ecological
 351 impact models. While the approach in defining the landscape classes is based on a consistent rule-
 352 set and prioritisation, each of the regions has different landscape classes, which is a consequence of
 353 the differences in location, jurisdictions and available spatial datasets.

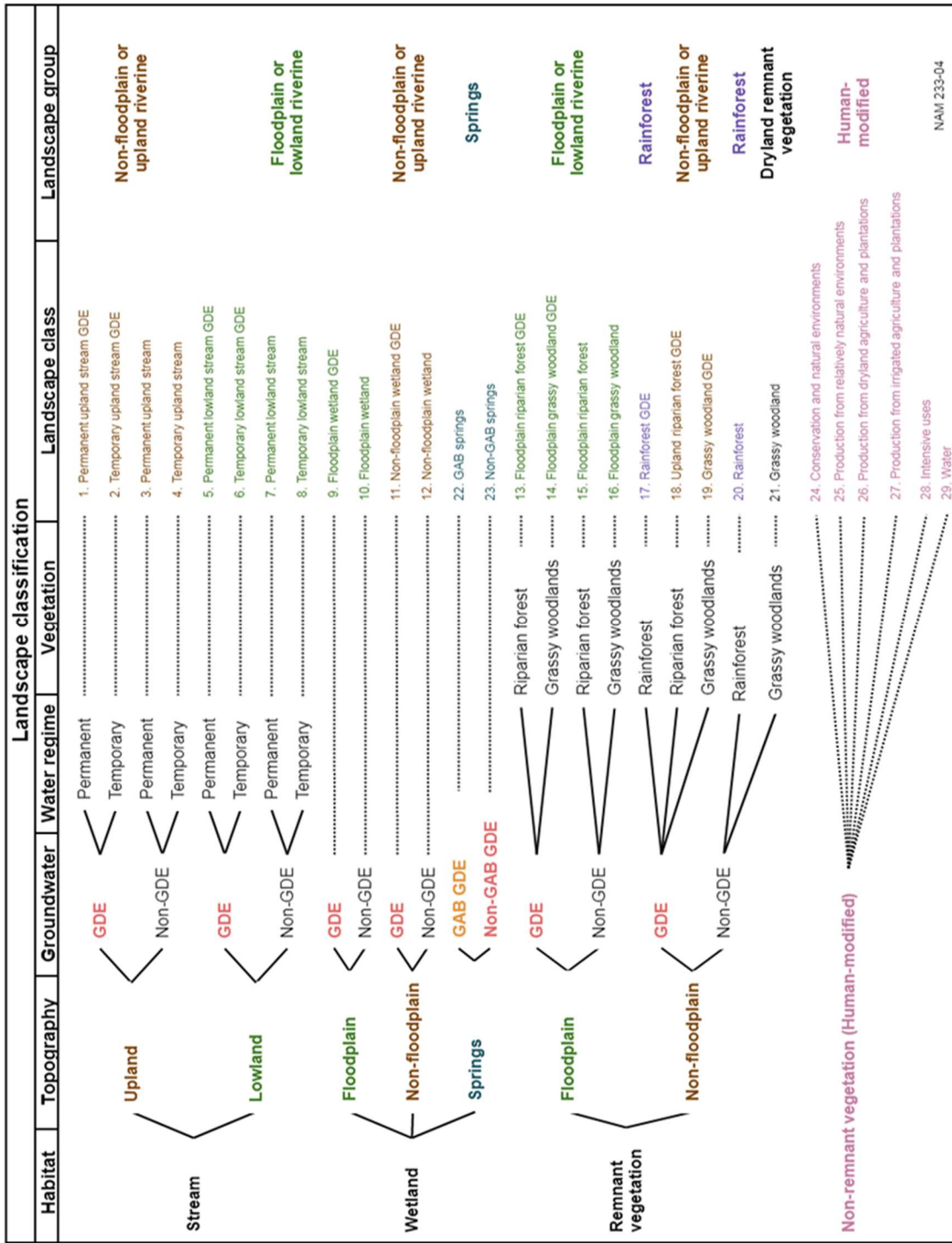
354 The rule-set deriving from the landform classification (Tables 1 to 3) and prioritisation of
355 hydrological features is the main outcome of our approach and we present the rule-set as a decision
356 pathway visually below (Figure 3). For example, for the Namoi region, the rule-set includes: (1)
357 identify the habitat (e.g. stream) (2) select by topography (e.g. upland), (3) identify the groundwater
358 associations (e.g. GDE), and so on until one derives at the final landclass level (see Figure 3).

359 **3.1 Landscape classes in the Namoi region**

360 There were 29 landscape classes within six landscape groups in the Namoi region (Figure 3). Of these
361 landscape groups, 'human-modified' (non-remnant) was the largest (59.3%; Table 4), and included
362 urban, agriculture, plantations and other intensive land uses. The dryland remnant vegetation was
363 the second largest landscape group and consisted of the grassy woodland landscape class (24.2%;
364 Table 4). This landscape class was considered non-water dependent as it did not intersect with
365 floodplain, wetland or GDE features. The rainforest landscape group was the smallest (0.5%; Table
366 4), with only a limited distribution (Figure 4a).

367 The stream network consisted of two landscape groups (floodplain or lowland riverine and non-
368 floodplain or upland riverine). The non-floodplain or upland riverine landscape group had a larger
369 proportion of stream network length (63.8%) compared to the floodplain or lowland riverine
370 landscape group (36.2%; Figure 4b). There were 22 springs identified within the Namoi region, with
371 seven of these associated with the GAB (Figure 4b).

372



373

374

Figure 3. Overview of the Namoi landscape classification schema, including criteria and attributes resulting in six landscape groups

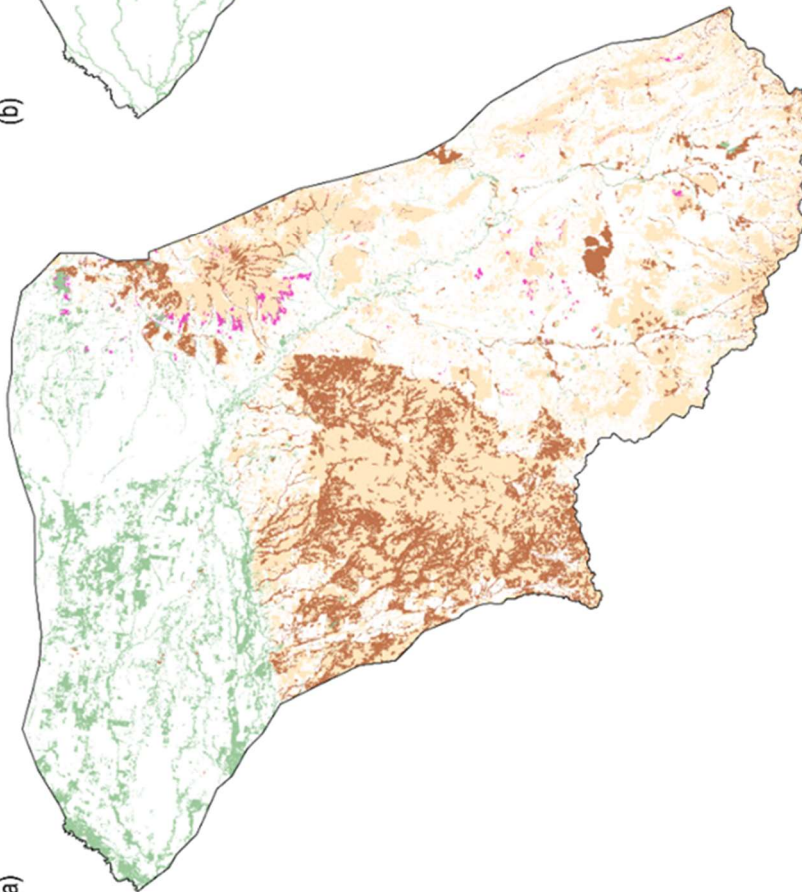
375 **Table 4. Percentage of area of each landscape group for the Namoi region**

Landscape group	Percentage of region (%)	Number of classes	Landscape classification attributes
Human-modified	59.3%	6	Conservation of natural environments, production from relatively natural environments, production from dryland agriculture and plantations, production from irrigated agriculture and plantations, intensive uses or waters for production/consumption
Dryland remnant vegetation	24.2%	1	Non-floodplain, non-GDE, grassy woodland remnant vegetation
Non-floodplain or upland riverine	9.8%	8	Upland or non-floodplain streams, wetlands or remnant vegetation
Floodplain or lowland riverine	6.2%	10	Lowland or floodplain streams, wetlands or remnant vegetation
Rainforest	0.5%	2	Non-floodplain GDE or non-GDE rainforest remnant vegetation
Springs	<0.1%	2	GAB or non-GAB springs

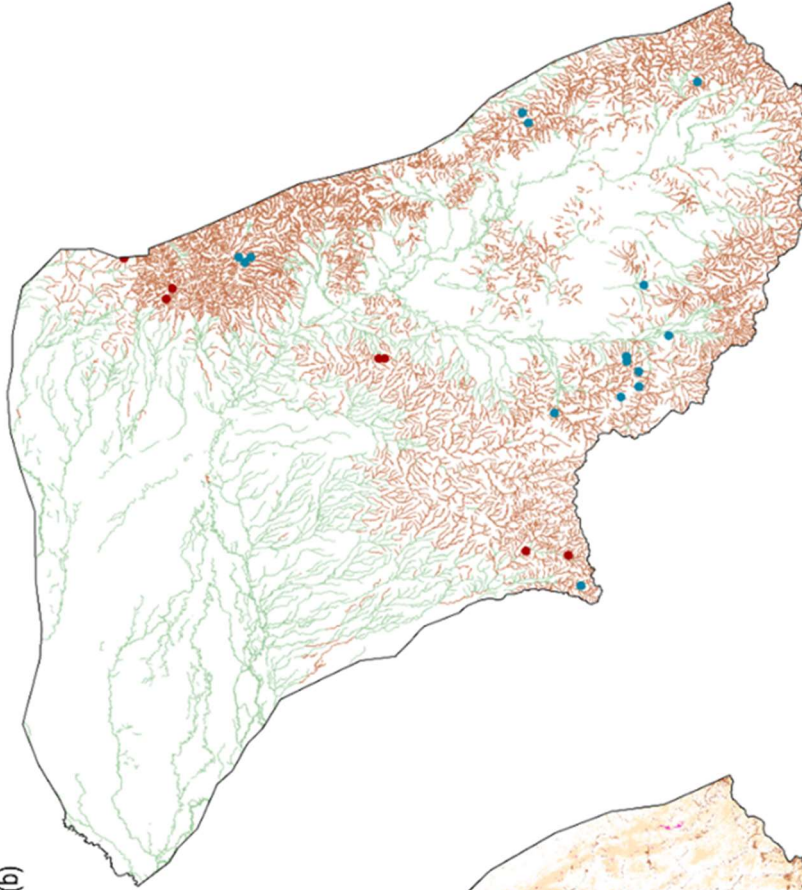
376

377

(a)



(b)



Landscape group

Floodplain or lowland riverine

Non-floodplain or upland riverine

Dryland remnant vegetation

Rainforest

GAB springs

Non-GAB springs

0 25 50
Kilometres

378
379
380

Figure 4. (a) Landscape groups (excluding the 'Human-modified'), and (b) Stream network classified as 'upland' or 'lowland' in the landscape classification and 'Springs' within the Namoi region. GAB = Great Artesian Basin
Data: Bureau of Meteorology (2012); Bioregional Assessment Programme (2017)

381 **3.2 Landscape classes in the Galilee region**

382 The Galilee region has 31 landscape classes organised into 11 landscape groups (Figure 5). The
383 dryland landscape group was the largest group within the region and the only group to have no
384 water dependency (68.5%; Table 5). The landscape groups that covered the floodplain areas were
385 the next most dominant classes, with floodplain, terrestrial GDE (12.94%; Table 5) and floodplain,
386 non-wetland (11.8%; Table 5). The remaining three non-floodplain landscape groups consisted of
387 disconnected wetlands, and terrestrial and wetland GDEs (4.9% combined; Table 5).

388 The stream network was classified as groundwater dependent or non-groundwater dependent.
389 Most of the streams in the region were non-GDEs (87.7% compared to 12.3% for the streams, GDE
390 landscape group). There were also over 3000 springs in the region.

391

392

Landscape classification						
Topography	Wetland	Groundwater	Water regime	Vegetation	Landscape class	Landscape group
<u>Vegetation</u>				Remnant vegetation	Wetland GDE, remnant vegetation	Floodplain, wetland GDE
		GDE		Non-remnant vegetation	Wetland GDE	
	Wetland		Saline	Remnant vegetation	Floodplain disconnected saline wetland, remnant vegetation	
		Disconnected		Non-remnant vegetation	Floodplain disconnected saline wetland	Floodplain, disconnected wetland
Floodplain			Non-saline	Remnant vegetation	Floodplain disconnected wetland, remnant vegetation	
	Non-wetland			Non-remnant vegetation	Floodplain disconnected wetland	
		GDE		Remnant vegetation	Terrestrial GDE, remnant vegetation	Floodplain, terrestrial GDE
		Disconnected		Non-remnant vegetation	Terrestrial GDE	
			Non-saline	Remnant vegetation	Floodplain disconnected non-wetland, remnant vegetation	Floodplain, non-wetland
		Disconnected		Non-remnant vegetation	Floodplain disconnected non-wetland	
		GDE		Remnant vegetation	Non-floodplain wetland GDE, remnant vegetation	Non-floodplain, wetland GDE
	Wetland			Non-remnant vegetation	Non-floodplain wetland GDE	
		Disconnected	Saline	Remnant vegetation	Non-floodplain disconnected saline wetland, remnant vegetation	
				Non-remnant vegetation	Non-floodplain disconnected saline wetland	Non-floodplain, disconnected wetland
Non-floodplain			Non-saline	Remnant vegetation	Non-floodplain disconnected wetland, remnant vegetation	
	Non-wetland			Non-remnant vegetation	Non-floodplain disconnected wetland	
		GDE		Remnant vegetation	Non-floodplain terrestrial GDE, remnant vegetation	Non-floodplain, terrestrial GDE
		Disconnected		Non-remnant vegetation	Non-floodplain terrestrial GDE	
<u>Stream network</u>			Non-saline	Remnant vegetation	Dryland, remnant vegetation	Dryland
				Non-remnant vegetation	Dryland	
		GDE	Temporary	Temporary upland GDE stream	Streams, GDE
	Upland		Near-permanent	Near-permanent upland GDE stream	
		Disconnected		Temporary upland stream	Streams, non-GDE
			Near-permanent	Near-permanent upland stream	
		GDE	Temporary	Temporary lowland GDE stream	Streams, GDE
	Lowland		Near-permanent	Near-permanent lowland GDE stream	
		Disconnected		Temporary lowland stream	
			Temporary	Near-permanent lowland stream	Streams, non-GDE
		Disconnected		Near-permanent lowland stream	
	Estuarine		Temporary	Temporary estuarine stream	
			Near-permanent	Near-permanent estuarine stream	
<u>Springs</u>				Springs
		GDE		Springs	

Figure 5. Landscape classification of the Galilee region

GAL-340-001

393

394 **Table 5. Percentage of area of each landscape group for the Galilee region**

Landscape group	Percentage of region (%)	Number of classes	Landscape classification attributes
Dryland	68.5%	2	Non-floodplain, non-wetland, disconnected and saline remnant or non-remnant vegetation
Floodplain, terrestrial GDE	12.9%	2	Floodplain, non-wetland and GDE remnant or non-remnant vegetation
Floodplain, non-wetland	11.8%	2	Floodplain, non-wetland, disconnected and non-saline remnant or non-remnant vegetation
Floodplain, disconnected wetland	1.1%	4	Floodplain, wetland, disconnected saline and non-saline remnant or non-remnant vegetation
Floodplain, wetland GDE	0.8%	2	Floodplain, wetland and GDE remnant or non-remnant vegetation
Non-floodplain, terrestrial GDE	3.4%	2	Non-floodplain, non-wetland and GDE remnant or non-remnant vegetation
Non-floodplain disconnected wetland	1.4%	4	Non-floodplain, wetland, disconnected, saline and non-saline remnant or non-remnant vegetation
Non-floodplain, wetland GDE	<0.1%	2	Non-floodplain, wetland and GDE remnant or non-remnant vegetation
Springs	<0.1%	1	GDE springs

395

396 **3.3 Landscape classes in the Maranoa–Balonne–Condamine region**

397 The landscape classification for the Maranoa–Balonne–Condamine resulted in 34 landscape classes
 398 within five landscape groups (Figure 6). The largest landscape group was the human-modified group
 399 (72.2%, Table 6), which included agricultural production, plantations and other intensive land uses.
 400 Of the remaining landscape groups, dryland remnant vegetation was the second most dominant
 401 (19.8%, Table 6). It was not considered water dependent, because it did not intersect with
 402 floodplain, wetland or GDE features.

403 There were three landscape groups that cover the stream network. The most dominant landscape
 404 group was floodplain or lowland riverine (including non-GAB GDEs) (47.8%), followed by non-
 405 floodplain or upland riverine (including non-GAB GDEs) (39.4%) and GAB GDEs (riverine, springs,
 406 floodplain or non-floodplain) (12.7%). There were 177 springs identified within the region. Most of
 407 the springs were GAB GDEs (riverine, springs, floodplain or non-floodplain) (86.4%, compared to
 408 13.6% for non-floodplain or upland riverine (including non-GAB GDEs)).

409
410
411
412

Landscape classification					
Topography	Groundwater	Wetland	Water regime	Landscape class	Landscape group
<u>Remnant vegetation</u>	GAB GDE Non-GAB GDE Non GDE	Wetland	Near-permanent Temporary	Floodplain GAB GDE, near-permanent wetland Floodplain GAB GDE, temporary wetland Floodplain GAB GDE	GAB GDEs (riverine, springs, floodplain, non-floodplain)
		Non-wetland	Near-permanent Temporary	Floodplain non-GAB GDE, near-permanent wetland Floodplain non-GAB GDE, temporary wetland	Floodplain or lowland riverine (including non-GAB GDEs)
		Wetland	Near-permanent Temporary	Floodplain non-GAB GDE Floodplain, near-permanent wetland Floodplain, temporary wetland Floodplain remnant vegetation	
<u>Floodplain</u>	GAB GDE Non-GAB GDE Non GDE	Wetland	Near-permanent Temporary	Non-floodplain GAB GDE, near-permanent wetland Non-floodplain GAB GDE, temporary wetland Non-floodplain GAB GDE	GAB GDEs (riverine, springs, floodplain, non-floodplain)
		Non-wetland	Near-permanent Temporary	Non-floodplain non-GAB GDE, near-permanent wetland Non-floodplain non-GAB GDE, temporary wetland	Non-floodplain or upland riverine (including non-GAB GDEs)
		Wetland	Near-permanent Temporary	Non-floodplain non-GAB GDE Non-floodplain, near-permanent wetland Non-floodplain, temporary wetland Dryland remnant vegetation	Dryland remnant vegetation
<u>Non-floodplain</u>	GAB GDE Non-GAB GDE Non GDE	Wetland	Near-permanent Temporary	Conservation and natural environments Production from relatively natural environments Production from dryland agriculture and plantations Production from irrigated agriculture and plantations Intensive uses Water	Human-modified
		Non-wetland	Near-permanent Temporary	Temporary upland GAB GDE stream Temporary upland non-GAB GDE stream Near-permanent upland stream Temporary upland stream Temporary lowland GAB GDE stream Temporary lowland non-GAB GDE stream Near-permanent lowland stream Temporary lowland stream	GAB GDEs... Non-floodplain or upland riverine (including non-GAB GDEs) GAB GDEs... Floodplain or lowland riverine (including non-GAB GDEs)
		Wetland	Near-permanent Temporary	GAB springs Non-GAB springs	GAB GDEs... Non-floodplain...
<u>Non-remnant vegetation</u>	GAB GDE Non-GAB GDE Non GDE	Wetland	Near-permanent Temporary		
		Non-wetland	Near-permanent Temporary		
		Wetland	Near-permanent Temporary		
<u>Human-modified</u>	GAB GDE Non-GAB GDE Non GDE	Wetland	Near-permanent Temporary		
		Non-wetland	Near-permanent Temporary		
		Wetland	Near-permanent Temporary		
<u>Stream network</u>	GAB GDE Non-GAB GDE Non GDE	Wetland	Near-permanent Temporary		
		Non-wetland	Near-permanent Temporary		
		Wetland	Near-permanent Temporary		
<u>Upland</u>	GAB GDE Non-GAB GDE Non GDE	Wetland	Near-permanent Temporary		
		Non-wetland	Near-permanent Temporary		
		Wetland	Near-permanent Temporary		
<u>Lowland</u>	GAB GDE Non-GAB GDE Non GDE	Wetland	Near-permanent Temporary		
		Non-wetland	Near-permanent Temporary		
		Wetland	Near-permanent Temporary		
<u>Springs</u>	GAB GDE Non-GAB GDE	Wetland	Near-permanent Temporary		
		Non-wetland	Near-permanent Temporary		
		Wetland	Near-permanent Temporary		

Figure 6. Landscape classification of the Maranoa–Balonne–Condamine region
 GAB = Great Artesian Basin, GDE = groundwater-dependent ecosystem, GAB GDEs... = GAB GDEs (riverine, springs, floodplain, non-floodplain), Non-floodplain... = Non-floodplain or upland riverine (including or non-GAB GDEs)

413 **Table 6. Percentage of area of each landscape group for the Maranoa–Balonne–Condamine region**

Landscape group	Percentage of region (%)	Number of classes	Landscape classification attributes
Human-modified	72.2%	6	Conservation of natural environments, production from relatively natural environments, production from dryland agriculture and plantations, production from irrigated agriculture and plantations, intensive uses or waters for production/consumption
Dryland remnant vegetation	19.8%	1	Non-floodplain, non-GDE, non-wetland remnant vegetation
Floodplain or lowland riverine (including non-GAB GDEs)	4.5%	9	Floodplain or lowland, non-GAB GDE or non-GDE, temporary or near-permanent wetland, non-wetland or stream
Non-floodplain or upland (including non-GAB GDEs)	2.2%	9	Non-floodplain or upland, non-GAB GDE or non-GDE, temporary or near-permanent wetland, non-wetland, stream or spring
GAB GDEs (riverine, springs, floodplain or non-floodplain)	1.3%	9	Floodplain, non-floodplain or upland GAB-GDE, temporary or near-permanent wetland, non-wetland, stream or spring

414 GAB = Great Artesian Basin, GDE = groundwater-dependent ecosystem

415 **4 Discussion**

416 In Australia, there is no consistent national classification that links ecosystems at landscape level
 417 with their underlying hydrological system. While there are many different land classifications that
 418 incorporate hydrological aspects, they do not provide linkages between hydrology and landscape
 419 elements that enable a broad scale ecological assessment of impacts associated with changes in
 420 water flow and availability, and they are not sufficiently generic for the purpose of assessing
 421 landscape level water related impacts on ecosystems in a spatially explicit manner (Kilroy et al.,
 422 2008; Elmore et al., 2003; Brown et al., 2014; Liermann et al., 2012; Doody et al., 2017; Poff et al.,
 423 2010; Aquatic Ecosystems Task Group, 2012; Olden et al., 2012; Gharari et al., 2011). However, the
 424 bioregional assessment program needed to assess impacts of coal resource developments on
 425 ecological systems via a water pathway. Hence, we needed to develop an ecological landscape
 426 classification that would be applicable to the different assessment regions.

427 While our spatially explicit landscape classification provided experts with the ability to readily
 428 identify cause and effect relationships between landscape elements and landscape hydrology, there
 429 are obvious differences between the landscape classifications in the three regions, reflecting their
 430 geographical differences (see Figure 3, Figure 5 and Figure 6). It provides the specificity that is
 431 required in a regional impact assessment, where the boundaries are based on a combination of
 432 geology, water resources and administrative conditions. The regionality also means that there is
 433 need for different datasets describing the landscape features that would not be available from a
 434 classification covering the whole of Australia.

435 Nevertheless, each landscape classification provides a typology with an explicit connection of water
 436 to the landscape class. This connection enables a causal link between hydrological change and
 437 impact to ecosystems represented by landscape classes. The causal linkage is dependent on (i) a

438 spatially explicit connection between water in the landscape and the landscape classes, (ii)
439 conceptual understanding how changes in water may result in a reaction of specific ecosystem
440 elements in the landscape class and/or landscape group and (iii) a way of modelling quantitative
441 changes in ecosystem elements related to changes in water. Our ecohydrological classification
442 approach for landscapes provides this spatially explicit connection and has implicit ecohydrological
443 elements that foster the conceptual understanding of the causal linkage. For example, spatially
444 modelling groundwater level drawdown enables a prediction on which landscape elements classified
445 as springs may be experiencing impacts from water extraction and, with additional ecological
446 modelling, by how much and when.

447 Subsequent ecological modelling using expert elicitation of potential impacts drew heavily on our
448 classification, which is based on a consistent rule-set and fosters conceptual understanding of
449 landscape processes and functions. It provides an essential framework for experts to understand and
450 conceptualise how modelled future hydrological changes from coal resource developments link to
451 potential ecological changes at the landscape level. It allows the incorporation of different data
452 sources and existing classification schemes. This consistency also makes the classification
453 development transparent, repeatable, and adjustable, should new data become available.

454 **4.1 Application of the landscape classification based impact assessment**

455 Here we show an application of how our classification approach can be used to assess the potential
456 impact coal resource developments have on ecology using the Namoi region as an example,
457 demonstrating the useability of our classification approach.

458 The purpose of developing the landscape classification was to assess the risk of coal resource
459 development on the ecology of a region via a water pathway. Our landscape classification provided
460 the spatial framework on which experts can base their assessment of risk from coal resource
461 development on the ecology of a region via a water pathway. Details of the predicted changes in
462 surface water and groundwater for the Namoi and Galilee regions are in Post et al. (2020). Here, we
463 demonstrate the assessment of potential ecological impacts using the Namoi region. For full details
464 of the analyses in each of the three regions see Holland et al. (2017); Herr et al. (2018b); and Lewis
465 et al. (2018). This work included expert assessment of ecological risk to ecosystem components
466 based on conceptual models. Hence, the models needed to identify water mitigated linkages
467 between hydrological changes, ecosystem components and the landscape classes. We briefly
468 describe the expert assessment approach in a 3-step process below. For details we direct the reader
469 to the above references and those listed below.

470 The following describes an application of the landscape classification (see also Figure 1), and in doing
471 so we demonstrate that it is a fit-for-purpose in the context of assessing potential ecological impact
472 resulting from potential surface water and groundwater changes at different locations within the
473 landscape. The 3-step process illustrates the utility of our landscape classification approach for
474 assessing the risk to ecosystems in the landscape classes and groups. The process included experts
475 identifying risk to landscape classes using their knowledge on local ecosystems. Specifically, the
476 experts used the broad landscape groups and their underlying hydrogeological features to develop
477 qualitative models initially that then fed into building quantitative models. Here the experts used
478 outputs from surface water and groundwater modelling to determine the potential changes in water
479 and what this may mean for ecological entities within the landscape classes and/or groups. These

480 models assessed the future hydrological changes and risks to the ecosystems in the landscape
481 groups (see also Figure 1).

482 The 3 step process included:

483 **Step 1:** Develop qualitative models to conceptualise and prioritise ecosystem components of each
484 landscape class and their linkage to hydrological variables.

485 Here we use the example of the upland riverine landscape class. A qualitative model for the upland
486 riverine landscape class agreed with the existing understanding that a reduction in overbank flows
487 and lowering of the water table resulted in a reduction in several ecosystem components, including
488 riparian habitat, amphibians and fish, and an increase in fine particulate matter, dissolved organic
489 matter and cyanobacteria (Holland et al., 2017; Herr et al., 2018b; Hosack et al., 2018). A qualitative
490 model has, at its basis, the conceptual understanding of ecosystem components and the direction of
491 their interactions, that is a positive, negative, or neutral influence of one component on another.
492 This understanding also incorporates feedback loops between the ecosystem components in the
493 form of sign directed graphs, and it enables time intensive quantitative model development to be
494 directed at variables with the highest importance. The method is based on a matrix level analysis of
495 the component interactions (see for example Herr et al., 2016; Ickowicz et al., 2018).

496 The hydrological variables, and relationships between ecosystem components that the qualitative
497 modelling process prioritised for upland riverine systems in the Namoi region were the
498 macroinvertebrate responses to riverine system change, presence of tadpoles and changes in
499 projected foliage cover in the riparian trees along the stream channel (Table 7).

500 **Step 2:** Use qualitative model priorities to develop quantitative models.

501 In this context, qualitative models highlighted critical relationships and variables that became the
502 focus of the quantitative models (see for example Herr et al., 2016; Hosack et al., 2018; Ickowicz et
503 al., 2018). The focus of the quantitative models was on three elements within the upland riverine
504 landscape classes (Table 7): (i) the response of upland riparian trees to changes in groundwater; (ii)
505 macroinvertebrate assemblage changes related to days with no consecutive water (zero-flow days)
506 and the longest zero flow event period; and (iii) the response of tadpoles to zero flow days and
507 longest zero flow event period. Table 7 provides a brief summary of these variables; specific details
508 of the variable definitions are in Ickowicz et al. (2018).

509

510

511 **Table 7: Upland riverine ecosystem quantitative modelling variables that experts prioritised in the qualitative model and**
 512 **associated ecological and hydrological variables used in the development of the quantitative impact model (after**
 513 **Ickowicz et al., 2018)**

Expert prioritised relationship	Ecological variable (with associated sample units)	Hydrological variable
Response of the upland riparian forest to changes in hydrological regime and groundwater	Annual mean projected foliage cover of species group that includes: Casuarina, yellow box, Blakely's red gum, <i>Acacia salicina</i> , <i>Angophora floribunda</i> , grey box. Transect of 50 m length and 20 m width that extends from first bench ('toe') on both sides of stream	<ul style="list-style-type: none"> • The mean annual number of events with a peak daily flow exceeding the overbank flow events. • Maximum difference in drawdown under a baseline and under the expected drawdown • The year with the maximum difference in drawdown relative to the baseline
Response of fast-water macroinvertebrates to changes in number of zero-flow days and maximum zero-flow event	Average number of families of aquatic macroinvertebrates in riffle habitat sampled using the NSW AUSRIVAS method for riffles	<ul style="list-style-type: none"> • The number of zero-flow days per year, averaged over a 30-year period. • The maximum length of spells (in days per year) with zero flow, averaged over a 30-year period.
Response of tadpoles to changes in number of zero-flow days and maximum zero-flow event	Probability of presence of tadpoles from <i>Limnodynastes</i> genus (species <i>dumerilii</i> , <i>salmini</i> , <i>interioris</i> and <i>terraereginae</i>) sampled using standard 30 cm dip net	<ul style="list-style-type: none"> • The number of zero-flow days per year, averaged over a 30-year period. • The maximum length of spells (in days per year) with zero low, averaged over a 30-year period.

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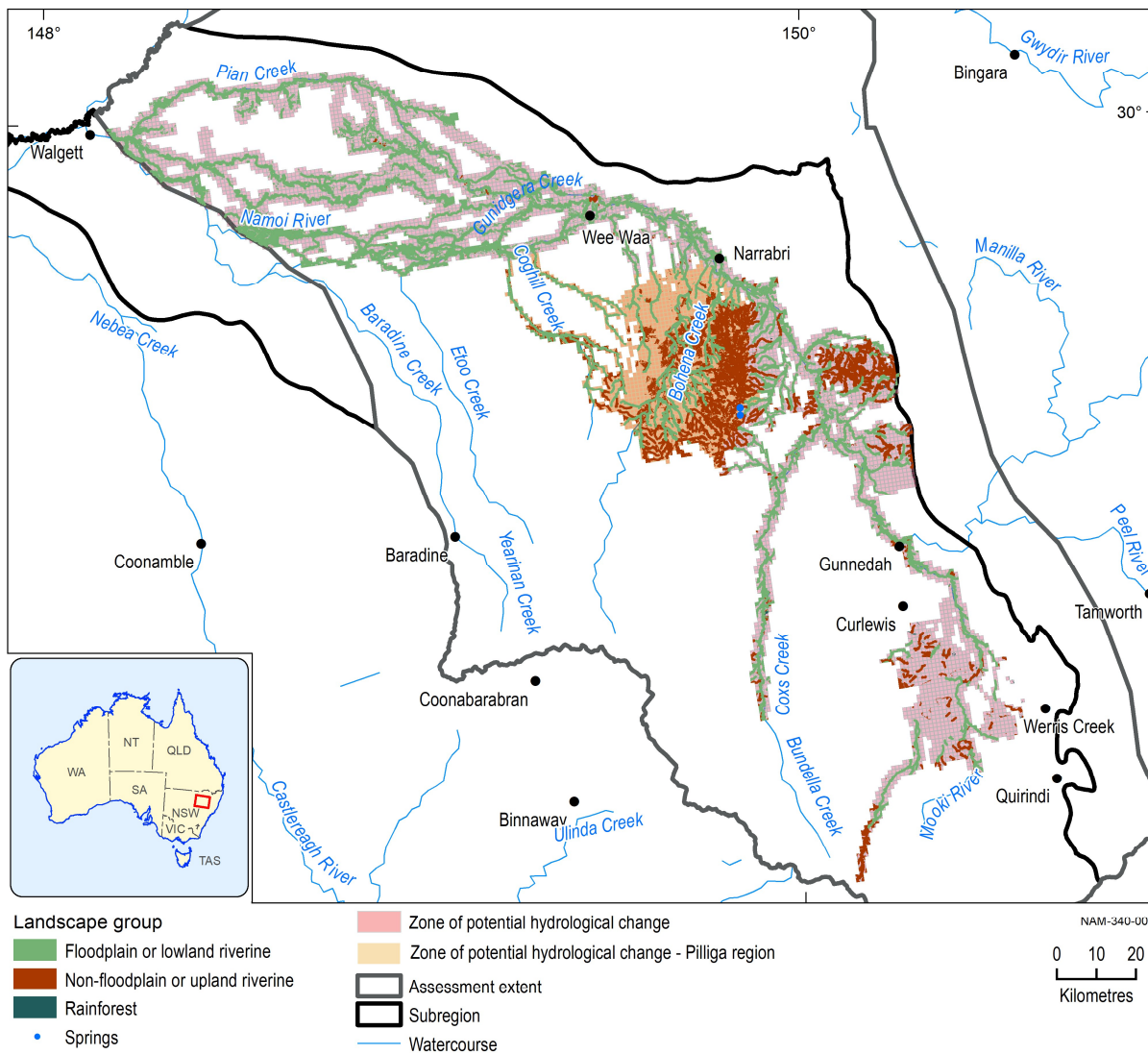
515 **Step 3:** Identify risk areas in the regions where quantitative modelling indicated significant changes
516 to landscape group components.

517 This quantitative modelling approach incorporated expert elicitation in a Bayesian framework to
518 predict changes in ecological system components because of expected changes in hydrology
519 conditions. The method dealt with complexity and limited knowledge that allows for updating with
520 new information, which is an important feature in evidence-based decision making (see for example
521 Hosack et al., 2017).

522 The modelling of risk to ecosystems at regional scale focuses on recognising which parts of the
523 region are potentially impacted and which parts are unlikely to experience harm. Using our
524 landscape classification as a crucial input, the modelling delineated impacted areas within each
525 region, based on a zone of potential hydrological change. This is the area in the landscape, where
526 hydrological modelling identified an expected change to surface water and groundwater from future
527 resource extraction. Risk levels across a landscape group are a result of aggregating individual risks
528 associated with each ecological variable and categorising the risks into three levels based on their
529 percentile spreads (for details see Herr et al., 2018b).

530 For the Namoi region, for example, dryland remnant vegetation, human-modified ecosystems, no-
531 floodplain and upland riverine ecosystems and rainforests, will not experience impacts, while
532 floodplain and lowland ecosystems area and streams of floodplain and lowland ecosystems will
533 potentially experience impacts (Herr et al., 2018a). Figure 7 (a) shows the landscape groups that are
534 at risk of impact from hydrological changes as they are situated within the zone of potential
535 hydrological change, and Figure 7 (b) shows the risk level to these landscape groups from the
536 quantitative models. Note that there is a category “Remaining unquantified ‘floodplain and lowland
537 riverine’ classes”. The expert could not develop quantitative models for these classes, because there
538 was no surface water hydrological model available that could predict changes to surface water flows.
539 This was related to the lack of gauging data and groundwater interaction details specific to the
540 lowland drainage channels. Having lowland riverine classes whose risk remains unquantified means
541 there is additional work needed before an assessment and potential mitigation of impacts from
542 hydrological changes is possible (Herr et al., 2018b).

543

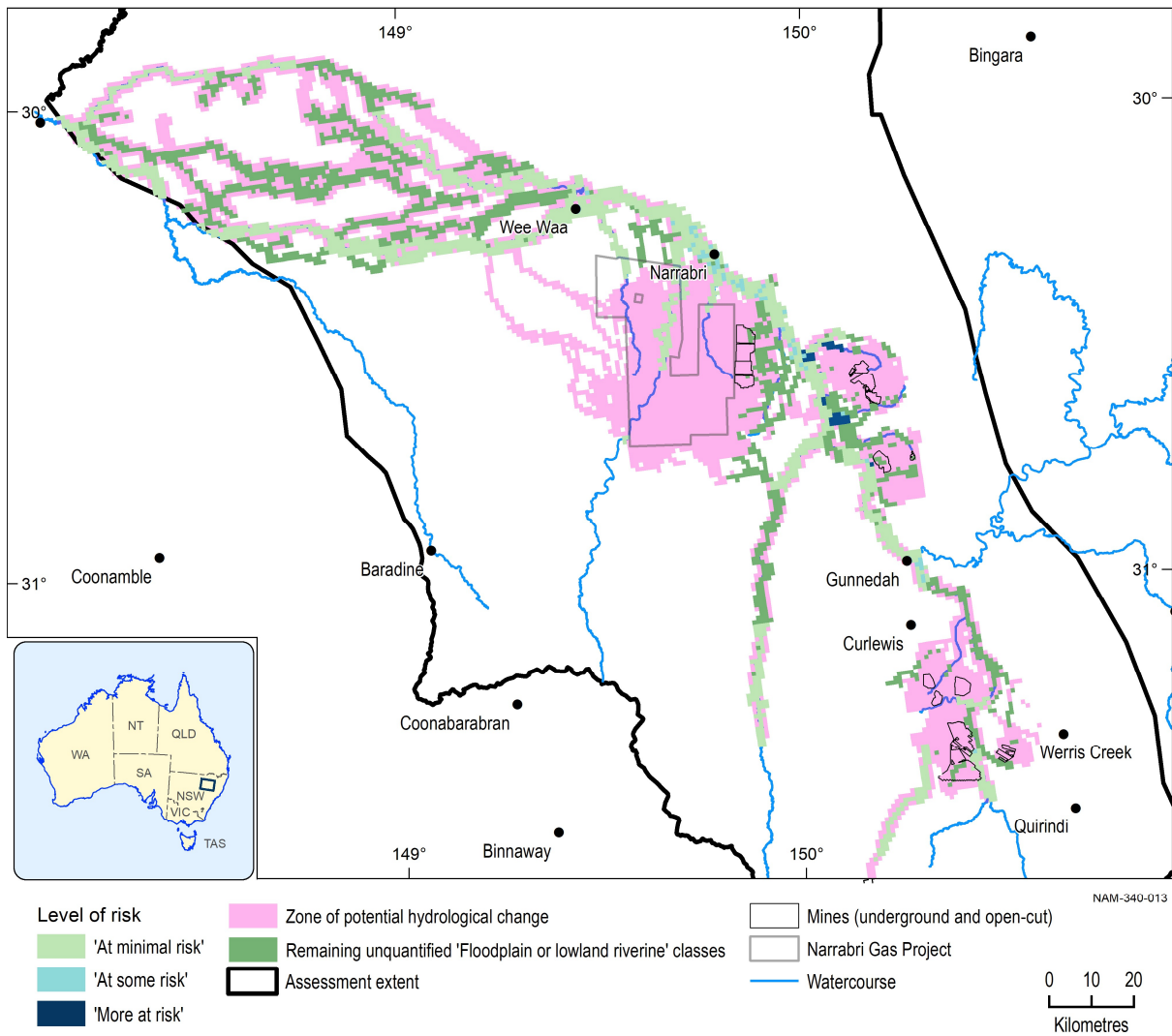


544

545 **Figure 7a: Landscape classes overlaying areas of potential hydrological change (Herr et al., 2018b)**

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547



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549

550

Figure 7b: Hydrological change risk level of lowland landscape class areas (Herr et al., 2018b)

551 **4.2 Limitations**

552 While the ecohydrological landscape classification approach provided the basis for the risk
553 assessment outlined above, there are some limitations that require consideration when attempting
554 to develop and apply this ecohydrological landscape classification approach.

555 An important issue for the landscape classification is formulating a typology that adequately reflects
556 both the functional and structural complexity of the ecosystem, while delivering a succinct and
557 consistent representation of the system that is 'fit for purpose' to assign hydrological connectivity
558 between the landscape classes, and within the general landscape. The systematic classification
559 imposes discrete boundaries among landscape components that may not adequately capture
560 gradients within and across landscape classes. This approach tends to simplify important
561 components of ecotones such as 'transition' zones or edges between landscape classes, where
562 ecosystem processes and/or biodiversity are likely to peak. If landscape classes are treated purely as
563 'closed' ecosystems, then the result may be a poor representation of the biotic interactions and
564 energy exchange between adjacent systems, and this could limit a conventional impact and risk
565 analyses. These conceptual challenges may be important considerations for subsequent impact
566 assessments, requiring special attention in assigning risk from human induced changes in hydrology.
567 However, conceptual modelling of impacts may be able to compensate for this shortfall, when for
568 example, incorporating riparian areas within riverine and wetland model development.

569 There are also spatial data issues that require additional consideration beyond just simply
570 incorporating existing data. There are several technical issues that constitute important gaps in the
571 landscape classification for the Namoi region, for example. Here, two different approaches to define
572 GDEs were required because one spatial dataset only included terrestrial vegetation and not riverine
573 systems mapped within the stream network (NSW Office of Water, 2015). A second GDE dataset
574 helped overcome this deficiency, and provided the basis to classify the stream network's
575 dependency on groundwater (Bioregional Assessment Programme, 2012).

576 Wetlands in large areas of Australia are not yet adequately mapped. The separation between
577 groundwater-dependent and surface water-dependent wetlands may not always be accurate. In
578 many areas there is little knowledge of groundwater – surface water interactions. There is also a
579 significant gap in the understanding of water thresholds for ecosystems associated with springs. In
580 part, this results from a lack of bores to provide meaningful groundwater data. Some examples of
581 these data gaps appear in the discussion of the functioning of springs in the Doongmabulla Springs
582 complex in the Galilee region, particularly in identifying the source aquifer (Fensham et al., 2016).

583 There is extensive work from Queensland that links regional ecosystems vegetation to their
584 groundwater needs, although the mapped areas are still small (Sattler and Williams, 1999;
585 Queensland Government, Queensland, 2016; Queensland Herbarium, 2021). However, in many parts
586 of Australia, GDE mapping and classification approaches are limited, and many areas lack systematic
587 ground-truthing. This is especially true in areas with extensive intact native vegetation remnants,
588 such as the Pilliga Forest of the Namoi region, where large areas of 'Grass woodland GDE' landscape
589 class exist, but the lack of published studies on vegetation–groundwater interactions limits a
590 definition of the nature of this interaction.

591 4.3 Conclusions

592 We showed that our landscape classification approach worked in the three geographically different
593 regions, with widely disparate information sources that fed into a landscape classification. This also
594 makes the approach resource efficient where existing spatial landscape or ecosystem classification
595 schemes, developed for other purposes, can be incorporated into the classification.

596 The study was able to formulate and implement an attribute-based classification scheme to define
597 and delineate water-dependent features across three large regions. We conclude that this approach
598 allowed us to repurpose several existing schemas into an adaptable and practical typology of a
599 landscape classification. The conceptual framework of landscape ecohydrology forms the basis for
600 this classification, which is used to focus subsequent analysis of potential cumulative impacts on
601 water resources from multiple coal resource developments. The classification enabled the
602 development of specific conceptual and qualitative models that linked changes in hydrology to
603 potential impacts on ecosystems using the landscape classes. The classification provided crucial
604 inputs for a risk analysis of landscape components subjected to hydrological changes.

605 Applying our approach to different regions showed that it is sufficiently general and flexible to
606 enable the development of ecohydrological classifications in regions in Australia and potentially in
607 other regions around the globe, given a sufficiently mature information base and data availability.

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869 **6 Author contributions**

870 AH, LM undertook the original draft preparation. All authors contributed to review and editing,
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872 **7 Competing interests**

873 The authors declare that they have no conflict of interest.

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