

A UAV- and field-based investigation of the land degradation and soil erosion at opencast coal mine dumps after 5 years' evolution of natural processes

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1 **A UAV- and field-based investigation of the land degradation and soil**
2 **erosion at opencast coal mine dumps after 5 years' evolution of natural**
3 **processes**

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11 **Abstract:**

12 Open-pit coal mining has a large impact on land surface, both at the mining pits themselves
13 and at waste sites. After artificial management is stopped, a reclaimed opencast coal mine dump is
14 affected by wind and water erosion from natural processes, resulting in land degradation and even
15 safety incidents. In this paper, the soil erosion and land degradation after 5 years of such natural
16 processes, at the Xilinhote open pit coal mine dump in Inner Mongolia, were investigated. A multi-
17 source data acquisition method was applied: the vegetation coverage index was extracted from GF-
18 1 satellite imagery, high-precision terrain characteristics and the location and degree of soil erosion
19 were obtained using an Unmanned Aerial Vehicle (UAV), and the physical properties of the topsoil
20 were obtained by field sampling. On this basis, the degree and spatial distribution of erosion cracks
21 were identified, and the causes of soil erosion and land degradation were analyzed using a
22 geographical detector. The results show that: 1) The multi-source data acquisition method can
23 provide effective basic data for the quantitative evaluation of the ecological environment at dumps;
24 2) slope aspect and vegetation fractional coverage are the main factors affecting the degree of
25 degradation and soil erosion. Based on this analysis, several countermeasures are proposed to
26 mitigate land degradation: 1) The windward slope be designed to imitate the natural landform; 2)
27 engineering measures should be applied at the slope to restrain soil erosion; 3) pioneer plants should
28 be widely planted on the platform at the early stage of reclamation.

29

30 **Keywords:** open-pit mining, dumps, ecological restoration, UAV, soil erosion, ecosystem stability,
31 geographical detector

32 **1. Introduction**

33 Surface mining is the most widely used mining method in the world. In the United States,
34 Germany, Australia, Russia and other developed countries, the proportion of opencast mining
35 reached more than 60%. Although surface mining has many advantages, such as high safety, high
36 mining efficiency, high recovery rate, low cost and etc. (Kennedy, 1990), it drastically damages the
37 eco-environment and produces a large amount of coal waste. The dump formed by the resulting coal
38 waste stack takes up a large area of land, where the surface soil is poor and it is pressed by heavy
39 trucks, resulting in problems for plant roots and an increase in surface runoff. During the rainy
40 season, concentrated precipitation flows into the settlement cracks, causing disasters such as col-
41 lapse, landslide and debris flow, and seriously threatening the lives of residents nearby (Neugirg et
42 al., 2016). Statistically, the land used for open-pit mine dumps accounts for 30%–50% of the total
43 land used for mining. As a large coal mining country, China's raw coal production accounts for more
44 than 70% of its total energy production (National Bureau of Statistics, 2017). There are a large
45 number of shallow coal seams in western China. The mining conditions are good and the coal quality
46 is excellent. The center of coal mining is predicted to move gradually westwards (Guo et al., 2018).
47 The proportion of open-pit production also increase gradually, from about 3.3% in 1998 to about
48 15% in 2015 (Huang et al., 2015) and the annual output of coal is expected to reach 3.9×10^9 tons by
49 2020. The average annual additional land area occupied is up to 1×10^4 hm² (10^8 m²), and the soil
50 erosion area increases annually by 3×10^3 hm² (3×10^7 m²) (Qi, 2017). Therefore, it is important to
51 take account of the safety and stability of the dump when assessing the production safety and eco-
52 nomic benefits of the open mining area. The monitoring and maintenance of the dump has become
53 an important aspect of the reclamation process of the open mining area.

54 The dump is mainly composed of waste material from open-pit mining. After the dump reaches
55 its full capacity, it is covered with a layer of topsoil to ensure a rapid restoration of vegetation. The
56 construction of the dump needs to be adapted to local conditions, and the appropriate disposal
57 method should be selected according to the geomorphologic characteristics of the mining area. Sev-
58 eral types can be distinguished, such as conical waste dump, piles type of shape, board shape, shape
59 of the terrace, shape of slope, ridge shape, straight shape, or they may take the form of flat coverings.
60 (Petra et al., 2015). Because of the high cost of land acquisition in China, flat coverings are rarely
61 used in filling the dumps. The terraced landform of the Loess plateau can be imitated by designing
62 terraced dumps with a relative height difference of 100 m–150 m between the platform and the slope
63 intersect (Fig. 1a). Alternatively, in hill and gully regions of the Loess plateau, fly ash and gangue
64 can be used to carry out a mixed ecological filling of the open pit and gully (Fig. 1b), laying a
65 foundation for subsequent ecological restoration (Shanxi provincial bureau of quality and technical
66 supervision, 2016). However, in the eastern grassland area of northern China, the terrain is flat, there
67 is an insufficient supply of topsoil, and the precipitation can be highly intense. In order to reduce
68 the amount of land they occupy, most dumps adopt a ladder distribution (Liu et al., 2011).

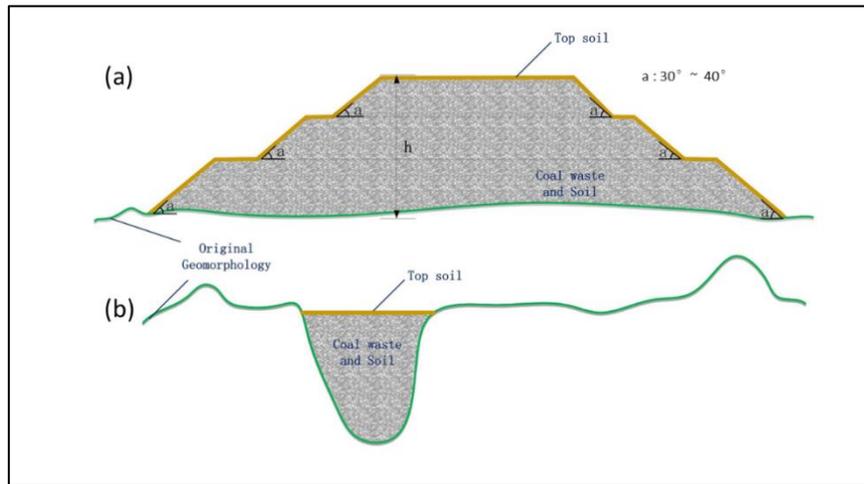


Fig. 1 Waste dump diagram. (a) Multi-stage dump. (b) Gully filled dump.

69

70

71 In view of a series of environmental problems and safety problems caused by surface mining,
 72 the land reclamation in mining areas has been carried out in many countries. However, in order to
 73 restore the mining eco-system to a stable top community as soon as possible, post-restoration mon-
 74 itoring is still needed. The monitoring of mine restoration mainly includes vegetation restoration,
 75 soil quality and soil erosion. Vegetation restoration is the key to the restoration of dump sites. Veg-
 76 etation restoration can make full use of the function of the soil-plant composite system, improve the
 77 local environment and promote a regional ecological balance (Srinivasan Madhusudan et al., 2015),
 78 and also significantly improve the soil bulk density, soil moisture content and soil porosity (Wang
 79 et al., 2016). Through the calculation of VFC, the vegetation growth status of the dump can be
 80 effectively quantified, serving as an important reference for soil erosion. This should give priority
 81 to local species with a high survival rate, strong resistance and an ability to improve the physical
 82 and chemical properties of the soil. The ideal procedure is to gradually strengthen the composition,
 83 structure and level of vegetation, improve the function of the soil and its water conservation and
 84 self-renewal ability, and finally to achieve self-balance, establishing a complete and stable ecosys-
 85 tem (Wang et al., 2014). At present, in order to improve the success rate and greening rate of recla-
 86 mation over a short period of time, most of dump sites adopt artificial management. Although a
 87 large area of vegetation coverage can be achieved in a short period of time, it is difficult to achieve
 88 a state of self-balance. The monitoring and analysis of vegetation degradation and soil erosion of
 89 typical dumps under natural conditions can improve the reclamation method and help dumps to
 90 reach ecological balance more quickly. However, there is a lack of research on ecological and envi-
 91 ronmental monitoring, or disaster monitoring, of the development of dumps under natural conditions
 92 (i.e., without artificial management and protection). In addition to this, the quality of the overlying
 93 soil layer (Khalil et al., 2014) affects the restoration of vegetation at the dump. The pore structure
 94 of the soil affects its hydraulic properties and associated water flow (Wang et al., 2014; Bao et al.,
 95 2017). The amount of soil organic carbon (SOC) (Fettweis et al., 2005; Yuan et al., 2018; Zhao et al.,
 96 2012), nitrogen (N) (Ye et al., 2017), phosphorus (P) (Liu et al., 2017) and potassium (K) (Zhou
 97 et al., 2017) present in the soil are major determining factors and indicators of its fertility and quality,

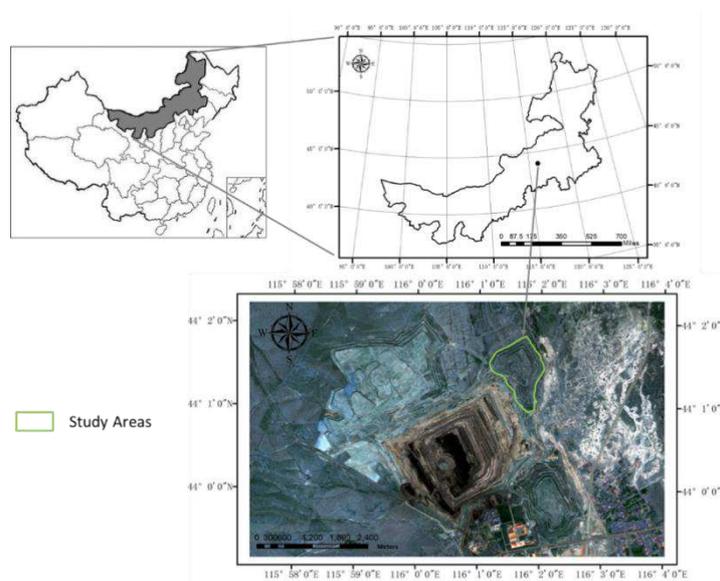
98 which are closely related to vegetation restoration. Heavy metals in a soil not only affect plant de-
99 velopment but also threaten the health of the surrounding inhabitants through the food chain (Zhao
100 et al., 2010). Vegetation drought resistance (Zhao et al., 2008) and water and soil conservation (Scor-
101 pio et al., 2018) can reduce soil erosion. Nevertheless, the high precipitation intensity (Neugirg et
102 al., 2016) in the rainy season causes the soil moisture to reach saturation, and plants and top soil can
103 easily be carried away due to heavy rainfall (Sahle et al., 2019). Inhomogeneous sedimentation
104 (Tomas et al., 2014), and the gradient (Lü et al., 2013) and aspect (Nyssen et al., 2010) of the dump
105 slope also have an important effect on soil erosion. Erosion gullies and surface collapses caused by
106 soil erosion lead to landslide, debris flow and other disasters, so it is very important to monitor soil
107 quality, vegetation and soil erosion at the dump.

108 Field measurement is often used in traditional monitoring methods (Jan et al., 2006; Martin
109 Duque et al., 2015; Nyssen, 2006b). With the development of geomechanics, some experts began to
110 use mechanical models to estimate the development of soil erosion (Garry et al., 2006). Considering
111 the natural factors affecting soil erosion, some scholars use the soil erosion equation to conduct
112 quantitative research on soil erosion (Wischmeier and Smith, 1978). Satellite remote sensing (Schro-
113 eter and Glaesser, 2011; Obade and Lal, 2013) and UAV remote sensing (Xiao et al., 2017; Ren et
114 al., 2020; Zhao et al., 2020) are also gradually applied to vegetation inversion and soil erosion mon-
115 itoring in mining areas, it makes multi-frequency and high-spatial-resolution monitoring possible.
116 At present, UAV is mainly used for small-scale topographic surveys for mine monitoring (Kršák et
117 al., 2016) and for large scale geomorphic feature characterization (Chen et al., 2015; Ren et al.,
118 2019). Most studies focus on measuring volume changes at different time scales (Messinger and
119 Silman, 2016; Siebert and Teizer, 2014) and changes in mine land use (Gui et al., 2008), and there
120 is little research carried out on mine dumps. In addition, there is a lack of systematic research on the
121 evolution of a dump, and the associated degradation mechanisms, under the influence of multiple
122 factors. The causes of water erosion (Neugirg et al., 2016) and wind erosion (Yang, 2016) are closely
123 related to topography, vegetation, soil, and slope aspect and gradient. The Geographical Detector
124 (Cao et al., 2013), based on spatial stratified heterogeneity, can be used to analyze the relationship
125 between variables.

126 By combining UAV remote sensing, satellite remote sensing and field investigation, the main
127 outcomes of this paper are as follows: 1) the acquisition of accurate topography and aetiology (ero-
128 sion and collapse) data of the dump; 2) a calculation of the vegetation fractional coverage (VFC) of
129 the dump and a sampling of the physical properties of the soil; 3) assessment of the importance of
130 different soil erosion factors, both individually and in combination, using the geographical detector
131 (spatial correlation analysis); 4) according to the main driving factors, some suggestions are put
132 forward to alleviate the soil erosion in the dump. It provides a reference for the land reclamation of
133 open pit mine dump.

134 **2. Study area**

135 The experimental plot is located at the Western One site of the Shengli coal field, with geo-
 136 graphic coordinates of 115° 58'–116° 2' 50" E, 43° 59'–44° 2' 15"N (Fig. 2), located in Xilinhot,
 137 Inner Mongolia Autonomous Region, which is the largest prairie area in China. The mining area has
 138 a typical temperate, semi-arid, continental monsoon climate, where the annual average temperature
 139 is 0.3 °C and the average wind speed is above 8.5 m/s. The total annual precipitation averages 294.9
 140 mm, with more than 70% occurring from June to August. The average annual effective evaporation
 141 is 1811.3 mm, which is more than six times greater than the amount of precipitation. The coal mine
 142 has a coal-bearing area of 342 km², with proven reserves of 15.932 billion tons, most of which are
 143 suitable for open-pit mining. There are three main external dumping sites, namely south dump, north
 144 dump and side dump. All the three waste dumps had been reclaimed, with a green area of 8.64 km².



145

146

Fig. 2 The study area location.

147

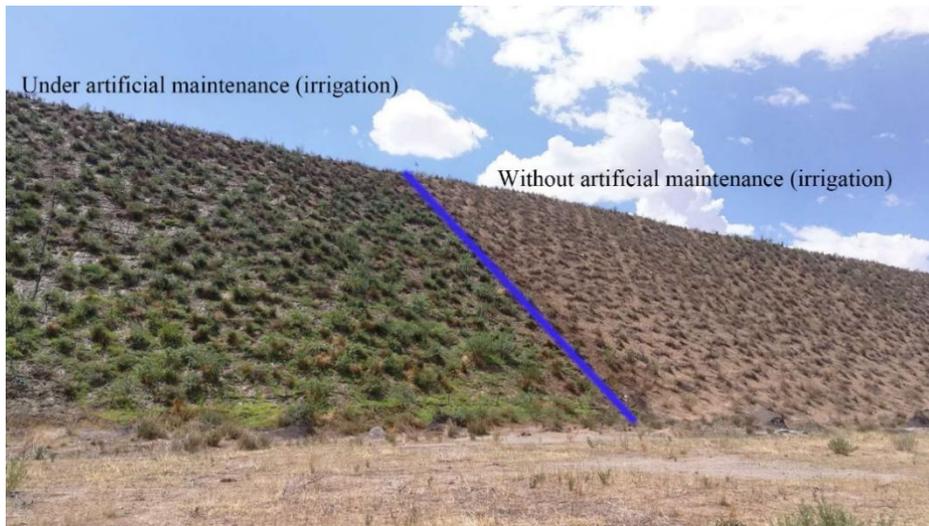
148 The specific study area is located in north dump of the Western One site of the Shengli coal
 149 mine, with an area of 1.07 km². The dump is divided into four platforms with an altitude range of
 150 980 m–1040 m and a slope angle of 33°. North dump has been reclaimed since 2006, with a total
 151 land area of 1.01 km², in which the slope greening area is 0.355 km², the platform greening area is
 152 0.655 km², and the green rate is 100%. Since the cessation of artificial conservation in 2013, the
 153 vegetation at the dump site has been degraded, and concentrated precipitation in summer has led to
 154 soil erosion, which can easily lead to slope erosion, collapse, debris flow and other disasters under
 155 the action of gravity.

156 3. Methodology and Materials

157 3.1 Framework

158 As a multi-stage dump, north dump is mainly composed of deep rock-soil, pulverized coal and
 159 surface soil close to the coal seam, and siltstone, mudstone and gravel of different sizes away from

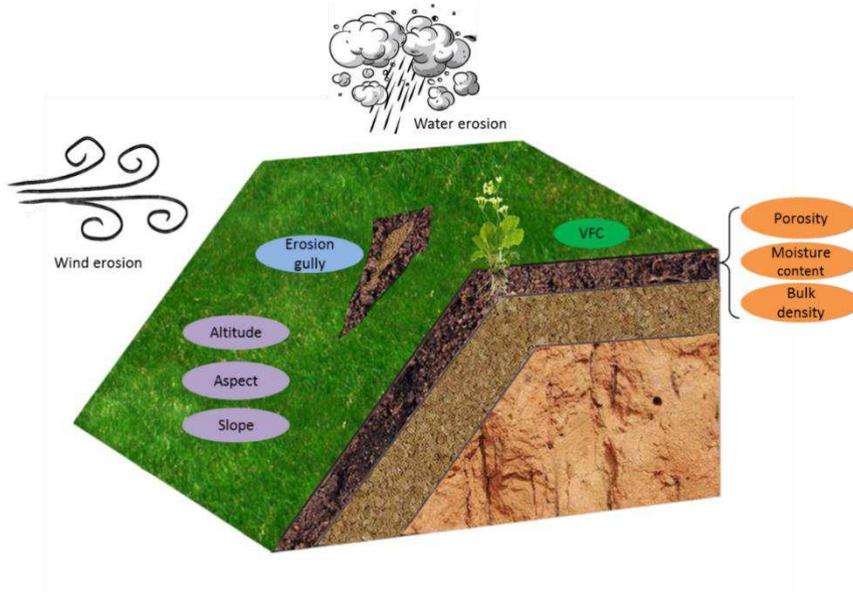
160 the slope. The soil is closely packed, with few internal gaps, the permeability coefficient is very
161 small, the seepage storage capacity is very low, and it contains essentially no humus. Without man-
162 agement and protection, dust and soil erosion can easily occur. Therefore, topsoil with a thickness
163 of 30–50 cm has been added to cover the surface. This part of the soil has a high organic matter
164 content and is fully matured. The soil has favorable air permeability, in which the number of seed
165 banks and microorganisms is large, which is conducive to plant growth. After completion of the soil
166 covering and before the arrival of the rainy season, forage grass seeds and shrub seeds with salt-
167 alkali resistance, drought resistance, strong adaptability and nitrogen fixation were planted. Artifi-
168 cial maintenance and management was then carried out for 4 years, including watering, sprinkling
169 and irrigation, adding topdressing fertilizer, disease and insect pest control, soil cultivation and
170 planting. Artificial maintenance and management are quite important for vegetation construction
171 (shown in Fig. 3). However, after this was stopped, north dump deformed to different degrees, and
172 its vegetation degenerated, under the influence of wind and rain erosion (Fig. 4). The intense pre-
173 cipitation which occurs during summer has led to soil and water loss, which can in turn easily lead
174 to erosion ditches, collapse, debris flow and other problems on the slope.



175

176 Fig. 3 A comparison of dumps with and without artificial maintenance in the south dumps of
177 Shengli coal field.

178



179

180 Fig. 4 Potential impact of wind and soil erosion on coal mine dumps under natural conditions.

181 Wind speed and direction in the study area affects wind erosion on the dump, while intense
 182 summer precipitation can cause water erosion. In addition to the vegetation cover and topsoil phys-
 183 ical properties themselves, the combined action of these factors will jointly lead to degradation of
 184 the dump and soil erosion. Obtaining these parameters quantitatively is particularly important. Jan
 185 et al. (2006) conducted interviews to understand the historical development of an erosion gully and
 186 calculated the average gully erosion rate through field measurement, Boggs et al. (2000) used GIS
 187 and modeling techniques to assess landform evolution, Hancock et al. (2008) quantified rills on an
 188 angle of repose slope of undisturbed mine spoil using a terrestrial laser scanner (TLS), and Tarolli
 189 and Sofia (2016) analyzed how surface mining affected the environmental landscape using remote
 190 sensing technology. However, the interview and modeling methods do not provide complete datasets,
 191 while TLS is not suitable for the acquisition of data over a large area, and the visible band of remote
 192 sensing is limited by spatial resolution. The research idea of this paper (Fig. 5) is to use a combina-
 193 tion of methods to assess the soil erosion and land degradation. The vegetation fractional coverage
 194 value of the dump is calculated using the RED and NIR bands of the GF-1 satellite. Obtaining high-
 195 precision topographic feature information and the abrasion degree of soil erosion furrows of the
 196 dump, in its unmanaged natural state, is done using aerial photogrammetry from a UAV equipped
 197 with a visible light camera. The physical properties of the topsoil are obtained by interpolation after
 198 field sampling. Finally, by analyzing the topsoil physical properties, VFC, elevation, step slope gra-
 199 dient, slope aspect, and morphological characteristics of erosion gullies in north dump using the
 200 Geographical Detector, the main factors leading to erosion can be identified, providing an effective
 201 reference for ecological recovery of the dump in the future.

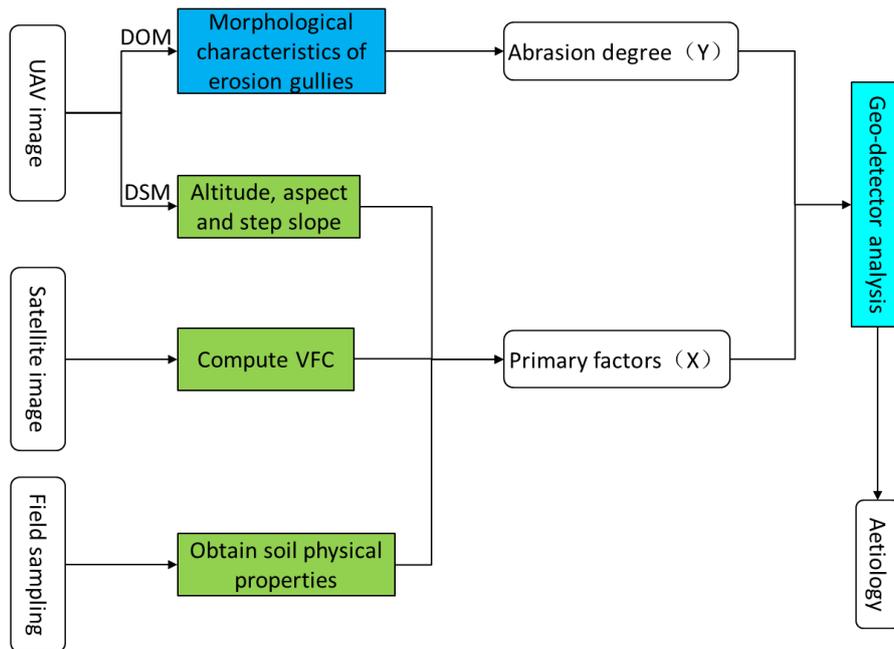


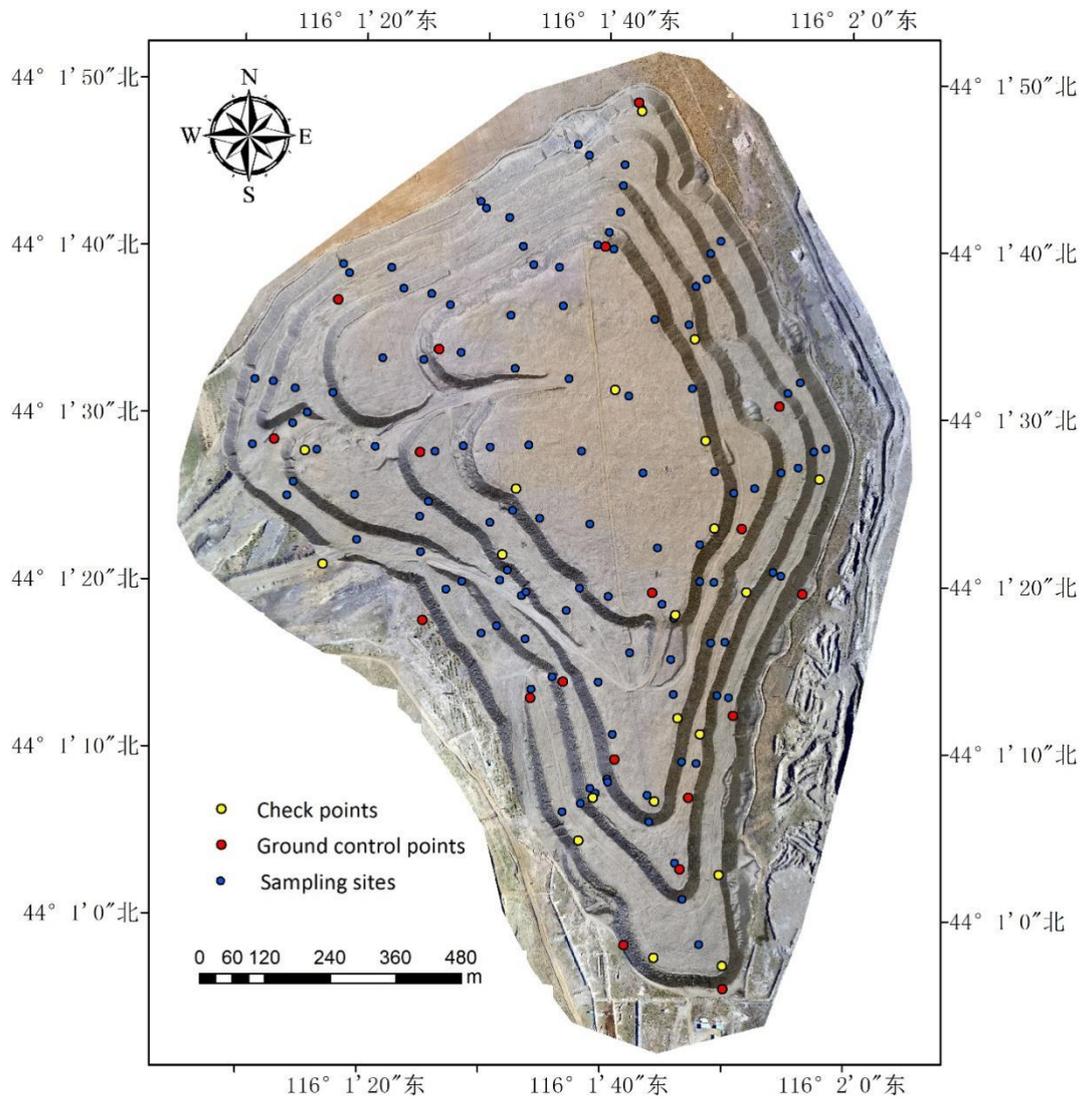
Fig. 5 Methodology flow chart.

3.2 Soil property

Soil is the foundation of vegetation restoration in dump lands, and determines the quality of reclamation (Liu et al., 2016a). The physical properties of the soil, such as its moisture content, bulk density, specific gravity, porosity, etc., can influence respiration and the water absorption of plant roots. Soils with deep root systems are characterized by high porosity, high soil infiltration and low soil bulk density (Wang et al, 2018). Nevertheless, the soils of overburden dumps are physically poor, and usually consist of a mixture of coarse-grained particles and rock fragments. In addition, the intensive traffic of heavy machinery used during reclamation can seriously compact soils, further degrading its physical quality. This anthropogenic activity has a significant influence on the soil stability to surface weathering, groundwater or infiltration, gully erosion and slope morphology (Huang et al., 2015).

The physical properties of the topsoil of north dump were obtained by field sampling. Samples from 117 points were collected in May 2017, uniformly spaced along the dump with a step length of 100 m. The sampling points were supplemented at soil erosion locations (Fig. 6) and the sampling depth was 15 cm. Samples at each sampling point were taken using a cutting ring, placed in a sealed bag, fresh weight of soil was measured on site, and sent to the laboratory 4 days later to measure their physical properties, including dry weight, porosity, water content and bulk density. The soil moisture content was determined using a drying method. Fresh weight of soil was measured on site, and dry weight was measured after drying in the laboratory. The soil bulk density and total porosity were determined using a cutting ring method. We pressed the ring knife vertically into the topsoil, and dig the ring knife out of the soil with a shovel and flatten the upper and lower ends.

226 Then the soil inside the ring cutter was transferred into the aluminum box without damage and
227 brought back to the laboratory for weighing.



228

229 Fig. 6 Sketch map of ground control points, control points and sampling sites.

230

231 **3.3 Vegetation fractional coverage**

232 Vegetation fractional coverage (VFC) is calculated from GF-1 remote sensing satellite images,
233 taken in August 2017. The GF-1 satellite is the first satellite of China's high-resolution Earth obser-
234 vation system, launched from the CZ-2D carrier rocket at 12:13:04 on April 26, 2013. It is charac-
235 terized by the addition of a multispectral camera with high spatial and temporal resolution (Table
236 1). It is widely used in the fields of geographical mapping, oceanic and atmospheric meteorological
237 observation, water conservation and forestry resources monitoring, fine management of urban areas

238 and transportation, epidemic situation assessment and public health emergency response, and sci-
 239 entific research on the Earth system. Based on the growth cycle of plants and the visibility of satellite
 240 images, the multi-spectral image with a spatial resolution of 8 m obtained by GF-1 on August 11,
 241 2016 was adopted. At this time, the plant growth condition was relatively good and there was rela-
 242 tively little cloud cover. To improve the quality and reduce the effects of terrain and atmospheric
 243 noise, the images were preprocessed using geometric correction, radiometric calibration, and atmos-
 244 pheric correction from ENVI 5.2.

245

Table 1 GF-1 specifications

Specifications	Details
<i>GF-1</i>	
Orbit type	Sun synchronous and repeating ground trace
Orbit altitude	645 km
Orbit inclination	98.0506°
Descending node local time	10:30 AM
Regression cycle	41 d
Spatial resolution	2 m (Panchromatic camera) 8 m & 16 m (Multispectral camera)
Revisit time	4 d

246

247 Vegetation fractional coverage (VFC) is based on the pixel dichotomy model established by Li
 248 et al. (2004). The estimation accuracy can reach 85%, but due to the limitation of spectral infor-
 249 mation, when the vegetation coverage is higher than 85%, the estimated value is obviously lower
 250 than the actual value, and the estimated value is no longer increased when it is about 90%. Because
 251 the study area is located in arid grassland and vegetation is sparse, where few areas have vegetation
 252 coverage of more than 90%. It is very suitable to take this model estimate the vegetation cover of
 253 the dump. Firstly, the normalized difference vegetation index (NDVI, denoted *NDVI*) is calculated
 254 from the RED (*R*) and NIR (*NIR*) bands of remote sensing images:

$$255 \quad NDVI = \frac{NIR - R}{NIR + R} \quad (1)$$

256 VFC (*VFC*) is then defined as a uniform NDVI value:

257
$$VFC = \frac{NDVI - NDVI_S}{NDVI_V - NDVI_S}, \quad (2)$$

258 with the following definitions:

259 $NDVI_S$: NDVI values of bare soil or areas covered with no vegetation.

260 $NDVI_V$: NDVI values of pixels completely covered by vegetation.

261 These quantities can be calculated from:

262
$$NDVI_S = \frac{VFC_{max} \cdot NDVI_{min} - VFC_{min} \cdot NDVI_{max}}{VFC_{max} - VFC_{min}}, \quad (3)$$

263
$$NDVI_V = \frac{(1 - VFC_{min}) \cdot NDVI_{max} - (1 - VFC_{max}) \cdot NDVI_{min}}{VFC_{max} - VFC_{min}}. \quad (4)$$

264 We take $VFC_{max}=1$ and $VFC_{min}=0$, so equation (2) can be expressed as:

265
$$VFC = \frac{NDVI - NDVI_{min}}{NDVI_{max} - NDVI_{min}}, \quad (5)$$

266 where NDVI values with accumulative probability of 5% and 95% were taken as $NDVI_{min}$ and
267 $NDVI_{max}$ respectively.

268 VFC values range from 0 to 1, and the following results are found in this study, through field
269 investigation and comparison with remote sensing images. An area with VFC greater than 0.8 is a
270 high VFC area, and the plants are mainly *pinus* (Genus), *ulmus pumila* and *armeniaca sibirica* trees.
271 Between 0.4 and 0.8 represents a medium VFC area, and the vegetation mainly includes shrubs such
272 as *C. korshinskii* and *hippophae rhamnoides*. Lower than 0.4 corresponds to a low VFC area, which
273 is dominated by herbs such as *medicago sativa* and *astragalus adsurgens* and bare land.

274 **3.4 UAV photogrammetry**

275 Traditional methods for geomorphological measurement use total station instrument (Kršák et
276 al., 2016) or Global Navigation Satellite Systems (GNSS). Although the measurement accuracy is
277 high, the data density is low, making it difficult to use for fine measurements (Aguera-Vega et al.,
278 2017). UAV photogrammetry can make use of the advantages of aerial triangulation to make a com-
279 plete and comprehensive observation of the area of interest, and has high measurement efficiency
280 and low cost, so it is widely used in topographic mapping of open mining areas.

281 To ensure the elevation accuracy of the digital surface model (DSM), 39 terrain points (Fig. 6)
282 were measured using CORS (Continuously Operating Reference Stations) with Trimble R8 GNSS
283 receivers. Of these, 19 points were chosen as check points, and the remaining 20 ground control

284 points (GCP) served as photogrammetric targets.

285 The Matrice 100 was chosen as the UAV used in this study. Its fuselage is made of carbon fiber
286 material, making it flexible and light, with a maximum flight duration of 40 min. It can be expanded
287 flexibly and has the possibility to be further developed by the user. The camera used to acquire the
288 images was the Zenmuse X3. The drone took off from the highest platform at a relative flight altitude
289 of 100 m. The longitudinal overlay of the photos was 80%, and the lateral overlay was 60%. The
290 survey took 57 min and 815 pictures were captured. The processing of the images was performed
291 using the Pix4D mapper software, which allows the reconstruction of a 3D model from the images
292 by applying SfM (Structure from Motion) algorithms. Pix4D mapper is highly automated, with easy
293 one-key manipulation. Users only need to input coordinates of GCPs and identify the location of
294 GCPs in the images, and it requires neither professional knowledge to process data and view the
295 results nor further human processing of the UAV data to automatically generate the digital evalua-
296 tion model (DEM) and the digital orthophoto map (DOM). Data results can be converted into other
297 formats for other software. The first step is to input them to the Pix4D desktop to carry out image
298 alignment after adjusting for chromatic aberration, noise, and the white balance of the pictures. We
299 obtained the camera position corresponding to each picture, the internal camera orientation param-
300 eters, and the sparse point cloud of the terrain, by using feature identification and feature matching.
301 The second step is to import and identify GCPs. The terrain profile, composed of a sparse point
302 cloud, can be seen where the aerotriangulation rays intersect, so that GCPs can be easily identified.
303 In the final step, the dense point cloud, 3D scene reconstruction, post-processing, the DSM and the
304 DOM are produced using a one-key procedure. The terrain of the whole dump can be obtained from
305 the DEM while particular locations on the dump can be extracted from the DOM.

306 **3.5 Geographical detector**

307 The geographical detector model was developed for exploring the relationships between spatial
308 patterns of landscapes and the factors which impact them, by Wang et al. (2010). Its principle is that
309 the spatial distributions of two attributes tend to be similar, if there are spatial relationships or inter-
310 actions between them. The spatial data do not have to be stratified in geospatial terms, although the
311 attributes can be stratified. The similarity between two attributes can be represented by the so-called
312 power of determinant ($P_{D,H}$), without requiring many external assumptions or constraints. It is de-
313 fined as:

$$314 \quad P_{D,H} = 1 - \frac{\sum_{i=1}^m (n_{D,i} \times \sigma_{D,i}^2)}{n\sigma^2}, \quad (6)$$

315 with the following variable definitions:

316 $P_{D,H}$: the power of determinant of the D factor to the abrasion degree (H).

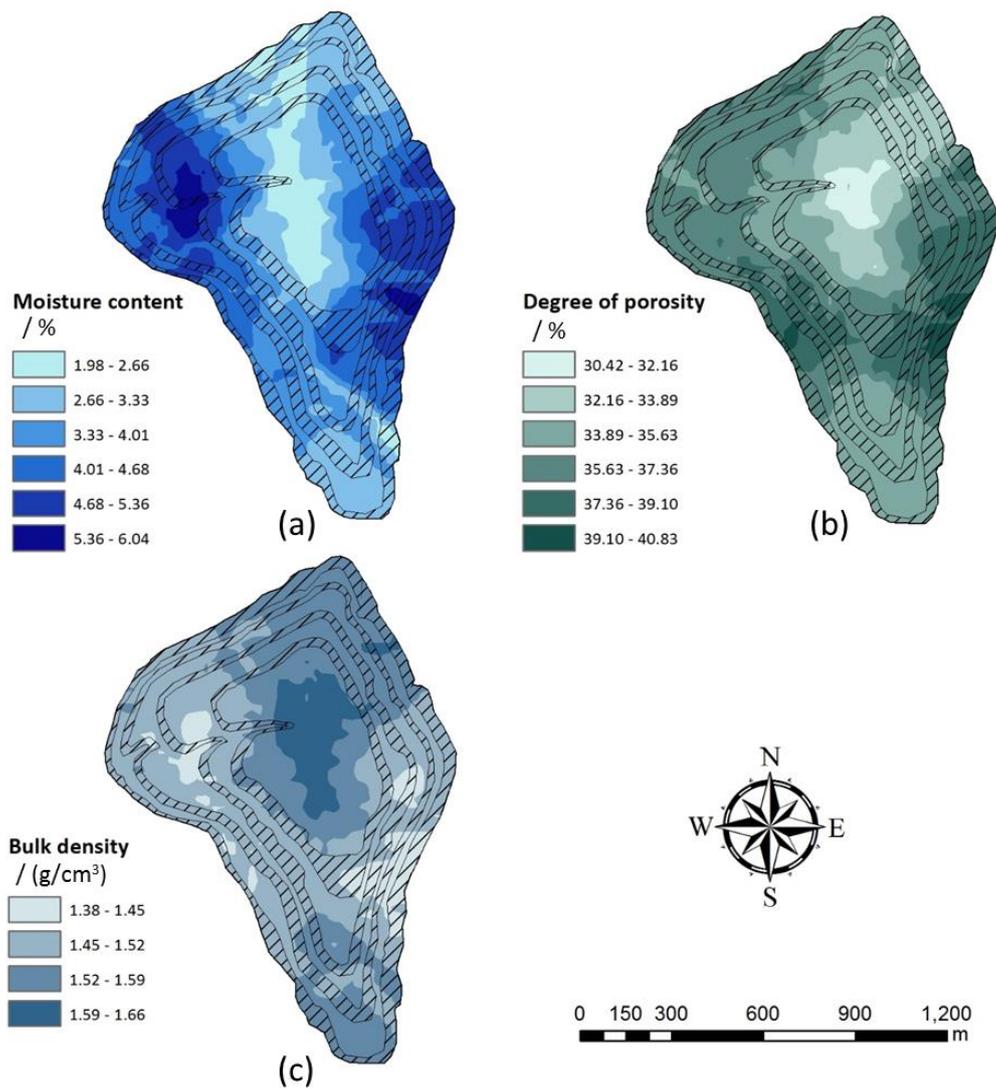
- 317 n : the total number of units in the study area.
318 $n_{D,i}$: the number of cells in the i th partition in the D factor.
319 σ : total standard deviation of the abrasion degree.
320 $\sigma_{D,i}$: standard deviation of the i th partition in the D factor.

321 The geographical detector model is effective at analyzing the effects of different driving factors
322 on spatial heterogeneity, and it has been applied to research on landscapes (Liang and Yang, 2016),
323 the environment (Zhang and Zhao, 2018) and animal habitats (Liao et al., 2016). Geogdetector was
324 employed in our study to objectively assign weights to different environment factors affecting soil
325 erosion hazards at the dump. As the dependent variable, soil erosion was described by the erosion
326 area. The area of gully erosion was described by the maximum length multiplied by the maximum
327 width, and the area of sheet erosion was identified manually. The independent variables were the
328 step slope gradient, elevation, VFC, degree of porosity, bulk density, soil moisture content and slope
329 aspect for each hazard location. In this way, the main driving factors that affect the amount of soil
330 erosion were assessed. The geographical detector consists of four parts, i.e., the factor detector,
331 interaction detector, risk detector, and ecological detector. We focused on the factor detector and the
332 interaction detector in this study.

333 **4. Results**

334 ***4.1 Soil property analysis***

335 The main soil components of the dump are siltstone, mudstone, carbonaceous mudstone and
336 chestnut soil. However, the proportion of chestnut soil is very small, with an average overburden
337 thickness of 30 cm, a sandy soil texture and a humus content between 1.5% and 3%. Table 2 shows
338 the physical properties of the soil samples. The average bulk density of the topsoil is as high as 1.51
339 g/cm³, and the average porosity of the soil is 35.89%. The soil tightness is much higher than that of
340 the soil under natural conditions. The average value of soil moisture content was 3.83%, indicating
341 the topsoil has a poor ability to conserve moisture, therefore it has a low fertility and is not conducive
342 to plant growth, making it susceptible to erosion. The sampling points were interpolated using the
343 Kriging method to obtain the physical properties (Mendes et al., 2019) of the entire dump (Fig. 7),
344 including soil moisture content, porosity and bulk density. The topmost platform is the most heavily
345 compacted, so that it has the smallest porosity, maximum bulk density and minimum water content.



346

347 Fig. 7 Soil physical properties. (a) Moisture content. (b) Degree of porosity. (c) Bulk density.

348

349

Table 2 soil physical properties

Statistical result	Bulk density /g•cm ³	Porosity /%	Moisture content/%	
			Original value	Logarith-metics
Maximum	1.860	55.770	13.298	1.124
Minimum	1.040	21.952	0.667	-0.176

Average	1.511	35.885	3.832	0.498
Median	1.520	35.943	3.019	0.480
Standard deviation	0.150	5.545	2.565	0.270
Variance	0.023	30.746	6.578	0.073

350

351 4.1.1 Porosity

352 The soil porosity of the dump site in the study area is low around and high in the middle, which
353 may be due to the fact that the top platform has been repeatedly rolled by heavy machinery, resulting
354 in soil hardening and topsoil hardening (Liu et al., 2016b). Although dust is reduced to a certain
355 extent, the soil with low porosity reduces the infiltration rate of water, resulting in a large amount
356 of surface water runoff, thus causing soil erosion.

357 4.1.2 Moisture content

358 The east and west sides of the dump have the highest water content. The high moisture content
359 on the east side is due to the short illumination time and the low evaporation on sunny slopes. The
360 high moisture content in the west is because there is a lane leading to the top on the platform in the
361 west of the dump. After long-term rolling, the platform in the west is sunken, which plays a certain
362 role in water collection. The low moisture content on the south side is due to sunny slopes (Pan et
363 al., 2017). On the north side, the dry and cold monsoon prevailing in spring and winter takes away
364 a large amount of water from the soil surface, so the soil moisture content is low.

365 4.1.3 Bulk density

366 The distribution of soil bulk density on slope surface is affected by vegetation and topography.
367 The moisture, nutrient and air content in soil also change with the slope position. As water flows
368 through the surface, it moves fine particles from the soil to the bottom of the slope. With the rain
369 erosion, some slope will also occur collapse. The soil surface structure from the upper slope to the
370 lower slope tends to change from compact to loose and porous (Lü et al., 2018).

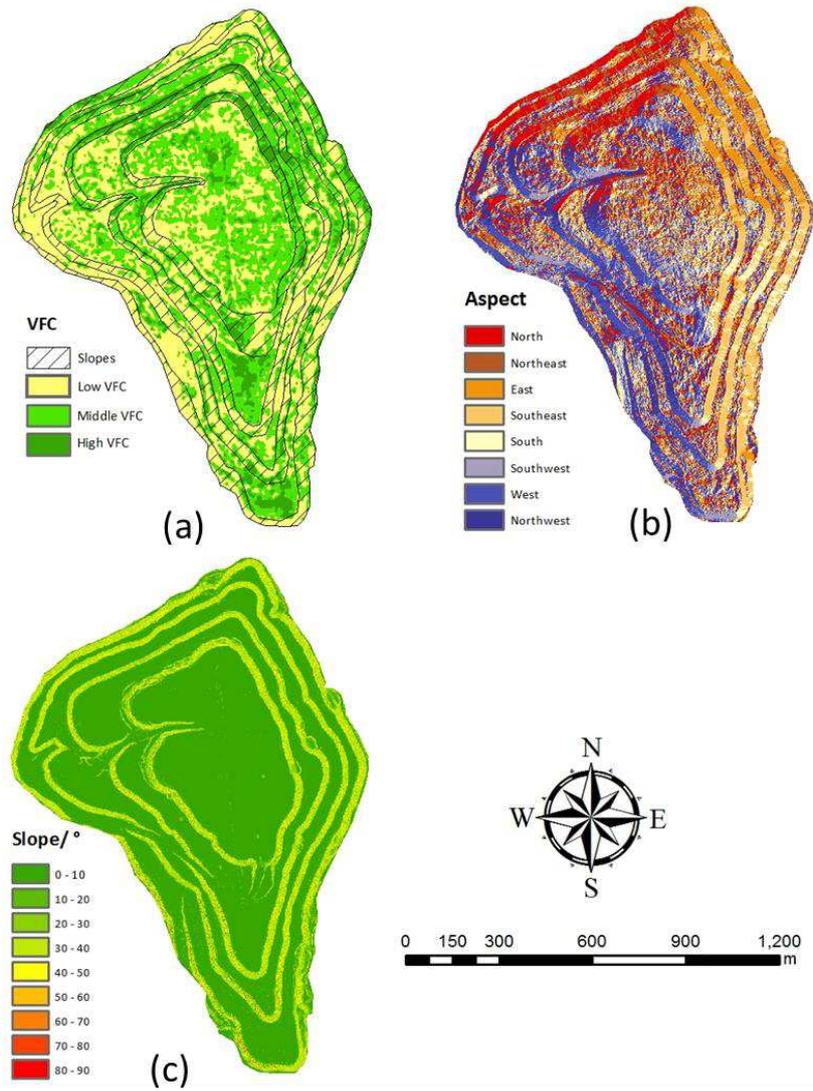
371 **4.2 VFC**

372 Vegetation coverage is affected by the physical properties of the topsoil(Huang et al., 2015),
373 plant species(Liu et al., 2016b) and local climate(Brown et al., 1997), and there is significant vari-
374 ation in its spatial distribution. The north dump covered 483,580 m² of low vegetation, 453,704 m²
375 of medium vegetation and 120188 m² of high vegetation on September 1, 2017. As can be seen from

376 Fig. 8a, the high VFC areas are mainly distributed on the slope surface, which has not been com-
377 pacted by heavy transport equipment and is more conducive to water infiltration and plant root
378 growth (Qi, 2017). In the high VFC area of the platform in the south of the dump, because of inten-
379 sive planting during reclamation, the tree cluster distribution makes it more stable and not so vul-
380 nerable to degradation. In different slope aspect, VFC of north slope and east slope was higher than
381 that of south slope and west slope. It indicates that the growth of vegetation is closely related to the
382 slope direction (Frankard et al., 2000). The north slope and the east slope are shady slopes with short
383 light time, weak plant transpiration and little water evaporation, while the south slope and the west
384 slope are sunny slopes with long light time, strong plant transpiration and much water evaporation,
385 which is difficult for vegetation restoration.

386 ***4.3 Geomorphology and gully erosion***

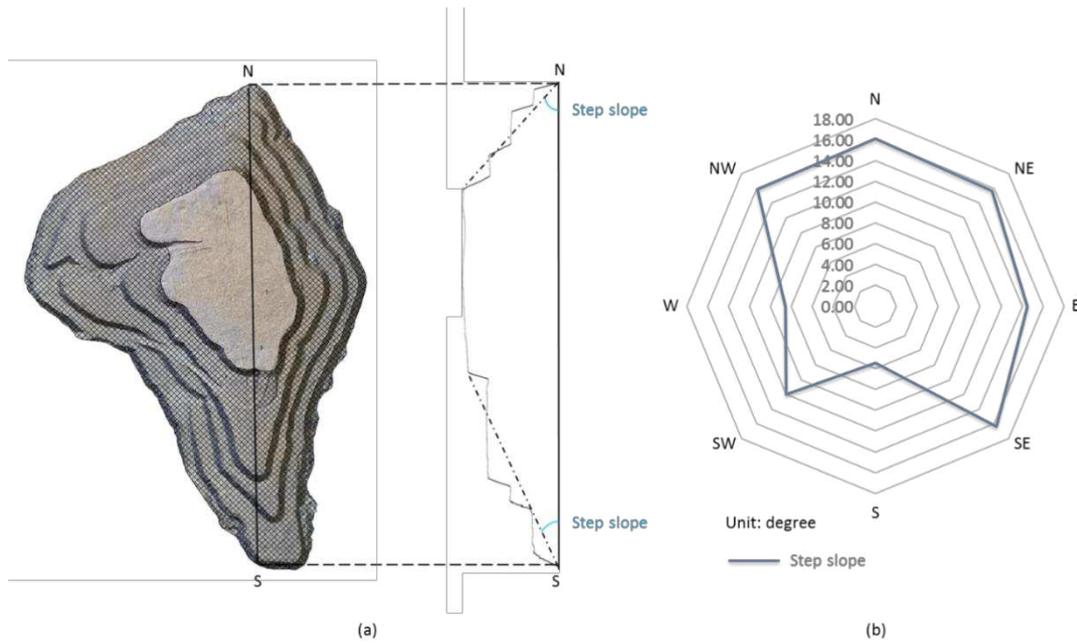
387 We utilized a SfM algorithm to process the aerial photos to acquire the geomorphology and
388 gully erosion of the dump. The slope aspect (Fig. 8b) and gradient (Fig. 8c) were then obtained from
389 the DSM. The slope gradients of each platform are approximately equal, at about 33 degrees, how-
390 ever, the step slope aspects are more variable (Fig. 9). In the north, northwest, northeast, east and
391 southeast directions, the step slope gradients are about 16 degrees, but in the west, southwest and
392 south directions, the step slope gradient is significantly shallower, even reaching as low as approx-
393 imately 6 degrees. The greater the step slope gradient is, the more vulnerable the surface is to cause
394 Rock fragment movement (Nyssen et al., 2006a) and wind-water erosion (Zhang et al., 2019).



395

396

Fig. 8 VFC, Slope aspect and gradient. (a)VFC. (b) Aspect. (c) Gradient.



397

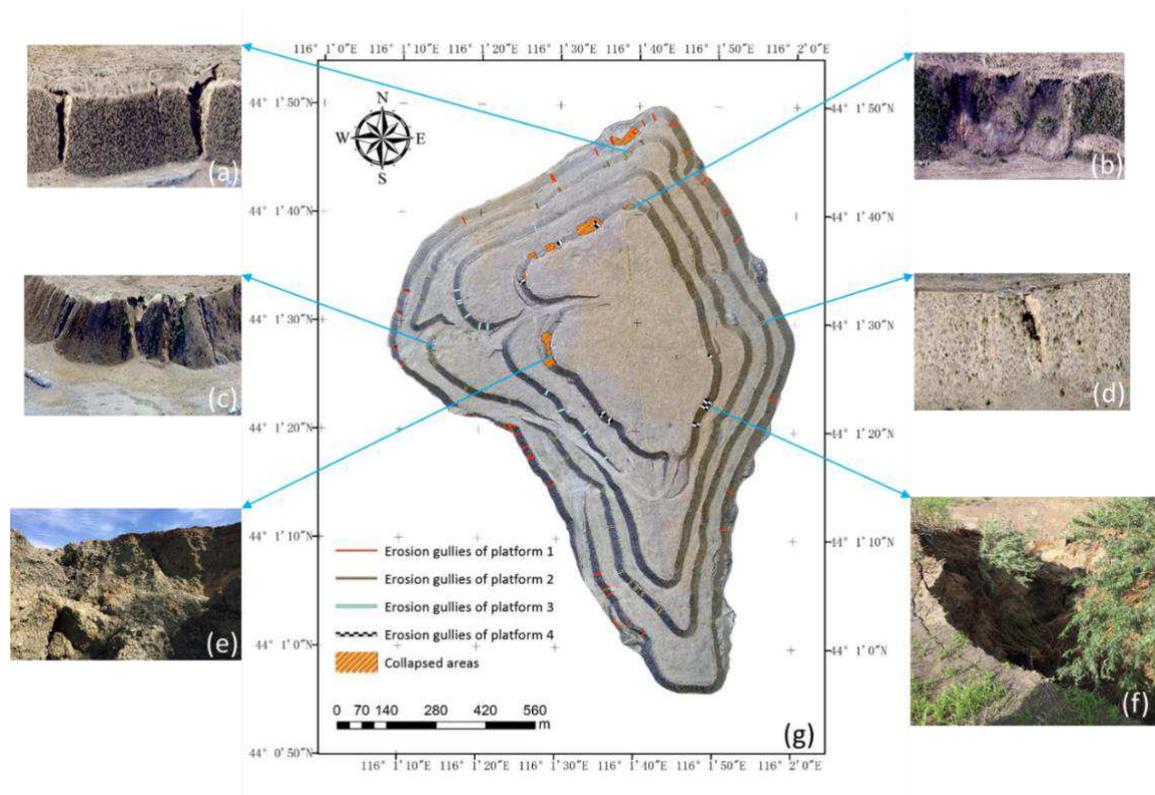
398 Fig. 9 Step slope. (a) The elevation profile from the DSM is shown on the right, with its location
 399 shown on the map on the left. (b) Radar chart of the step slope gradient.

400 The resolution of the DOM was 6 cm. Fig. 10 shows the distribution of erosion gullies and
 401 collapsed areas. All the geological hazards are located in slope areas. The heavy mechanical equip-
 402 ment used in the dump compacts the dump platform, making it less susceptible to infiltration of
 403 heavy rain. However, on the slopes of the dump, loose rocks and soils, poor stability, and low water
 404 content make it vulnerable to water-wind erosion (Kainthola et al., 2011). The erosion gullies radar
 405 chart (Fig. 11) shows the erosion extent on four platforms in different directions. It can be found in
 406 combination with Table 3, the region with the most serious soil erosion was located west to north-
 407 west of the dump, where five collapse areas were concentrated, totaling 7399.57 m², and accounting
 408 for 53.72% of the total erosion area. Erosion gullies were mainly distributed on the west side of the
 409 dump, accounting for 50% of the total erosion gullies. There were 13 erosion gullies to the west
 410 (247.5° to 292.5°), 16 erosion gullies to the northwest (292.5° to 337.5°), and 20 erosion gullies to
 411 the southwest (202.5° to 247.5°). The number of erosion gullies on the east side (E, NE, SE) was
 412 smaller, accounting for 28.26% of the total. The slope aspect with more severe erosion is in line
 413 with the prevailing local wind direction (Nyssen et al., 2010). The soil erosion of the fourth platform
 414 slope(crest) is the most serious, while the soil erosion of the first and second layers is lighter than
 415 that of the other two, as also shown by Cha et al. (2000). This indicates that wind erosion is a
 416 significant cause of these hazards.

417

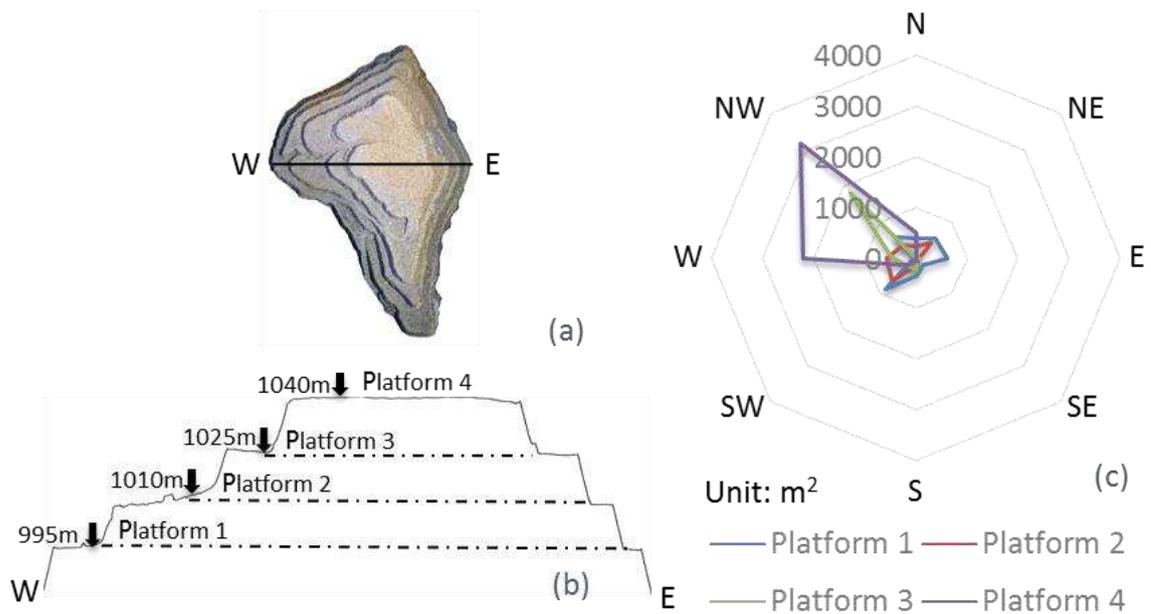
418

419



420

421 Fig. 10 Geologic hazards at north dump. (a–d) Hazards shown in 3D models. (e–f) Photos taken in
 422 the field. (g) Distribution of hazards.



423

424 Fig. 11 Erosion gullies radar chart. (a) Section line. (b) Transverse section. (c) Radar chart of ero-
 425 sion extent of each platform: the distance from the center represents the erosion extent.

426

427

428

429

Table 3 Distribution of soil erosion

Aspect	Erosion gullies	Cumulative gullies ratio	Erosion areas	Cumulative area ratio
N	10	10.87%	1108.72	8.79%
NE	13	25.00%	948.12	16.31%
E	8	33.70%	747.39	22.23%
SE	5	39.13%	272.57	24.39%
S	10	50.00%	827.58	30.95%
SW	20	71.74%	1933.28	46.28%
W	10	82.61%	827.58	52.84%
NW	16	100.00%	5948.07	100.00%

430

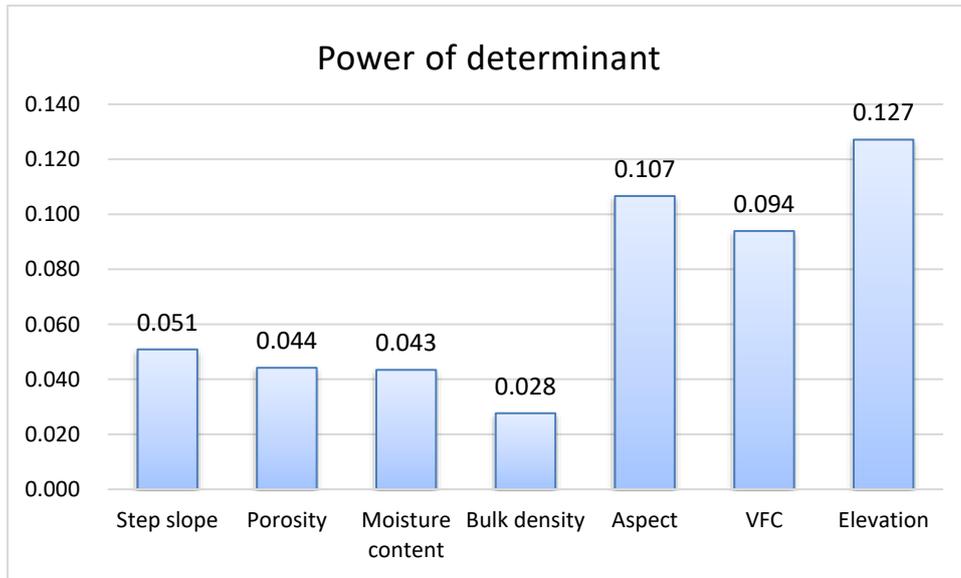
431

432 **4.4 Analysis of dump hazards using geographical detector**

433 4.4.1 Factor detector

434 The factor detector reveals the impact of a single driving factor on the degree of soil erosion.
 435 The power of determinant ($P_{D,H}$) of each factor is shown in Fig. 12. This shows that the most im-
 436 portant contributing factors are elevation (0.127), slope aspect (0.107), and VFC (0.094), followed
 437 by step slope gradient (0.051), degree of porosity (0.044), soil moisture content (0.043), and bulk
 438 density (0.028). The larger wind force at the top of the dump accelerates the development of erosion
 439 gullies into collapse areas. The collapse areas are mainly concentrated on the slope of the fourth
 440 platform, indicating that the higher the elevation of an area is, the more vulnerable it is to soil ero-
 441 sion(Zhang et al., 2012). The slope aspect is related to wind direction. The prevailing westerly wind
 442 in spring and winter means that erosion gullies are mainly distributed on the west side of the dump.
 443 For regions without soil erosion, vegetation can play a role as a wind break and in sand fixation,
 444 thus preventing soil erosion. For regions with soil erosion, vegetation can effectively inhibit the
 445 development of erosion gullies and prevent the combination of adjacent collapse areas. Gentle to-
 446 pography helps reduce soil erosion, especially on windward slopes, where a smaller gradient slows

447 the acceleration of upwind airflow and gravity subsidence of the topsoil. A surface soil with large
 448 porosity is conducive to plant growth and the infiltration of water, thereby reducing water erosion
 449 and inhibiting the further development of fissures. The climate in the experimental area is dry. The
 450 higher initial soil moisture content is conducive to a stable soil aggregate structure, enhanced soil
 451 permeability, and reduced soil disintegration rate, making the soil less susceptible to erosion (An-
 452 drea et al., 1997). The power of determinant $P_{D,H}$ of bulk density is small, indicating that bulk den-
 453 sity does not have a large direct impact on soil erosion in the dump.



454

455 Fig. 12 The influences of different driving factors on the degree of erosion.

456 4.4.2 Interaction detector

457 Soil erosion on the slope of the dump is the result of the joint action of many environmental
 458 factors. In a real environment, the distribution and change of soil erosion cannot be affected by only
 459 a single factor or a single natural cause (Xiao et al., 2018). In order to establish whether any two
 460 environment factors influence the degree of soil erosion independently, or interact with each other,
 461 the interaction detector is applied. The interactive influence between each pair of driving factors is
 462 categorized as follows:

463 Weaken and nonlinear: $P_{D,H}(x \cap y) < \text{Min}(P_{D,H}(x), P_{D,H}(y))$

464 Weaken and univariate: $\text{Min}(P_{D,H}(x), P_{D,H}(y)) < P_{D,H}(x \cap y) < \text{Max}(P_{D,H}(x), P_{D,H}(y))$

465 Enhance and bivariate: $P_{D,H}(x \cap y) > \text{Max}(P_{D,H}(x), P_{D,H}(y))$

466 Independent: $P_{D,H}(x \cap y) = P_{D,H}(x) + P_{D,H}(y)$

467 Enhance and nonlinear: $P_{D,H}(x \cap y) > P_{D,H}(x) + P_{D,H}(y)$.

468 This study found that the interaction detectors for abrasion degree can be divided into two types:
 469 enhance and bivariate, and enhance and nonlinear. Though the power of determinant of each pair of

470 driving factors was bigger than that of each individual driving factor, there were significant differ-
 471 ences in the interaction strength of different factors (Table 4). The interactive effects of the step
 472 slope gradient and elevation, soil moisture content and bulk density, and slope aspect and elevation
 473 were greater than the maximum of their separate effects. However, the interactive effects of other
 474 pairs were greater than the sum of the effects of the corresponding factors individually, indicating a
 475 strong synergistic effect between the two factors. The interactive influence power of VFC with other
 476 factors is higher than 0.36, indicating that VFC can significantly enhance the impact of various
 477 environmental factors on abrasion degree. VFC can be used as an auxiliary indicator factor for the
 478 monitoring of the aetiology of erosion and degradation (Tong et al., 2014). In addition, driving factor
 479 pairs with strong synergistic effects also include: porosity and moisture content, porosity and bulk
 480 density, slope aspect and moisture content, and slope aspect and bulk density. The physical proper-
 481 ties of the soil are intrinsically related, and the synergistic effect of slope aspect, moisture content
 482 and bulk density is mainly reflected in the change of these physical properties caused by wind ero-
 483 sion.

484 Table 4 Results of the interaction detector

	Step slope gradient	Porosity	Moisture	Bulk density	Slope aspect	VFC	Elevation
Step slope	0.051						
Porosity	0.202	0.044					
Moisture	0.228	0.331	0.043				
Bulk density	0.170	0.481	0.086	0.028			
Slope aspect	0.116	0.243	0.395	0.328	0.107		
VFC	0.443	0.650	0.367	0.399	0.487	0.094	
Elevation	0.194	0.214	0.247	0.206	0.248	0.427	0.127

485 * Footnote: values in bold are stronger than other interactions.

486

487 5. Discussions

488 5.1 Implications

489 Dumps are inevitable troubles that were formed by open pit mining. Through the analysis in
 490 section 4.4, it is concluded that elevation, VFC, and aspect are the main driving factor affecting soil
 491 erosion. Soil property and aspect affect the distribution of VFC. VFC, elevation and aspect have
 492 combined effects on soil erosion. Slope stability management (Sotiris et al., 2018) and soil erosion
 493 (Marchi et al., 2018) are worldwide problems. The temporal and spatial evolution of soil erosion
 494 was analyzed by remote sensing (Shivesh Kishore et al., 2016) and model (Wischmeier and Smith,

495 1978; Rodriguez Caballero, 2015). Soil erosion is usually controlled by changing land use pat-
496 terns(Nearing et al., 2005) and geomorphic reclamation(Martin et al., 2015). As it was mainly
497 caused by wind erosion and water erosion, the following three suggestions are put forward accord-
498 ing to the characteristics of this area:

499 (1) Given the serious soil erosion on the windward slope in the west, we suggest reducing the
500 ladder slope gradient to the northwest. The slope would then better reflect the natural landform
501 before its disturbance by human influence, leading to a reduction in wind erosion.

502 (2) VFC is not only the main factor affecting soil erosion but also an auxiliary indicator for
503 monitoring soil erosion. Increasing VFC is important for the prevention and treatment of land deg-
504 radation at dump sites. We can introduce dominant species and appropriately increase the density
505 and diversity of vegetation communities. At the same time, sand barriers and biological arpeggios
506 could be adopted to enhance the stability of vegetation community.

507 (3) Strong wind erosion leads to coarser soil surface particles and a great loss of nutrients.
508 Compaction by large machinery reduces soil porosity and increases bulk density, which is not con-
509 ducive to plant growth. However, plant roots can reach deep into the soil, promote the agglomeration
510 of micro-aggregates with small particle size, reduce soil bulk density and increase soil porosity,
511 forming a positive feedback loop. Therefore, shrubs and herbaceous pioneer plants with strong
512 adaptability and good survival rates should be widely planted in the early stage of reclamation, so
513 as to provide wind protection, fix the soil and improve the soil structure.

514 The site-specific suggestions put forward in this paper are applicable to the whole grassland
515 area of Eastern Inner Mongolia. The grassland area of Eastern Inner Mongolia is more than 380,000
516 km². More than 40 mining areas are scattered in the grasslands of eastern Inner Mongolia. The total
517 coal resources are 249 billion tons, and the annual mining volume is 150 - 300 million tons. The
518 total area of the dump is more than 14,460 km².

519 ***5.2 Limitation of the method and future work***

520 In this paper, a method of soil erosion monitoring based on multi-source data and combined
521 with geographical detector can effectively identify the main factors causing soil erosion, but there
522 are still some scientific problems to be solved. On the one hand, the UAV platform can be equipped
523 with multi-spectrum or hyper-spectrum in the future to obtain more spectral information for con-
524 structing vegetation parameters. Soil information can be obtained by inverting vegetation parame-
525 ters, and soil erosion warning can be carried out. On the other hand, using UAV loaded RTK (Real
526 - time Kinematic) can greatly reduce the time of data acquisition, so as to construct a more rapid
527 and efficient monitoring method.

528 In addition, this thesis only proposes to alleviate wind erosion in the dump by reducing wind-
529 ward slope, but the appropriate gradient is not discussed in depth. In fact, as far back as 1969 geo-
530 morphic design approach has been used in landscape planning (McHarg, 1969). At present, the ge-

531 omorphic remodeling of open-pit mines in the United States and Australia is mainly based on geo-
532 morphic design approach. Common software includes GeoWEPP (Hao et al., 2015), GeoFluv (Han-
533 cock et al., 2019), and so on. Because of the high cost of land acquisition in China, mining compa-
534 nies will reduce the land acquisition area as much as possible. How to construct the terrain resistant
535 to wind and water erosion is an important direction of mine restoration in the future.

536 The combined methodology that integrating satellite remote sensing imagery for vegetation
537 monitoring, UAVS platform for accurate terrain and soil erosion investigation, and field sampling
538 for soil property analysis, could fill the gap of different method, that satisfied the demand of coal
539 mine dumps land reclamation monitoring, investigation, assessment.

540 **6. Conclusions**

541 This paper investigates an outer dump at which artificial management has been stopped, as an
542 example to explore the soil erosion after 5 years of evolution of natural processes, and to determine
543 the main factors influencing the degree of soil erosion. The main conclusions are as follows:

544 1) At a height of 100 m, UAV pictures combined with a SfM algorithm can be used to accu-
545 rately and quickly obtain a 3D model of the dump with a total height of 60 m. The accuracy is
546 sufficient to position and identify the soil erosion area.

547 2) The step slope gradient of the coal mine dumps is significantly steeper than the natural land-
548 form, which makes the windward slope vulnerable to wind erosion. It is suggested that the windward
549 slope be designed to imitate the natural landform, for example by reducing its gradient.

550 3) Vegetation coverage is the main factor affecting soil erosion. Measures such as sand barriers
551 and biological basketry should be applied at the slope to reduce and restrain gully erosion and planar
552 erosion. Pioneer plants that can improve soil structure should be widely planted on the platform at
553 the early stage of reclamation, so as to facilitate water infiltration and enhance the diversity and
554 stability of vegetation communities during the natural recovery period.

555 4) The combined methodology that integrating satellite remote sensing imagery for vegetation
556 monitoring, UAVS platform for accurate terrain and soil erosion investigation, and field sampling
557 for soil property analysis, could fill the gap of different method, that satisfied the demand of coal
558 mine dumps land reclamation monitoring, investigation, assessment.

559 **7 Conflict of interest**

560 The authors have declared no conflict of interest.

561 **8 Acknowledgements**

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Figures

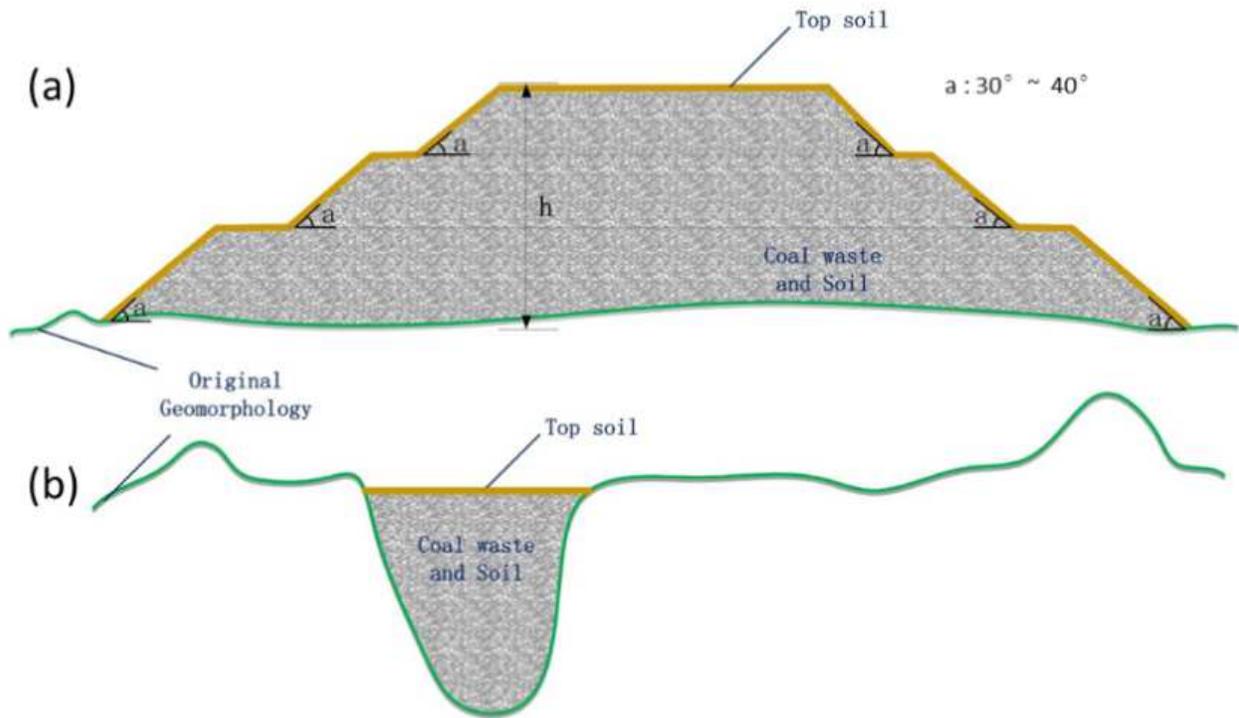


Figure 1

Waste dump diagram. (a) Multi-stage dump. (b) Gully filled dump.

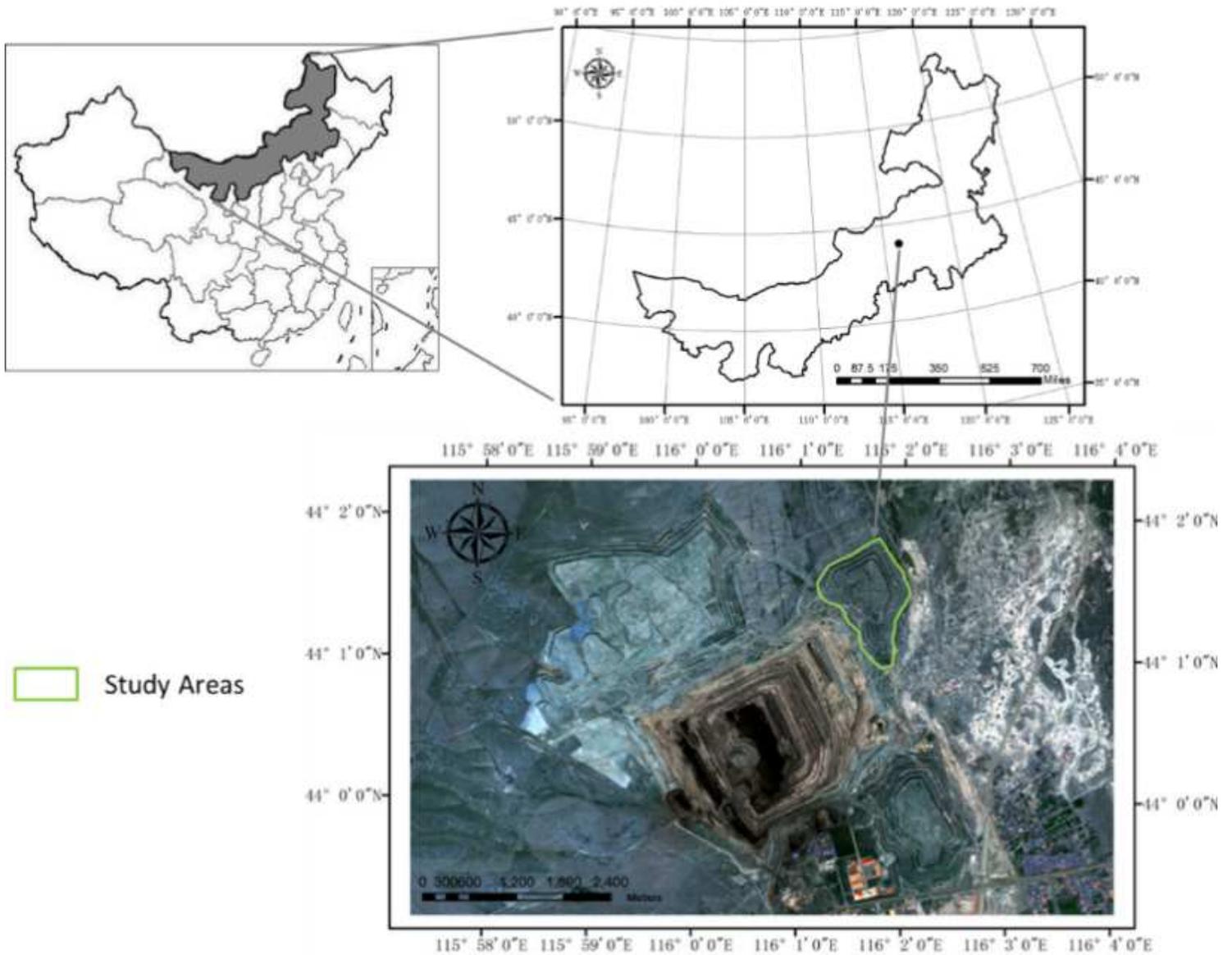


Figure 2

The study area location. Note: The designations employed and the presentation of the material on this map do not imply the expression of any opinion whatsoever on the part of Research Square concerning the legal status of any country, territory, city or area or of its authorities, or concerning the delimitation of its frontiers or boundaries. This map has been provided by the authors.

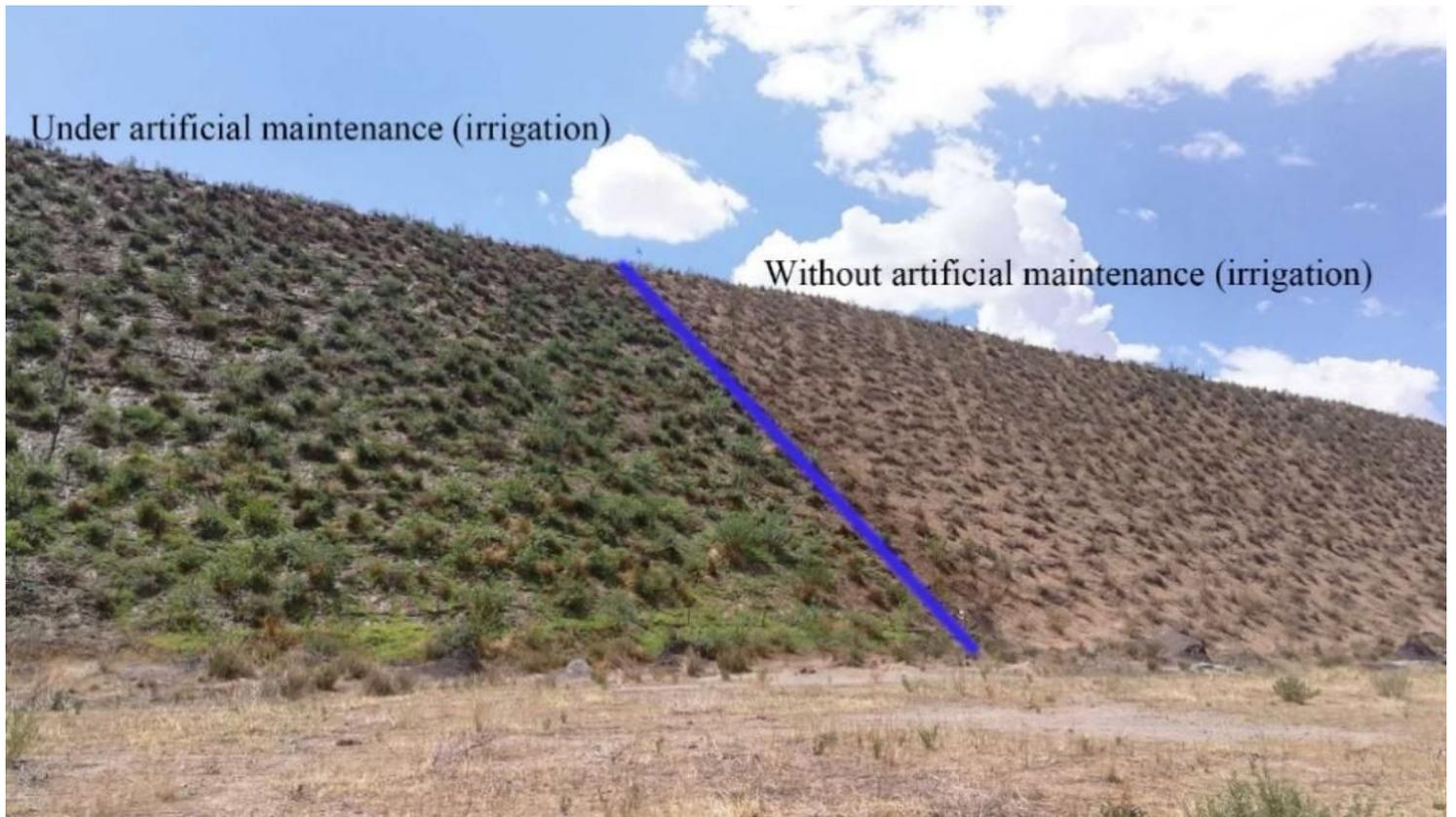


Figure 3

A comparison of dumps with and without artificial maintenance in the south dumps of Shengli coal field.

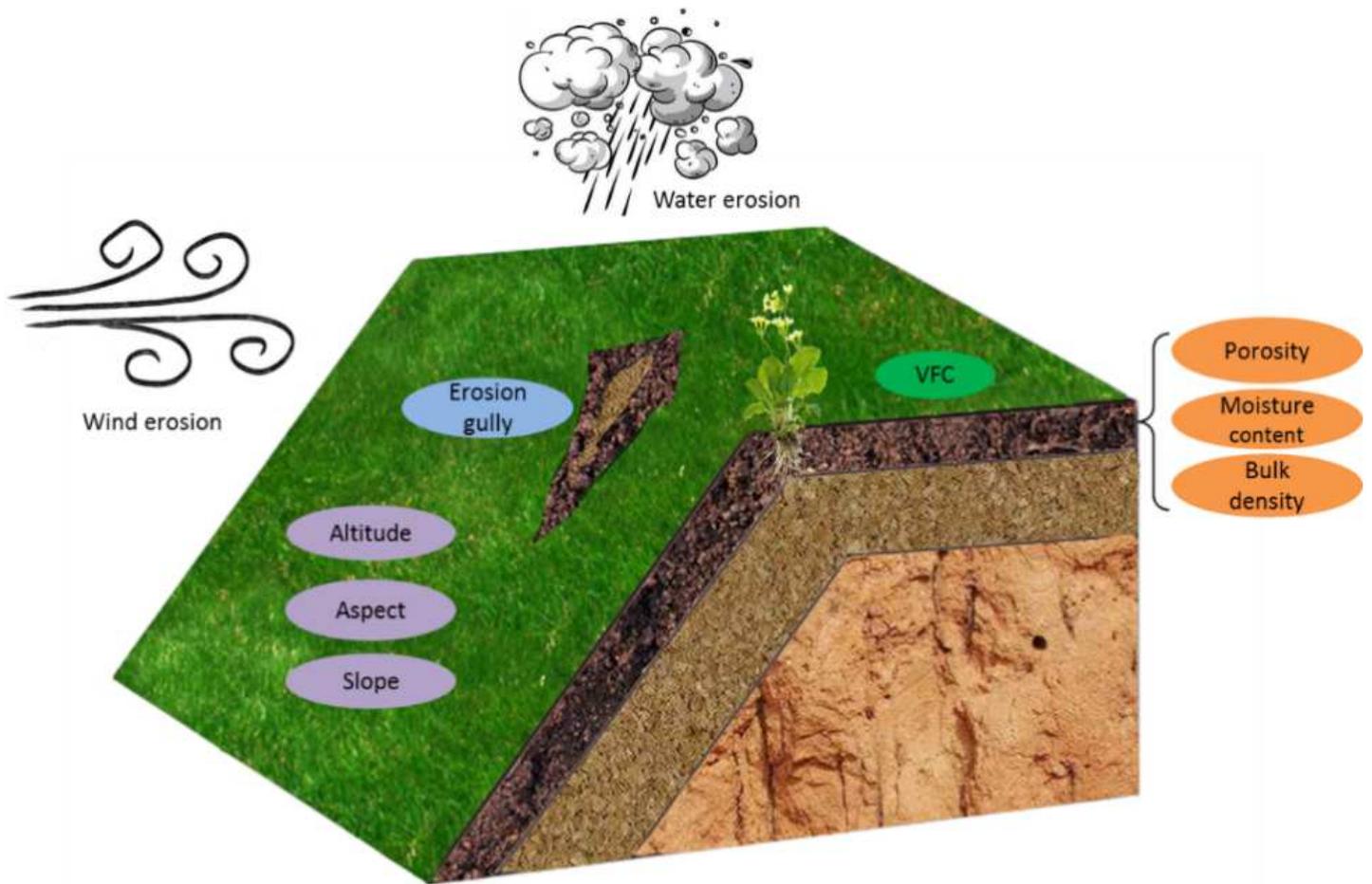


Figure 4

Potential impact of wind and soil erosion on coal mine dumps under natural conditions.

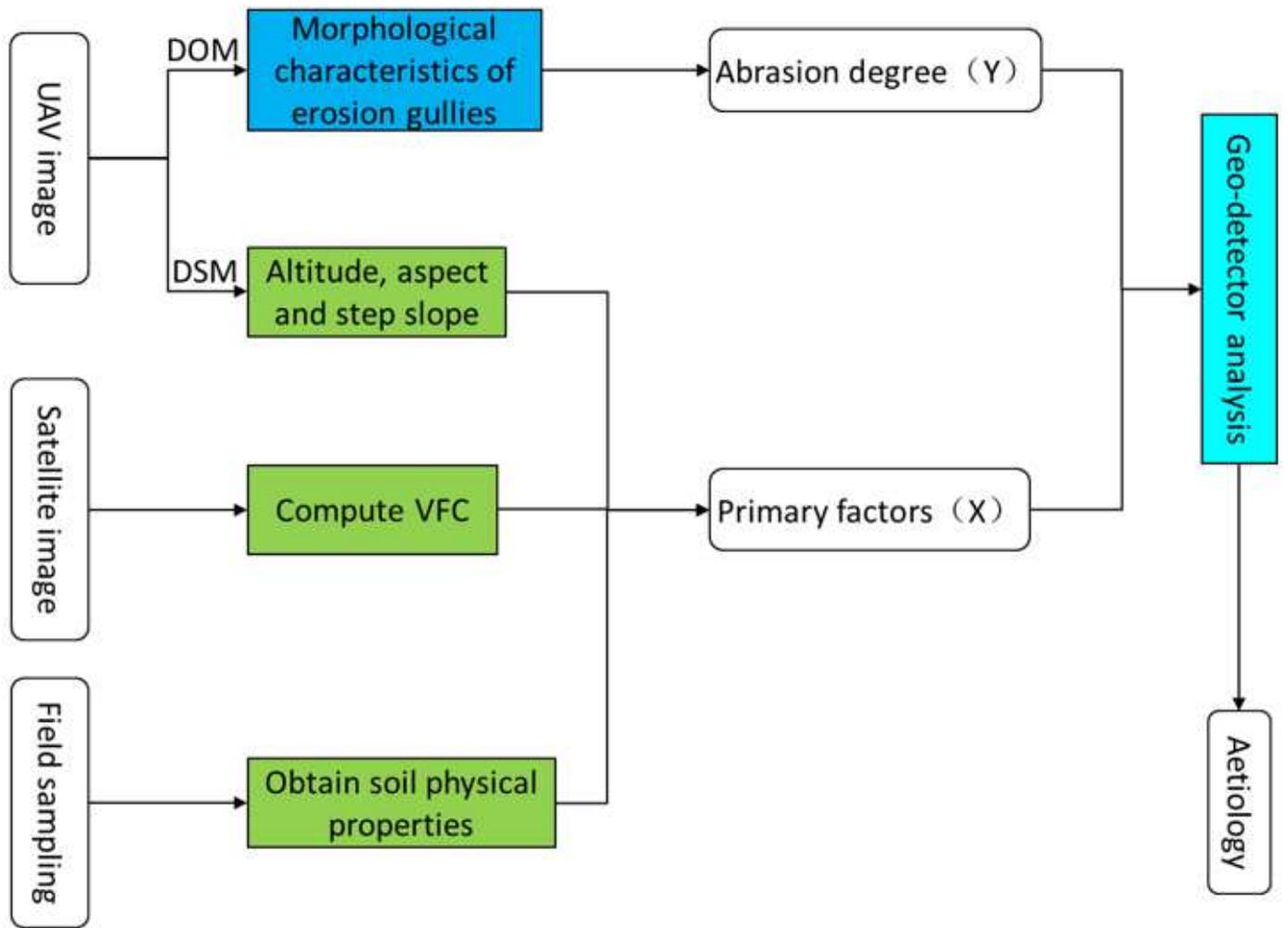


Figure 5

Methodology flow chart.

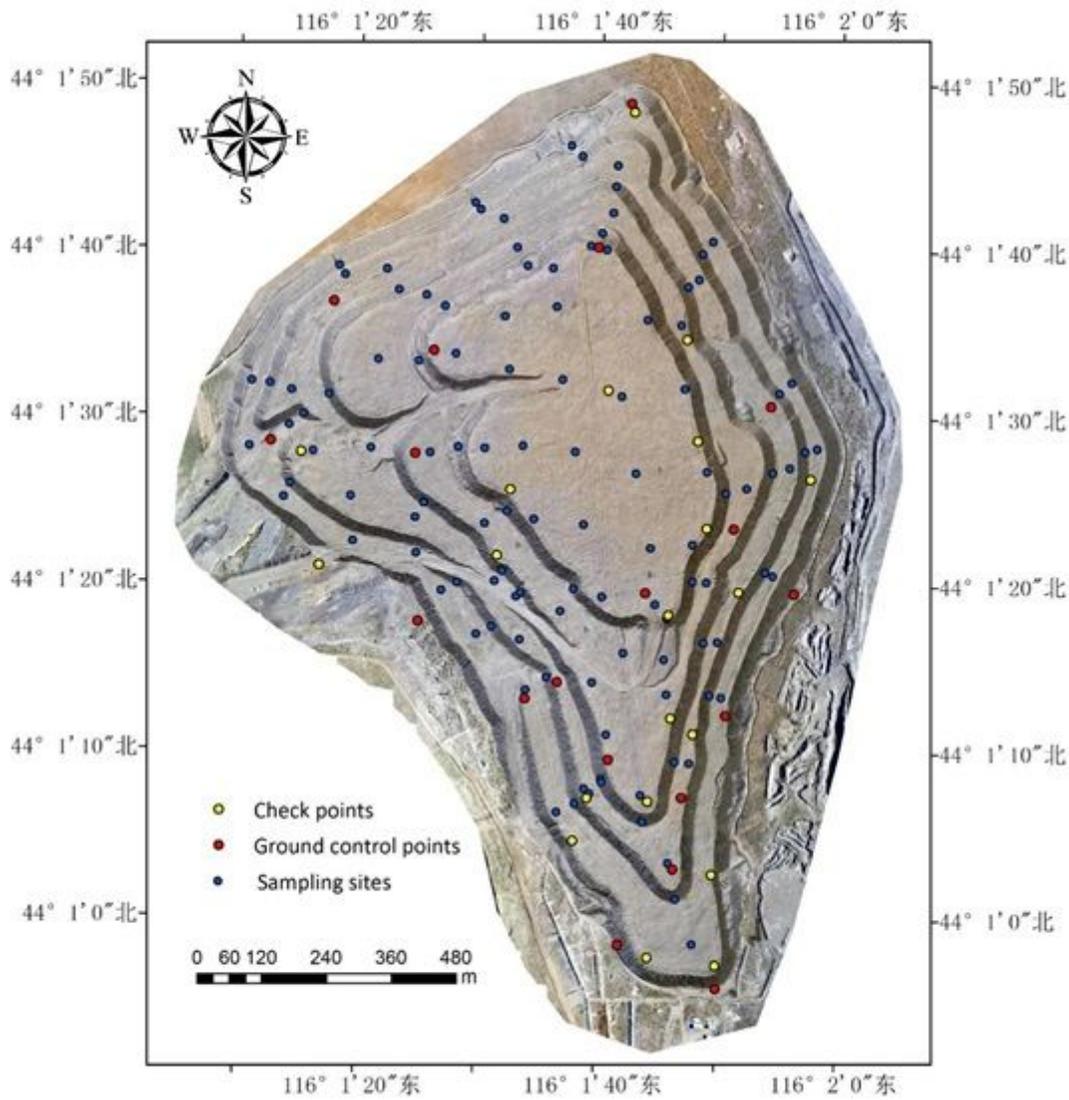


Figure 6

Sketch map of ground control points, control points and sampling sites.

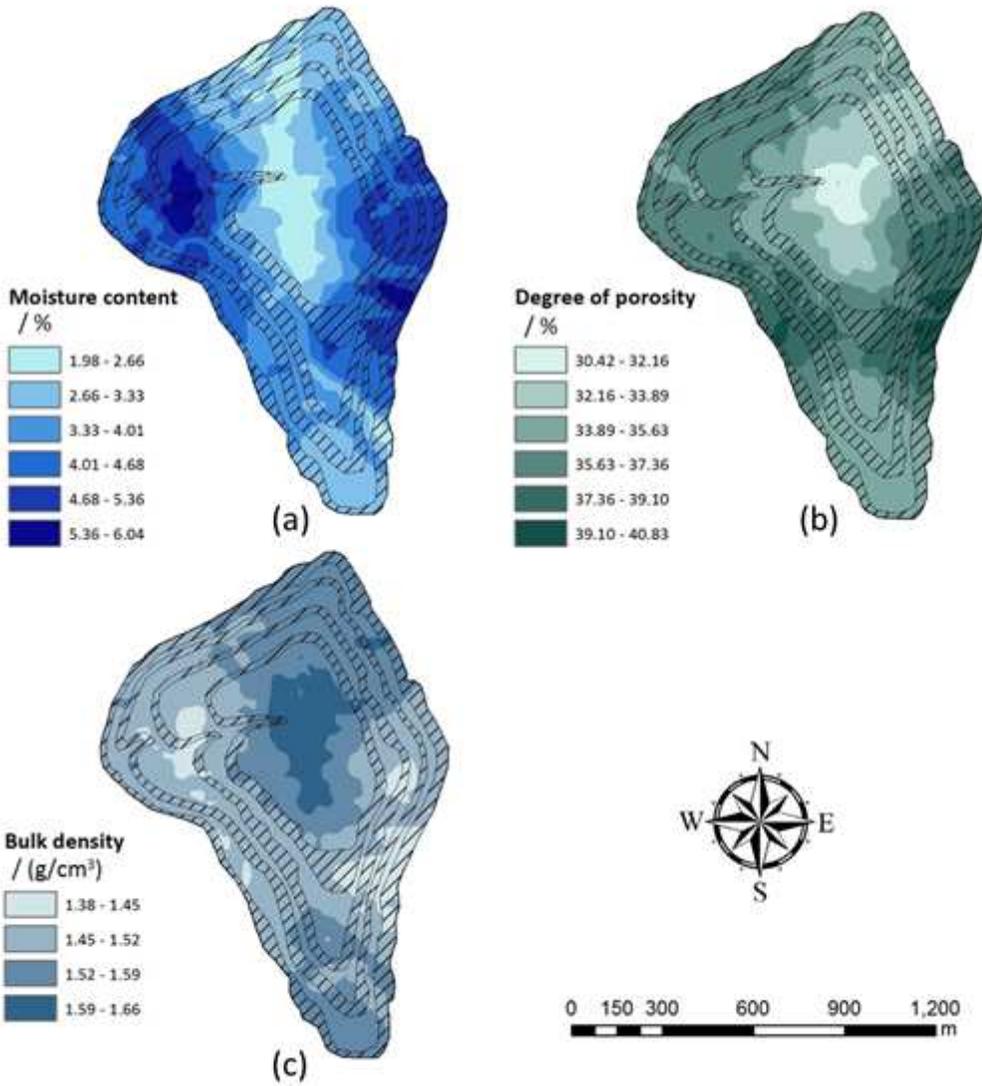


Figure 7

Soil physical properties. (a) Moisture content. (b) Degree of porosity. (c) Bulk density.

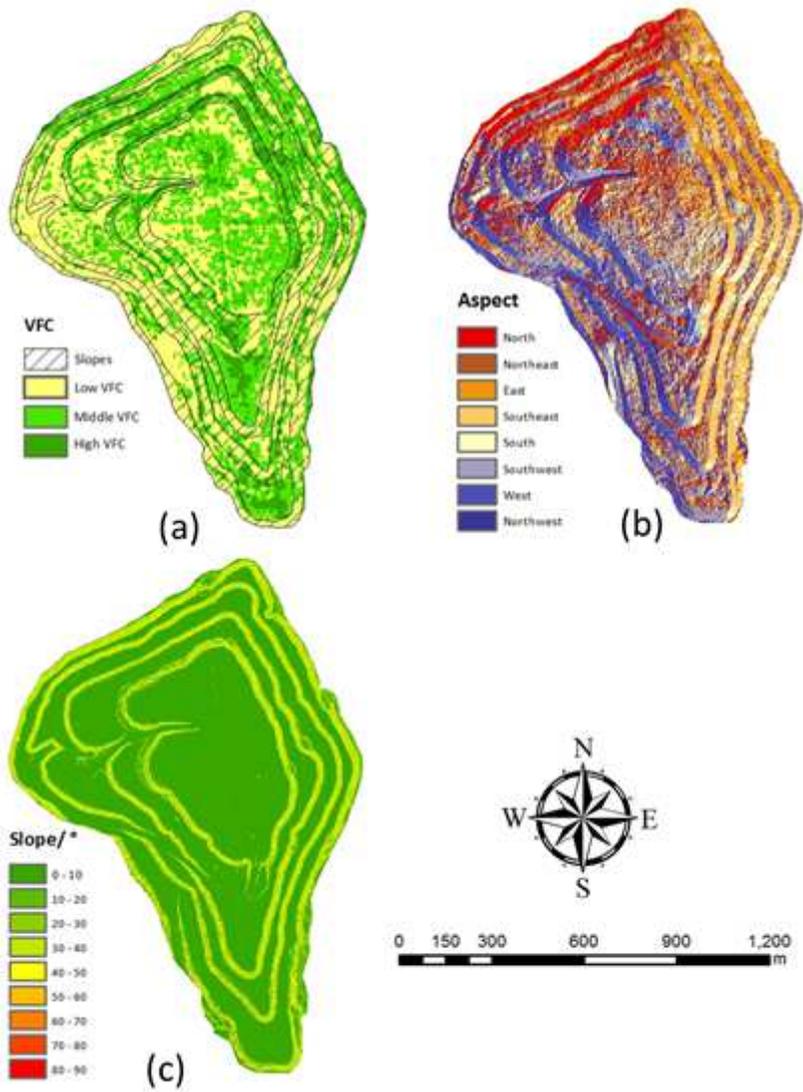


Figure 8

VFC, Slope aspect and gradient. (a)VFC. (b) Aspect. (c) Gradient.

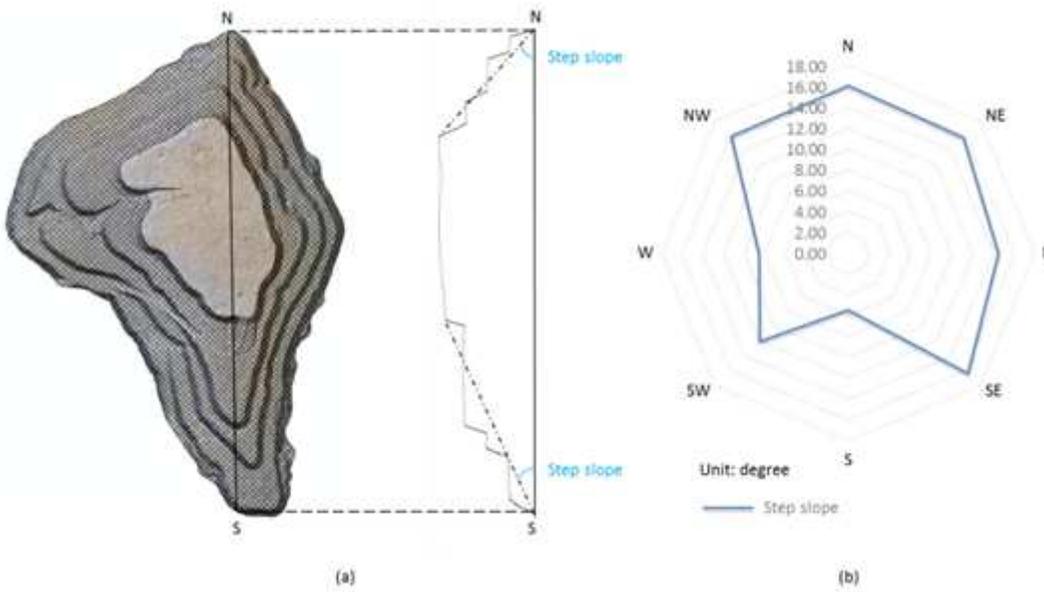


Figure 9

Step slope. (a) The elevation profile from the DSM is shown on the right, with its location shown on the map on the left. (b) Radar chart of the step slope gradient.

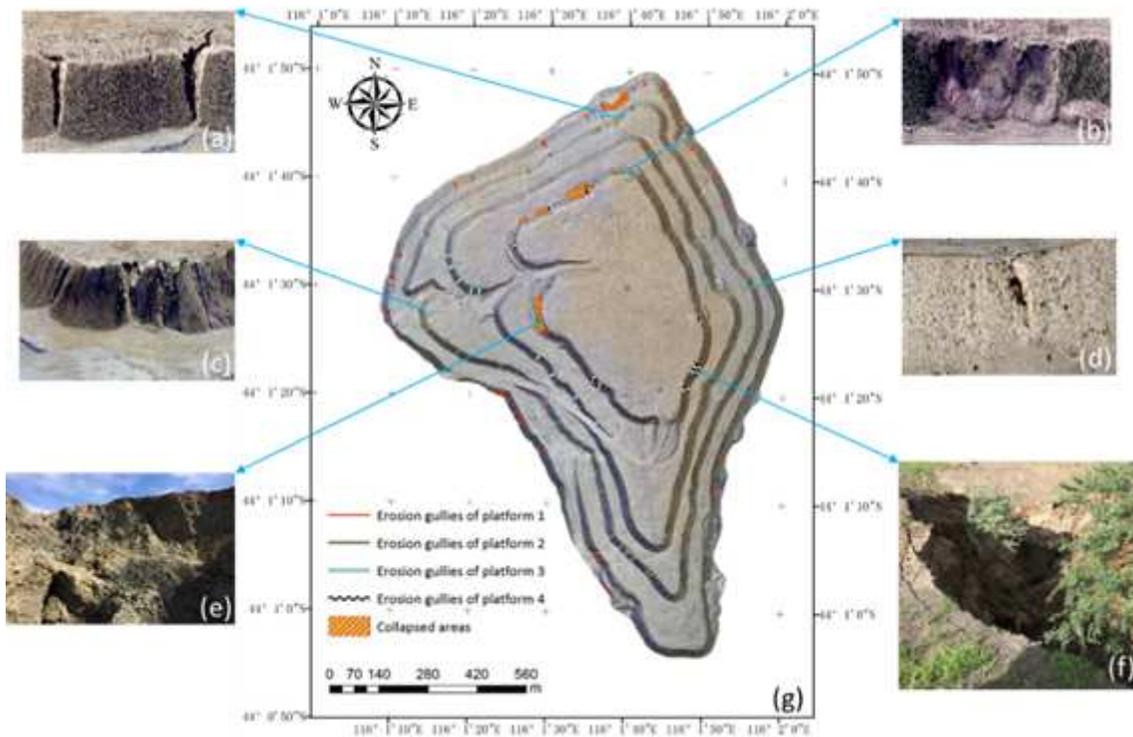


Figure 10

Geologic hazards at north dump. (a–d) Hazards shown in 3D models. (e–f) Photos taken in the field. (g) Distribution of hazards.

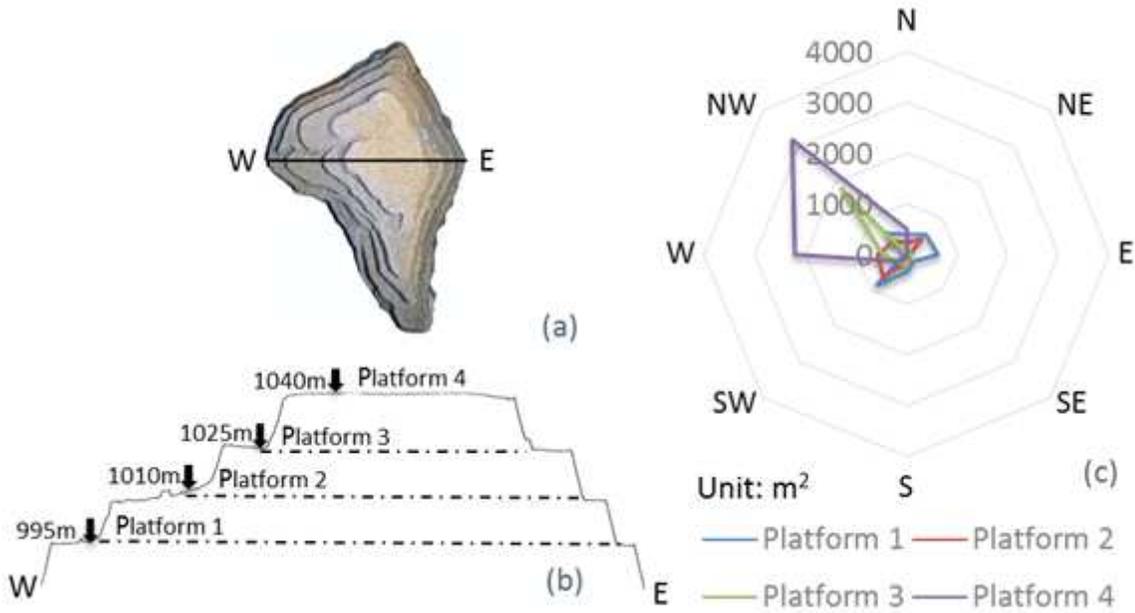


Figure 11

Erosion gullies radar chart. (a) Section line. (b) Transverse section. (c) Radar chart of erosion extent of each platform: the distance from the center represents the erosion extent.

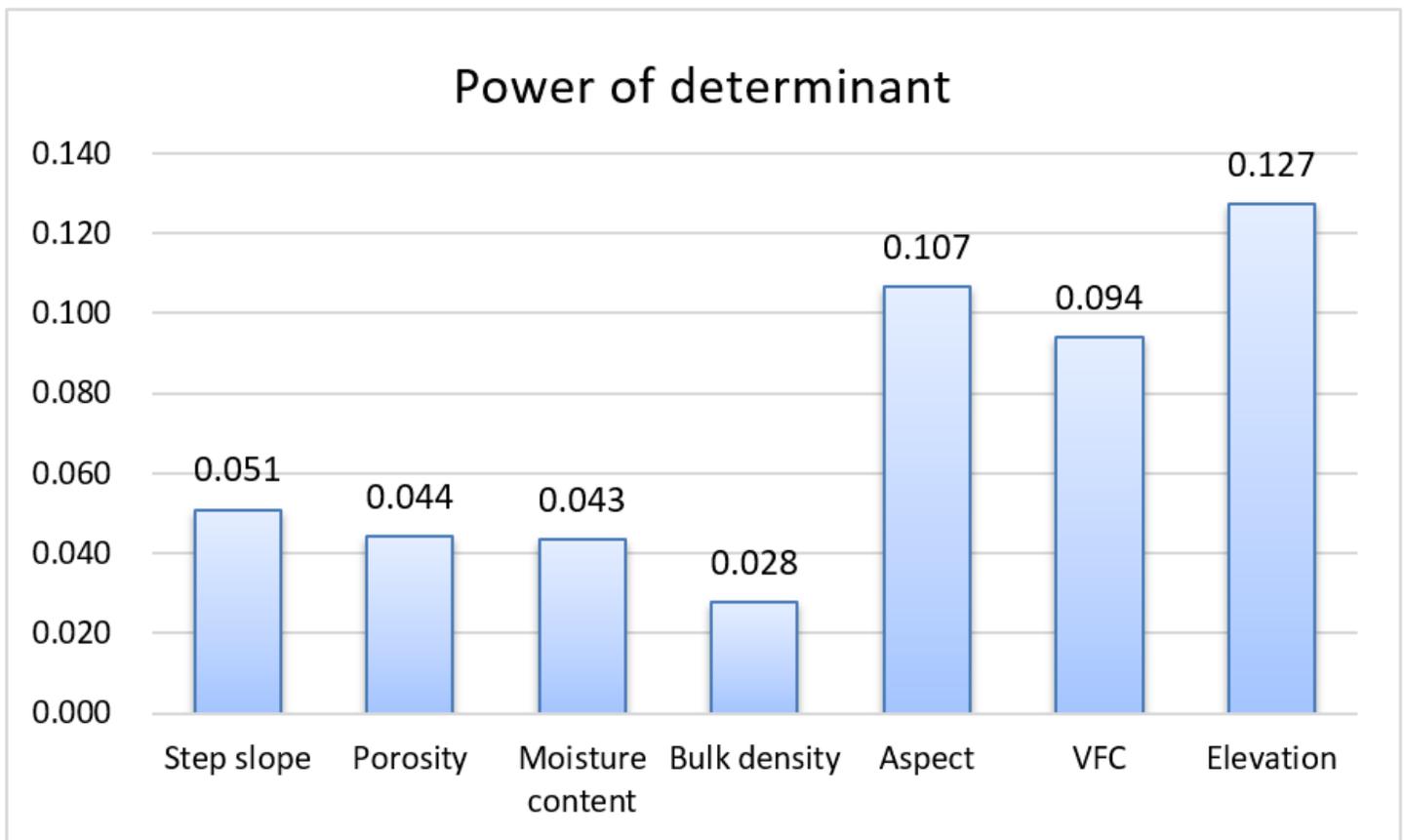


Figure 12

The influences of different driving factors on the degree of erosion.