

# Potential of Normalized Difference Vegetation Index for Mapping of Soft Clay Area in Paddy Fields of Kedah, Malaysia

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## Key words:

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**Abstract** Mapping of soft clay area in paddy fields uses remote sensing and GIS technique is the fastest way to obtain an accurate location of soft clay in a large scale area. It can be an alternative way to change conventional method like in-situ observation that is expensive and labor intensive. Therefore, this study aimed to investigate the normalized difference vegetation index (NDVI) to map soft clay area in paddy fields Kedah, Malaysia. To analyze soft clay area comprehensively, the study was carried out in three different periods; before paddy planting, after paddy planting and harvest. Ground-truth data of soft clay area was collected from study area during fieldwork activity and compared with NDVI values that produced from Landsat 8 image. Result of study showed NDVI map in period of before paddy planting could be a good indicator for mapping soft clay area because it gave a higher accuracy value than the other periods, with overall accuracy (85%) and kappa coefficient (0,84). Total area of soft clay from the highest value was showed in period of before paddy planting (1.856,97 ha), followed by after paddy planting (656,73 ha) and harvest (401,85 ha) periods, respectively.

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## 1. Introduction

Soft clay is infertile soil that characterized with low strength of soil. It affects rice production because agricultural machines cannot be operated at the soil. In Kedah, Malaysia, 8,107 ha of total paddy fields has experienced a great deal of soft clay problem (New Straits Time, 2013). With no proper soil improvement carried out at this area, slowing down in yield will increase. In natural condition, soft clay is distinguished with bare area and sometimes patches of water around paddy fields (Rendana et al., 2019a; Rendana et al., 2019b). They are some unique features that identified as the link with soft clay occurrence. But, to obtain an accurate ground-based observation of soft clay in a large scale area is difficult, labor intensive, expensive and time consuming. These challenges have led many researchers studied the remote sensing (RS) and geographical information system (GIS) technique because both offered a quick way in classifying land cover features (Rendana et al., 2016; Shafian & Maas, 2015).

The Landsat 8 satellite captures land cover image in a specific area that affected by soil and vegetation characteris-

tic. This is a basic of using remote sensing technique in assessing the land cover in various studies. Results of many studies have reported a great capacity of spectral band that derived from vegetation index for assessing soil and vegetation (Rendana et al., 2017). Over the years, the vegetation index has been applied in land cover evaluation in response on spatial and temporal variations (Mallick et al., 2009; Holzman et al., 2014).

To date, limited works have been done previously on the Landsat 8-based normalized difference vegetation index (NDVI) in mapping soft clay area. To distinguish the soft clay area in paddy fields using the NDVI index, firstly, we need some indicators from study area such as bare area and patches of water around the paddy fields. In addition, the soft clay soil has higher soil moisture content than normal soil. Many studies have successfully applied some vegetation indices from multispectral satellite in detecting soil moisture content. Multispectral indices like the soil adjusted vegetation index (SAVI), enhanced vegetation index (EVI) and vegetation condition index (VCI) have been widely used to

evaluate soil moisture and drought event (Huete et al., 2002). Specifically, the normalized difference vegetation index (NDVI) has been central to regional scale of soil moisture detection. The NDVI becomes a key parameter of the indices listed above and commonly applied in analyzing soil moisture content because it is capable to assess available moisture or water content in plant cell and soil (Wang et al., 2007; Chen et al., 2014; Ahmed et al., 2017). As a whole, it can be a suitable parameter for the study. Therefore, the study aimed to investigate potential of normalized difference vegetation index (NDVI) in mapping of soft clay area in the paddy fields. Findings from the study were expected to give further understanding of remote sensing and GIS technique in evaluating spatial variation of the soft clay area.

**2. The Methods**

Study area was located in MADA paddy cultivation area (Region-III) Kedah, Malaysia (Figure 1). It was divided by three regions; B-III Region (6°6'00.00"N, 100°27'00.00"E), C-III Region (6°3'00.00"N, 100°24'00.00"E) and D-III Region (6°3'00.00"N, 100°30'00.00"E). Total area of study was about 16.284,60 ha. The area has experienced low rice yield because of soft clay problem (Nadzim et al., 2014). Soft clay was characterized with high compressibility and low strength (Azizul, 2008; Rendana et al., 2018). The low soil strength value lead to decline ground pressure capability of agricultural machine.

Landsat 8 operational land imager (OLI) satellite image was downloaded from USGS Earth Explorer website. Landsat 8 OLI & thermal infrared sensor (TIRS) had nine spectral bands with spatial resolution of 30m for band 1 to 7 and 9, 15m for panchromatic band 8 and 100m for thermal band 10 and 11. In this study, mapping of soft clay area was carried out in three periods of paddy planting such as before paddy planting, after paddy planting and harvest. It was done to obtain the best result for mapping of soft clay area.

Atmospheric correction was a process for minimizing negative effects on satellite image before it could be used for land cover analysis such as fog, dust, tiny particle and the other disturbances on the atmosphere. It consisted of two main processes; (i) top of Atmosphere (TOA) reflectance and (ii) TOA reflectance with the sun angle correction. Top of Atmosphere (TOA) reflectance was a conversion of digital number (DN) from radiance to reflectance value. The atmospheric correction was calculated using formula from USGS (U.S. Geological Survey) in equation 1.

$$R_r = M_1 Q_{cal} + A_r \tag{1}$$

- $R_r$  = TOA reflectance without sun angle correction,
- $M_1$  = parameter "REFLECTANCE\_MULT\_BAND\_x" in metadata file.
- $Q_{cal}$  = standardized pixel number.
- $A_r$  = parameter "REFLECTANCE\_MULT\_BAND\_x" in metadata file.

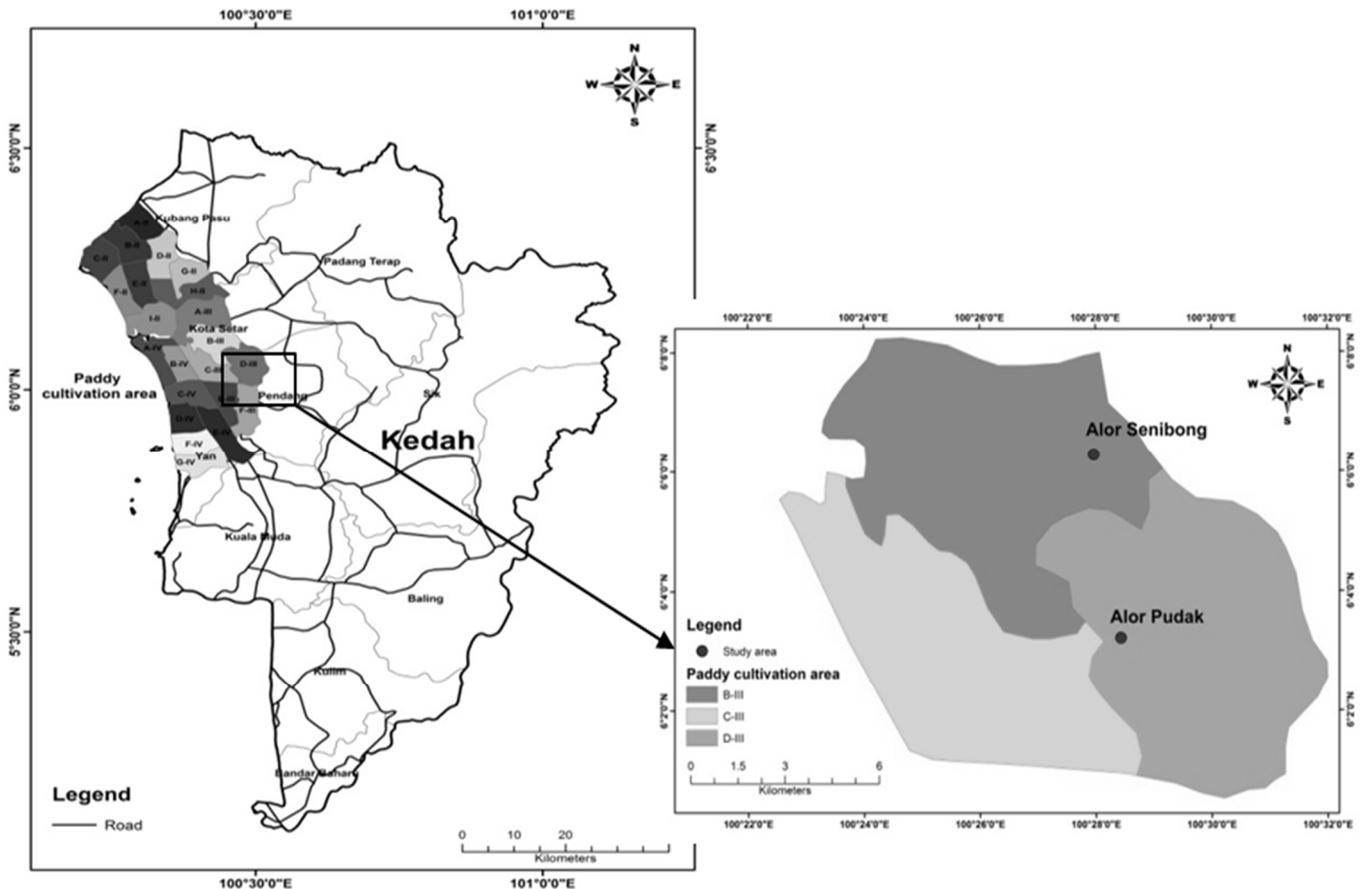


Figure 1. Location of study area.

$R_1 = R_r / \cos(\text{sun zenith angle})$  or  $R_1 = R_r / \sin(\text{sun elevation angle})$

$R_1$  = TOA reflectance,  $R_r$  = TOA reflectance without correction.

Sun elevation angle = parameter “SUN\_ELEVATION” in metadata file.

Sun zenith angle = 90 - Sun elevation angle.

Landsat 8 images were downloaded in three periods of paddy planting; before paddy planting, after paddy planting and harvest period. In many studies, the normalized difference vegetation index (NDVI) has been used to study canopy growth and fertility of plant (Gopalapillai & Tian, 1999). Near-infrared band was a sensitive in changing of chlorophyll content thus it could be used to determine land cover features in a certain area. The NDVI capability for assessing land cover features was a very useful method for mapping of soft clay area. Because of the soft clay had unique characteristics such as bare area and patches of water at soil surface (Figure 2). For this reason, the NDVI could be used to detect available moisture or water content in the soil. Therefore, the NDVI became a good parameter selected in the study. In general, the NDVI was calculated using the near-infrared band 5 and red band 4 from spectral band of the Landsat 8 image as shown in equation 2. NDVI values normally ranged from -1 to 1. For dense canopy or vegetation area, it showed the NDVI value near 1, while for bare area or water body showed the NDVI value near -1. It was classified in accordance classification from Rouse et al. (1974). Overall of GIS analysis was conducted using ArcGIS software Ver. 10.

$$NDVI = (NIR \text{ (band 5)} - Red \text{ (band 4)}) / (NIR \text{ (band 5)} + Red \text{ (band 4)}) \quad (2)$$

Sampling activity was conducted to obtain ground-truth data of soft clay location in study area. Total of 31 sampling points has been successfully recorded using DGPS Trimble Geo XM (Figure 3). It was used to validate NDVI maps from

$$K = \frac{N \sum_{i=1}^r x_{ii} - \sum_{i=1}^r (x_i + Xx_{+1})}{N^2 - \sum_{i=1}^r (x_{ii} Xx_{+1})} \quad (3)$$

GIS analysis. Accuracy assessment was calculated using kappa statistic. The kappa statistic formula has been shown in equation 3 that referred from Jensen (1996). Kappa coefficient was equal to 1 implied a perfect agreement while for the value closed to zero implied the agreement was no better than it would be expected by chance.

$r$  = number of rows and columns in error matrix.

$N$  = total number of observations (pixels).

$X_{ii}$  = observation in row  $i$  and column  $i$ .

$X_{i+}$  = marginal total of row  $i$ .

$X_{+i}$  = marginal total of column  $i$ .

### 3. Results and Discussion

#### Normalized difference vegetation index (NDVI) classification

In this study, classification of NDVI value was adopted from Jwan et al. (2013). Sparse vegetation ranged from 0,2 to 0,4, moderate vegetation ranged from 0,4 to 0,6 and dense vegetation ranged from 0,6 to 0,9. While, for the NDVI value was less than 0,2, it indicated water body or area without vegetation cover. Based on result of study and ground-truth data, the study believed the NDVI value (<0,2) indicated the soft clay area. Nadzim et al. (2014) stated the soft clay had soil strength was less than 0,40 MPa thus it inhibited paddy growth. The soft clay area was exhibited by blue color at NDVI maps (Figure 4). Bambang and Ahmad (2014) explained both water body and bare area had a lower NDVI value than vegetation area because of chlorophyll content. The chlorophyll absorbed more sun light in visible spectrum, especially in red spectrum. Total of the soft clay area obtained through GIS analysis in period of harvest (401,85 ha) was a slightly lower than period of after paddy planting (656,73 ha) and before paddy planting (1.856,97 ha) (Table



Figure 2. Pictures of soft clay in study area.

1). Many soft clay areas were covered by irrigation water in period of after paddy planting that resulted in decreasing of them. Contrarily, in period of harvest and before planting, paddy fields were in drying condition that attributed in increasing total of reflectance of soft clay area.

There was a significant change at dense vegetation class (NDVI 0,6-0,9) in period of after paddy planting. It showed increasing total area of the class (8,881.65 ha). It might occur because paddy plants have grown with high density thus it raised NDVI value. This notion was consistent with a finding of Ajith et al. (2017) who found the NDVI value in paddy fields during growing season was around 0,81. For harvest period, sparse vegetation class (NDVI 0,2-0,4) was a predominant in study area. It dominated approximately of 48% from total area of study. Mosleh and Hassan (2014) found the NDVI value of paddy fields in Bangladeshi during harvest season was around 0,45. The value was a slightly higher than current study. When it was compared with previous studies, this finding might be affected by several aspects such as location of area and species of paddy. The variation of paddy species could reflect distinct chlorophyll and NDVI values.

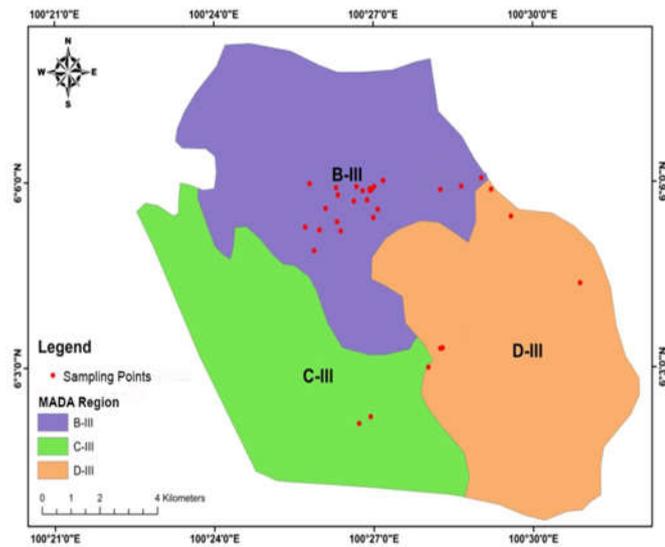


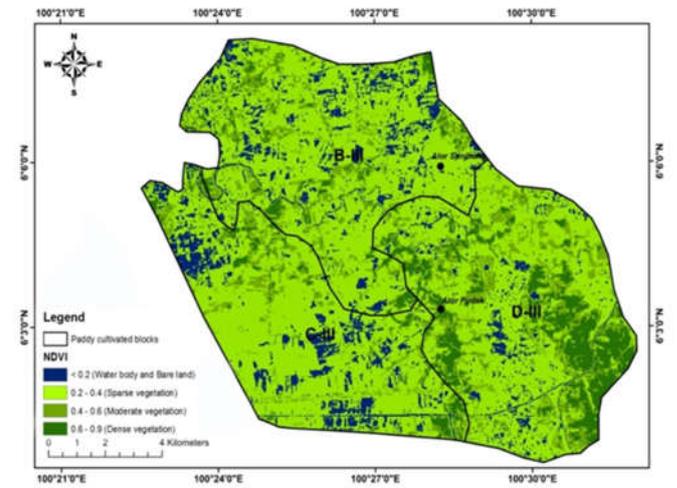
Figure 3. Distribution of sampling point in study area.

Table 1. Total area of NDVI classes in three periods of paddy planting.

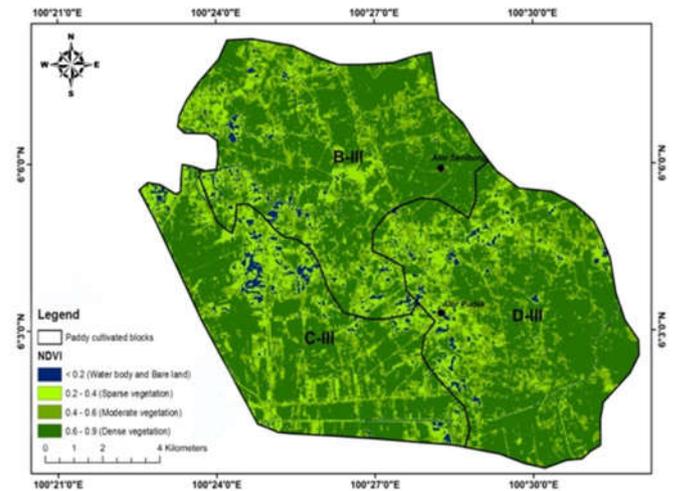
NDVI Classes	Period of Paddy Planting		
	Before Planting (ha)	After Planting (ha)	Harvest (ha)
Water Body or Bare Area (< 0,2)	1,856.97	656,73	401,85
Sparse Vegetation (0,2 - 0,4)	10,480.23	2,487.24	7,870.23
Moderate Vegetation (0,4 - 0,6)	2,512.26	4,258.98	4,523.58
Dense Vegetation (0,6 - 0,9)	1,435.14	8,881.65	3,488.94

Accuracy assessment

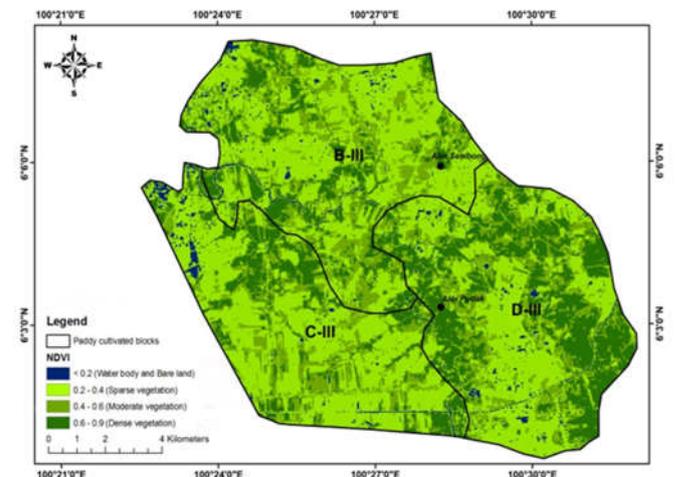
Accuracy assessment was carried out using confusion matrix on NDVI map (Table 3-5). In this study, the highest value of overall accuracy and kappa coefficient was showed in the period of before paddy planting with values of 85%



(a)



(b)



(c)

Figure 4. NDVI maps in study area (a) period of before paddy planting, (b) period of after paddy planting and (c) period of harvest.

Table 2. Accuracy assessment of NDVI classes based on statistical parameters.

NDVI Classes	Periods											
	Before Paddy Planting				After Paddy Planting				Harvest Period			
	Parameters				Parameters				Parameters			
	OE	CE	UA	PA	OE	CE	UA	PA	OE	CE	UA	PA
Water Body or Bare Area	0,15	0,104	0,85	0,85	1	0	0	0	0,70	0	1	0,30
Sparse Vegetation	0,15	0,111	0,71	0,85	0,95	0,24	0,05	0,05	0,45	0,188	0,46	0,55
Moderate Vegetation	0,15	0,016	0,94	0,85	0,70	0,121	0,40	0,30	0,70	0,135	0,38	0,30
Dense Vegetation	0,15	0,016	0,94	0,85	0,75	0,36	0,16	0,25	0,10	0,258	0,53	0,90

\*OE: Omission Error; CE: Commission Error; PA: Producer’s Accuracy; UA: User’s Accuracy.

Table 3. The matrix table of NDVI classification before paddy planting period.

NDVI Classes	Reference Value				Total	Correct Sampled
	Water Body or Bare Area	Sparse Vegetation	Moderate Vegetation	Dense Vegetation		
Water Body or Bare Area	17	3	0	0	20	17
Sparse Vegetation	3	17	2	2	24	17
Moderate Vegetation	0	0	17	1	18	17
Dense Vegetation	0	0	1	17	18	17
Total	20	20	20	20	80	68
Overall Accuracy					85%	
Kappa (k) Value					0.84	

Table 4. The matrix table of NDVI classification after paddy planting period.

NDVI Classes	Reference Value				Total	Correct Sampled
	Water Body or Bare Area	Sparse Vegetation	Moderate Vegetation	Dense Vegetation		
Water Body or Bare Area	0	1	5	7	13	0
Sparse Vegetation	5	1	6	8	20	1
Moderate Vegetation	2	7	6	0	15	6
Dense Vegetation	13	11	3	5	32	5
Total	20	20	20	20	80	12
Overall Accuracy					15%	
Kappa (K) Value					0.10	

and 0,84, respectively. Misclassification rate was obtained from commission error (CE) and omission error (OE) values. CE was defined as number of points that was included in a certain class while they did not belong to the class. Whilst, OE defined as number of points that did not fall in a certain class while they were belong to the class. Sparse vegetation class showed the highest CE (0,111) and OE (0,15), while moderate and dense vegetation classes showed the lowest CE (0,016) and OE (0,15) (Table 2). The accuracy value normally depended on spectral reflectance of land cover (Kamaruzaman et al., 2009).

User’s accuracy value ranged from 71% to 94% while producer’s accuracy value was 85%. Soft clay area class that

classified as water body or bare area was found to be more reliable with 85% of user’s accuracy. With the high accuracy value, it indicated NDVI map in period of before paddy planting was the most suitable period for mapping of soft clay area. In this study, the overall kappa coefficient (0,84) was classified as a good rating criteria of kappa statistic. The mapping of soft clay area was more accurate in period of before paddy planting than the other periods. This was because soft clay area tended to be apparently detected in drying condition. In drying condition, there was not any plant canopy or water that constrained reflectance of soft clay area.

In the study area, soft clay area appeared as bare spots around paddy fields that consisted of patches of water. To

Table 5. The matrix table of NDVI classification in the harvest period.

NDVI Classes	Reference Value				Total	Correct Sampled
	Water Body or Bare Area	Sparse Vegetation	Moderate Vegetation	Dense Vegetation		
Water Body or Bare Area	6	0	0	0	6	6
Sparse Vegetation	10	11	1	2	24	11
Moderate Vegetation	3	7	6	0	16	6
Dense Vegetation	1	2	13	18	34	18
Total	20	20	20	20	80	41
Overall Accuracy					51.2%	
Kappa (K) Value					0.49	

prove the assumption, we have conducted accuracy analysis using kappa statistic. Result showed 85% of NDVI value that classified as bare and water body area were accord with ground-truth data. In period of after paddy planting, many soft clay areas were covered by water because paddy fields were drained. Because of this condition, many soft clay areas were not able to be identified by NDVI index. While, in harvest period, paddy fields were completely dried thus many soft clay areas have also been dried that resulted in decreasing of total area of soft clay or NDVI class (<0,2). As a whole, for the best accuracy, the study recommended mapping of soft clay area should be conducted in the first period or before paddy planting.

#### 4. Conclusion

Result of study concluded NDVI index had a good potential for mapping of soft clay area in paddy fields. The best estimation of the NDVI index for mapping of soft clay area was showed in period of before paddy planting with overall accuracy (85%) and kappa coefficient (0,84). Total area of the soft clay from the highest value was showed in period of before paddy planting (1.856,97 ha) followed by after paddy planting (656,73 ha) and harvest (401,85 ha) periods, respectively. A low total area of soft clay in period of after paddy planting because the paddy fields were drained by irrigation water thus many soft clay areas were covered by water. While, in harvest period, the paddy fields were dried completely thus many soft clay areas were also dried that resulted in decreasing of total area of them. Therefore, based on result of the study, it recommended using the period of before paddy planting for mapping of soft clay area. Soft clay area was detected accurately when the paddy field was in drying condition but it was not too dry.

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