Analisis Perbandingan Algoritma Pattern Matching Knuth-Morris-Pratt dan Boyer-Moore Untuk Metode Pemrosesan Citra Template Matching

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Abstract— Pemrosesan citra template matching adalah teknik penting dalam pengolahan citra yang digunakan untuk mencari dan mencocokkan pola tertentu dalam citra. Dalam makalah ini, proses pattern matching digunakan untuk pemrosesan citra template matching. Selain itu, dilakukan analisis perbandingan antara dua algoritma populer dalam proses pattern matching, yaitu algoritma Knuth-Morris-Pratt (KMP) dan algoritma BoyerMoore. Algoritma KMP dan Boyer-Moore memiliki pendekatan yang berbeda dalam mencocokkan pola. Algoritma KMP menggunakan informasi dari pencarian sebelumnya untuk mempercepat proses pencarian berikutnya, sedangkan algoritma Boyer-Moore menggunakan informasi dari pola yang ingin dicari dan melakukan penggeseran pintar berdasarkan karakter terakhir yang tidak cocok. Penelitian ini bertujuan untuk membandingkan kinerja kedua algoritma berdasarkan kecepatan eksekusi, banyak proses perbandingan, dan akurasi hasil pencocokan.

Import Modules

```
# import modules
import cv2
import numpy as np
import matplotlib.pyplot as plt
import copy
import pandas as pd
import random
import seaborn as sns
import time
```

Definisikan Knuth-Morris-Pratt Search, Boyer-Moore Search, dan Brute Force Search

```
def KMPSearch(pattern, txt):
    patternLen = len(pattern)
    textLen = len(txt)

compareCounter = 0
    results = []

# create lps[] that will hold the longest prefix suffix
# values for pattern
```

```
lps = [0]*patternLen
    patternIdx = 0 # index for pat[]
    # Preprocess the pattern (calculate lps[] array)
    computeLPSArray(pattern, patternLen, lps)
    textIdx = 0 # index for txt[]
    while (textLen - textIdx) >= (patternLen - patternIdx):
        compareCounter += 1
        if pattern[patternIdx] == txt[textIdx]:
            textIdx += 1
            patternIdx += 1
        if patternIdx == patternLen:
            results += [textIdx - patternIdx]
            patternIdx = lps[patternIdx - 1]
        # mismatch after j matches
        elif textIdx < textLen and pattern[patternIdx] !=</pre>
txt[textIdx]:
            if patternIdx != 0:
                patternIdx = lps[patternIdx-1]
            else:
                textIdx += 1
    return results, compareCounter
# Function to compute LPS array
def computeLPSArray(pat, plen, lps):
    len = 0 # length of the previous longest prefix suffix
    lps[0] = 0
    i = 1
    while i < plen:
        if pat[i] == pat[len]:
            len += 1
            lps[i] = len
            i += 1
        else:
            if len != 0:
                len = lps[len-1]
            else:
                lps[i] = 0
                i += 1
def BMSearch(pattern, text):
    # Preprocessing
```

```
patternLen = len(pattern)
    textLen = len(text)
    compareCounter = 0
    last = {}
    for i in range(patternLen):
        last[pattern[i]] = i
    # Searching
    i = patternLen - 1 # Index of the last character in the pattern
   j = patternLen - 1 # Index of the last character in the text
being examined
    results = []
    while j < textLen:</pre>
        compareCounter += 1
        if pattern[i] == text[j]:
            if i == 0:
                results.append(j)
                i = patternLen - 1
                j += patternLen
            else:
                i -= 1
                i -= 1
        else:
            if text[j] in last:
                j += patternLen - min(i, last[text[j]] + 1)
                j += patternLen
            i = patternLen - 1
    return results, compareCounter
def BFSearch(pattern, text):
    # Preprocessing
    patternLen = len(pattern)
    textLen = len(text)
    compareCounter = 0
    # Searching
    results = []
    for i in range(textLen - patternLen + 1):
        j = 0
        compareCounter += 1
        while j < patternLen and pattern[j] == text[i + j]:</pre>
            compareCounter += 1
```

```
i += 1
        if j == patternLen:
            results.append(i)
    return results, compareCounter
def templateMatch(template, image, algo):
    # preprocessing
    # ubah citra dan template menjadi 8-bit
    image = cv2.cvtColor(image, cv2.COLOR RGB2GRAY)
    template = cv2.cvtColor(template, cv2.COLOR RGB2GRAY)
    tempRow = len(template)
    tempCol = len(template[0])
    pattern = template[0].flatten().tolist()
    n = len(image)
    compareCounter = 0
    # memanggil algoritma pattern matching yang sesuai
    for i in range(n):
        text = image[i].flatten().tolist()
        if algo == 'kmp':
            found, cnt = KMPSearch(pattern, text)
        elif algo == 'bm':
            found, cnt = BMSearch(pattern, text)
        else:
            found, cnt = BFSearch(pattern, text)
        compareCounter += cnt
        # mengecek validitas template pada citra
        for j in found:
            if (np.array equal(template, image[i:i+tempRow,
j:j+tempCol])):
                return i, j, compareCounter
    return -1, -1, compareCounter
```

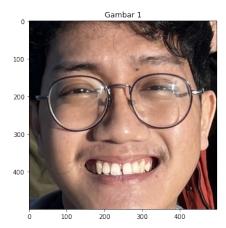
Membaca dan Menampilkan Citra dan Template

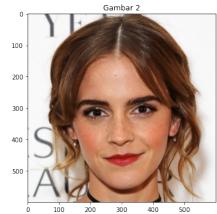
```
# Membaca citra dan template
image = cv2.imread(r'images\image.png')
image = cv2.cvtColor(image, cv2.COLOR_BGR2RGB)

template1 = cv2.imread(r'images\template.png')
template1 = cv2.cvtColor(template1, cv2.COLOR_BGR2RGB)
```

```
template2 = cv2.imread(r'images\emma watson.png')
template2 = cv2.cvtColor(template2, cv2.COLOR BGR2RGB)
fig, axes = plt.subplots(1, 3, figsize=(15, 5))
# Display the first image on the first subplot
axes[0].imshow(image)
axes[0].set title('Citra')
# Display the second image on the second subplot
axes[1].imshow(template1)
axes[1].set_title('Gambar 1')
# Display the third image on the third subplot
axes[2].imshow(template2)
axes[2].set title('Gambar 2')
# Adjust the spacing between subplots
plt.tight layout()
# Show the figure
plt.show()
```







Mendapatkan Hasil Template Matching Menggunakan KMP, BM, dan Brute Force Pada Gambar 1

```
t1 = time.time()
i3, j3, bfCount = templateMatch(template1, image, 'bf')
t2 = time.time()
i1, j1, kmpCount = templateMatch(template1, image, 'kmp')
t3 = time.time()
i2, j2, bmCount = templateMatch(template1, image, 'bm')
t4 = time.time()

bfTime = t2 - t1
kmpTime = t3 - t2
```

```
bmTime = t4 - t3
print(f"Ketiga hasil cocok : {(i1,j1) == (i2,j2) == (i3,j3)}")
print(f"Koordinat Gambar 1 : ({i1}, {j1})\n")
print(f"Waktu Brute Force : {bfTime}\ts")
print(f"Waktu Knuth-Morris-Pratt : {kmpTime}\ts")
print(f"Waktu Boyer-Moore : {bmTime}\ts\n")
print(f"Banyak Perbandingan Brute Force : {bfCount}")
print(f"Banyak Perbandingan Knuth-Morris-Pratt : {kmpCount}")
print(f"Banyak Perbandingan Boyer-Moore : {bmCount}")
Ketiga hasil cocok : True
Koordinat Gambar 1 : (2000, 950)
Waktu Brute Force
                       : 1.9515295028686523
Waktu Knuth-Morris-Pratt : 2.637633800506592
Waktu Boyer-Moore : 0.21509647369384766
Banyak Perbandingan Brute Force : 3543720
Banyak Perbandingan Knuth-Morris-Pratt: 3539769
                               : 12310
Banyak Perbandingan Boyer-Moore
```

Menampilkan Hasil Template Matching Pada Gambar 1

```
result = copy.copy(image)
w, h = template1.shape[:-1]

if ((i1, j1) != (-1, -1)):
    cv2.rectangle(result, (j1, i1), (j1 + h, i1 + w), (0, 255, 255),
50)

plt.figure(figsize=(10, 10))
plt.imshow(result)
plt.show()
```



Mendapatkan Hasil Template Matching Menggunakan KMP, BM, dan Brute Force Pada Gambar 2

```
t5 = time.time()
i4, j4, bfCount2 = templateMatch(template2, image, 'bf')
t6 = time.time()
i5, j5, kmpCount2 = templateMatch(template2, image, 'kmp')
t7 = time.time()
i6, j6, bmCount2 = templateMatch(template2, image, 'bm')
t8 = time.time()
bfTime2 = t6 - t5
kmpTime2 = t7 - t6
bmTime2 = t8 - t7
print(f"Ketiga hasil cocok : {(i4,j4) == (i5,j5) == (i6,j6)}")
print(f"Koordinat Gambar 2 : ({i4}, {i4})\n")
print(f"Waktu Brute Force : {bfTime2}\ts")
print(f"Waktu Knuth-Morris-Pratt : {kmpTime2}\ts")
print(f"Waktu Boyer-Moore : {bmTime2}\ts\n")
print(f"Banyak Perbandingan Brute Force
                                        : {bfCount2}")
print(f"Banyak Perbandingan Knuth-Morris-Pratt : {kmpCount2}")
print(f"Banyak Perbandingan Boyer-Moore : {bmCount2}")
Ketiga hasil cocok : True
Koordinat Gambar 2: (-1, -1)
Waktu Brute Force : 2.214761972427368
Waktu Knuth-Morris-Pratt : 5.476287364959717
                                              S
Waktu Boyer-Moore : 0.8662610054016113 s
Banyak Perbandingan Brute Force
                                     : 6731432
Banyak Perbandingan Knuth-Morris-Pratt : 6729497
Banyak Perbandingan Boyer-Moore
                               : 15729
```

Menampilkan Hasil Template Matching Pada Gambar 2



Generate 10 Random Sampel Lalu Ambil Hasil Template Matching Pada Sampel

```
data = {
    'size': [],
    'BF': [],
    'KMP': [],
    'BM': [],
    'BF time (ms)': [],
    'KMP time (ms)': [],
    'BM time (ms)': [],
    'accurate': [],
    'row': [],
    'col': [],
    'width': [],
    'height': []
}
w, h = image.shape[:-1]
for i in range(10):
    r1 = random.randint(0, w-10)
    r2 = random.randint(r1+1, w-1)
    c1 = random.randint(0, h-10)
    c2 = random.randint(c1+1, h-1)
    data['size'].append((r2-r1) * (c2-c1))
    temp = image[r1:r2, c1:c2]
    t1 = time.time()
    i1, j1, bfCount = templateMatch(temp, image, 'bf')
    t2 = time.time()
    i2, j2, kmpCount = templateMatch(temp, image, 'kmp')
    t3 = time.time()
    i3, j3, bmCount = templateMatch(temp, image, 'bm')
    t4 = time.time()
    data['BF'].append(bfCount)
    data['BF time (ms)'].append((t2 - t1)*1000)
    data['KMP'].append(kmpCount)
    data['KMP time (ms)'].append((t3 - t2)*1000)
    data['BM'].append(bmCount)
    data['BM time (ms)'].append((t4 - t3)*1000)
    data['accurate'].append(r1 == i1 == i2 == i3 and c1 == j1 == j2 ==
j3)
    data['row'].append(r1)
    data['col'].append(c1)
    data['width'].append(r2-r1)
    data['height'].append(c2-c1)
```

```
df = pd.DataFrame(data)
df
      size
                 BF
                         KMP
                                  BM
                                      BF time (ms)
                                                     KMP time (ms)
0
                                        3410.575390
                                                       5075.389385
       756
            8572875
                     8495186
                              254418
                                                       2257.345438
1
    498892
            3518453
                     3517447
                               13933
                                        1132.727146
2
     34380
            3876302
                     3864823
                               31865
                                        1296.786070
                                                       4106.243134
3
    101904
            2671851
                     2645565
                               28832
                                                       1615.803242
                                        1046.411037
4
            2923123
                                                       2228.004456
    222865
                     2885193
                               17654
                                        1275.613546
5
    898586
            2418976
                     2389150
                               53548
                                        819.306374
                                                       1633.229733
                                         907.504559
                                                       2839.982986
6
    205226
            2366499
                     2356548
                               13676
7
   2805720
             219528
                      197499
                                1813
                                        103.229284
                                                        317.312241
            1915169
                     1899815
                                5141
                                        1106.501102
                                                       2198.771715
8
   881973
9
                                        1454.519033
                                                       2907.816172
   1204800
           3517851
                     3484468
                               28418
   BM time (ms)
                                       width
                 accurate
                            row
                                   col
                                               height
0
     236.806154
                     True
                           3799
                                  735
                                           21
                                                   36
                           1863
                                  823
                                         1306
1
     167.150259
                     True
                                                  382
2
                           3439
    1072.239637
                     True
                                  920
                                           30
                                                 1146
3
      96.963167
                     True
                           1272
                                 2013
                                          528
                                                  193
4
     299.212217
                     True
                           2019
                                 1220
                                         265
                                                  841
5
     167.706013
                     True
                           1345
                                 1086
                                         1819
                                                  494
6
     702.826023
                     True
                          3054
                                  415
                                         137
                                                 1498
7
                                   29
     92.001915
                     True
                           156
                                         2724
                                                 1030
8
     191.809654
                           1059
                                         1849
                                                  477
                     True
                                 1610
9
                     True 2371
     433.033943
                                  574
                                         1506
                                                  800
df.describe()
                               BF
                                             KMP
                                                             BM BF
               size
time (ms) \
count 1.000000e+01 1.000000e+01 1.000000e+01
                                                      10.000000
10.000000
                                                44929.800000
mean
       6.855102e+05 3.200063e+06 3.173569e+06
1255.317354
       8.538765e+05 2.153084e+06 2.137927e+06
                                                   75119.035131
std
843,679298
       7.560000e+02 2.195280e+05 1.974990e+05 1813.000000
min
103.229284
25%
       1.277345e+05 2.379618e+06 2.364698e+06
                                                   13740.250000
942.231178
       3.608785e+05 2.797487e+06 2.765379e+06
                                                   23036.000000
1119.614124
75%
       8.944328e+05 3.518302e+06 3.509202e+06 31106.750000
1291.492939
       2.805720e+06 8.572875e+06 8.495186e+06 254418.000000
max
3410.575390
       KMP time (ms) BM time (ms)
                                             row
                                                         col
width
```

```
10.000000
                          10.000000
                                       10.000000
                                                     10.00000
count
10.000000
         2517.989850
                         345.974898
                                     2037.700000
                                                    942.50000
mean
1018.500000
         1333.838877
std
                         314.558089
                                     1145.300645
                                                    576.91695
949.684538
                          92.001915
                                      156.000000
                                                     29.00000
min
          317.312241
21.000000
         1774.615228
                         167.289197
                                     1290.250000
25%
                                                    614.25000
169.000000
50%
         2242.674947
                         214.307904
                                     1941.000000
                                                    871.50000
917.000000
75%
                         399.578512
                                     2883.250000
         2890.857875
                                                   1186.50000
1740.750000
         5075.389385
                        1072.239637
                                     3799.000000
                                                   2013.00000
max
2724.000000
            height
         10.000000
count
        689.700000
mean
        455.176901
std
min
         36.000000
25%
        405.750000
        647.000000
50%
        982.750000
75%
       1498.000000
max
```

Menampilkan Sampel

```
image_samples = []

for i in range(10):
    r, c = df.at[i, 'row'], df.at[i, 'col']
    w, h = df.at[i, 'width'], df.at[i, 'height']

    image_samples.append(image[r:r+w+1, c:c+h+1])

fig, axes = plt.subplots(2, 5, figsize=(10, 8))
axes = axes.flatten()

for i, (sample, ax) in enumerate(zip(image_samples, axes)):
    ax.imshow(sample, cmap='gray')
    ax.set_title(f"Sample {i+1}")
    ax.axis('off')
```





Sample 3



Sample 4













Menampilkan Hasil Template Matching Sampel

```
result = copy.copy(image)
for i in range(10):
    r, c = df.at[i, 'row'], df.at[i, 'col']
w, h = df.at[i, 'width'], df.at[i, 'height']
    if ((r, c) != (-1, -1)):
         cv2.rectangle(result, (c, r), (c + h, r + w), (0, 255, 255),
25)
plt.figure(figsize=(10, 10))
plt.imshow(result)
plt.show()
```

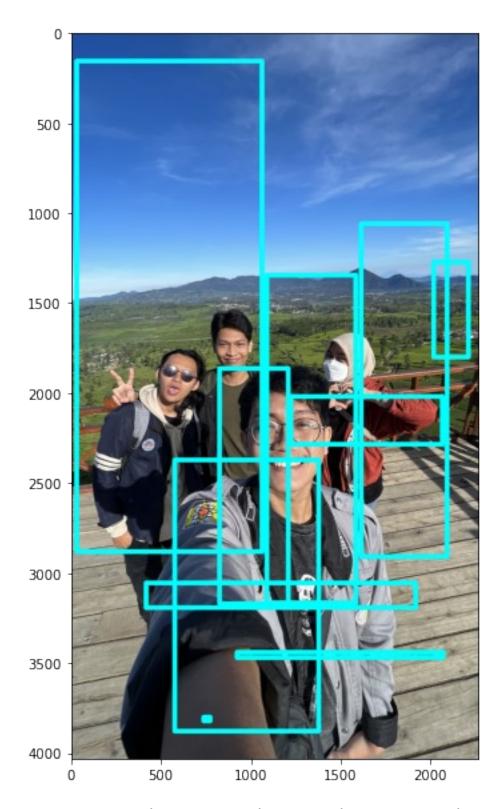


Diagram Venn Akurasi Template Matching Ketiga Algoritma

```
value_counts = df['accurate'].value_counts()
value_counts.plot(kind='pie', autopct='%1.1f%%')
```

```
plt.axis('equal')
plt.legend()
plt.title('Akurasi Template Matching')
plt.show()
```

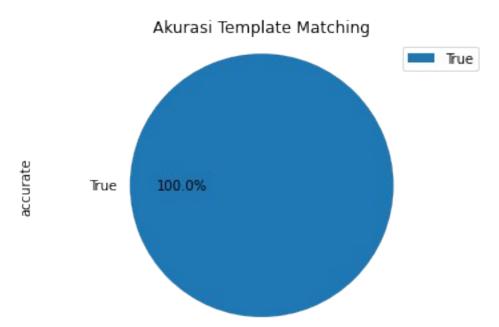


Diagram Batang Banyak Perbandingan Pixel Setiap Algoritma

```
fig, ax = plt.subplots()
# Lebar setiap bar
bar width = 0.25
# Menghitung posisi setiap bar
r1 = range(len(df))
r2 = [x + bar width for x in r1]
r3 = [x + bar width for x in r2]
# Membuat bar chart untuk setiap kategori dan data
plt.bar(r1, df['BF'], color='b', width=bar_width, edgecolor='black',
label='BF')
plt.bar(r2, df['KMP'], color='g', width=bar width, edgecolor='black',
label='KMP')
plt.bar(r3, df['BM'], color='r', width=bar width, edgecolor='black',
label='BM')
# Mengatur label sumbu x
# plt.xlabel('Uji')
plt.xticks([r + bar width for r in range(len(df))], [f'Uji {r+1}' for
r in range(len(df))])
```

```
# Mengatur label sumbu y
plt.ylabel('Banyak perbandingan')

# Menambahkan judul grafik
plt.title('Banyak Perbandingan Pixel Setiap Algoritma')

# Menambahkan legend
plt.legend()
plt.grid()

# Menampilkan grafik
plt.show()
```

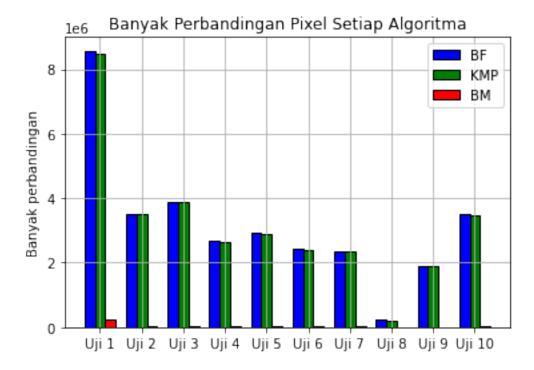


Diagram Batang Horizontal Waktu Eksekusi Setiap Algoritma

```
fig, ax = plt.subplots()

y_pos = np.arange(len(df))
bar_width = 0.2

# Membuat horizontal bar chart untuk setiap kategori dan data
plt.barh(y_pos, df['BF time (ms)'], height=bar_width, color='b',
edgecolor='black', label='BF')
plt.barh(y_pos + bar_width, df['KMP time (ms)'], height=bar_width,
color='g', edgecolor='black', label='KMP')
plt.barh(y_pos + (2 * bar_width), df['BM time (ms)'],
height=bar_width, color='r', edgecolor='black', label='BM')
```

```
# Mengatur label sumbu y
plt.yticks(y_pos + bar_width, [f'Uji {r+1}' for r in range(len(df))])
# Mengatur label sumbu y
plt.xlabel('Waktu Eksekusi (ms)')
# Menambahkan judul grafik
plt.title('Waktu Eksekusi Setiap Algoritma')
# Menambahkan legend
plt.grid()
plt.legend()
# Menampilkan grafik
plt.show()
```

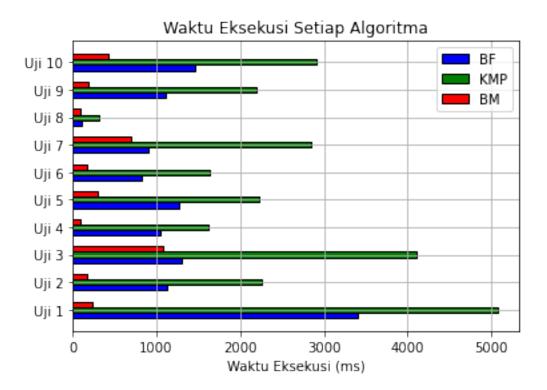


Diagram Batang Rata-Rata Perbandingan Setiap Algoritma

```
# Calculate the average values
selected_columns = ['BF', 'KMP', 'BM']
df_selected = df[selected_columns]

# Calculate the average values
avg_values = df_selected.mean()

# Create a bar plot using Seaborn
sns.barplot(x=avg_values.index, y=avg_values.values)
```

```
# Set the plot title and labels
plt.title('Average Values of BF, KMP, and BM')
plt.xlabel('Columns')
plt.ylabel('Average Value')
# Display the plot
plt.show()
```

