



Decision Tree

Ramaditia D

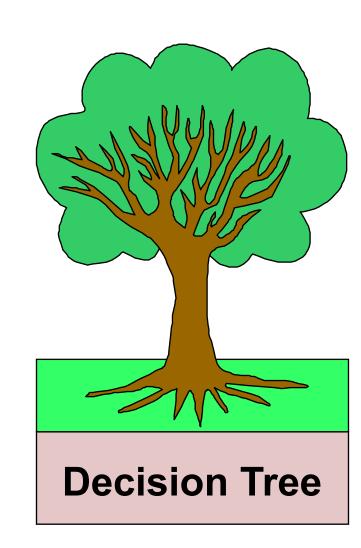
Ouline

- Pendahuluan
- Contoh Decision Tree
- Proses Membentuk Pohon
- Metode ID3

Konsep Decision Tree (Pohon Keputusan)

 Merupakan representasi visual berupa struktur pohon dari suatu pilihan aksi atau kondisi

 Merupakan mekanisme untuk menyederhanakan situasi kompleks ke dalam skenario yang lebih mudah dimengerti

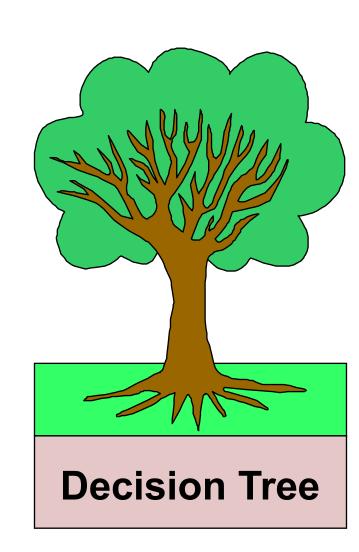


△ Kons

Konsep Decision Tree (Pohon Keputusan)

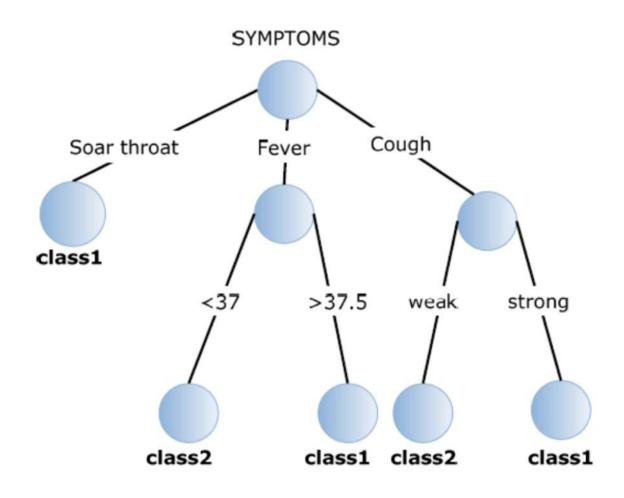
 Pembelajaran Decision Tree (DT) adalah metode memperkirakan fungsi target bernilai diskret, dimana fungsi tersebut disajikan sebagai pohon keputusan.

- RepresentasiPohonKeputusan:
 - Setiap Node internal menguji suatu atribut
 - Setiap cabang terkait dengan nilai atribut
 - Setiap Node daun menunjukkan suatu klasifikasi



Contoh Decision Tree (1)

Flue (Class1) atau Tidak (Class2)?



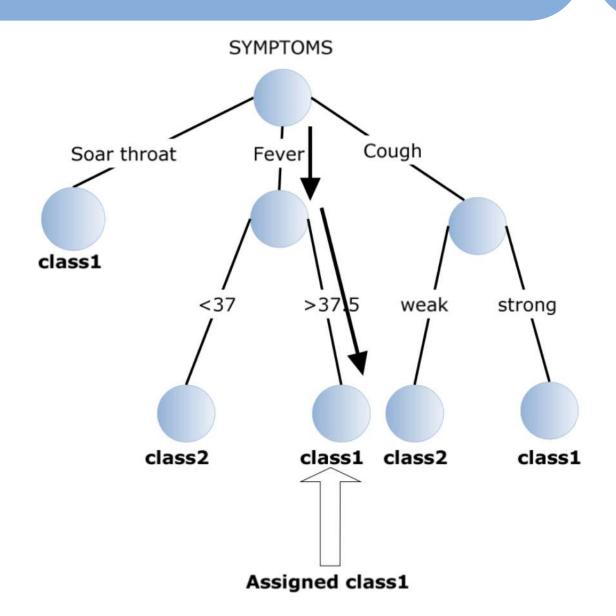


Contoh Proses Klasifikasi Decision Tree

Flue (Class1) atau Tidak (Class2)?

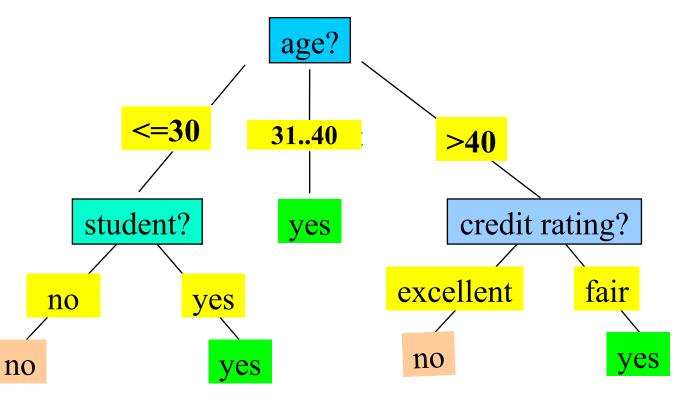
Testing Data:

Pasien Demam (>37,5)

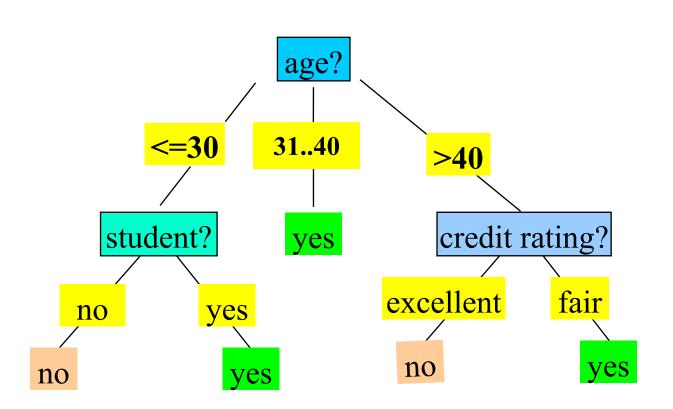


Contoh Decision Tree (2)

| age | income | student | credit_rating | buys_computer |
|------|--------|---------|---------------|---------------|
| <=30 | high | no | fair | no |
| <=30 | high | no | excellent | no |
| 3140 | high | no | fair | yes |
| >40 | medium | no | fair | yes |
| >40 | low | yes | fair | yes |
| >40 | low | yes | excellent | no |
| 3140 | low | yes | excellent | yes |
| <=30 | medium | no | fair | no |
| <=30 | low | yes | fair | yes |
| >40 | medium | yes | fair | yes |
| <=30 | medium | yes | excellent | yes |
| 3140 | medium | no | excellent | yes |
| 3140 | high | yes | fair | yes |
| >40 | medium | no | excellent | no |



Representasi Tree sebagai Rule

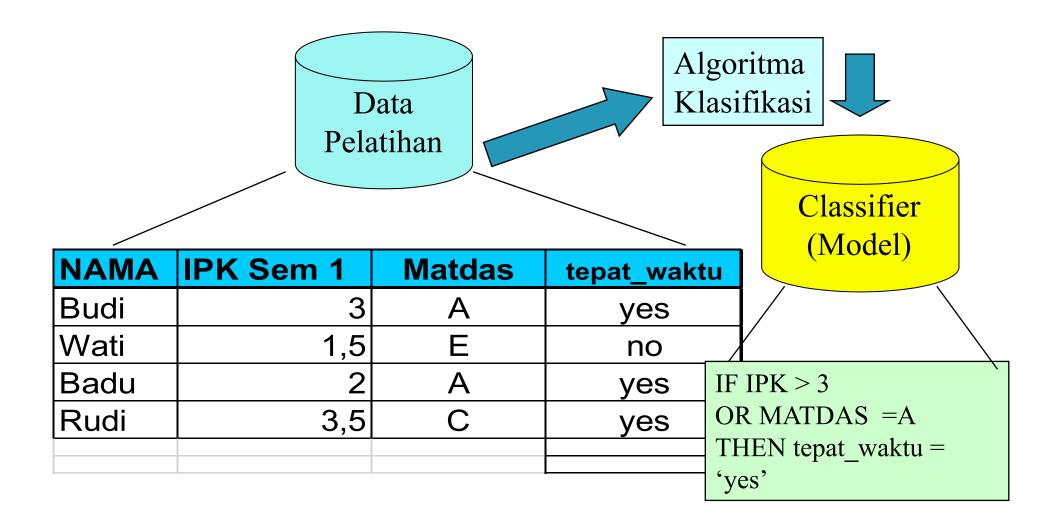


```
IF
 ((age \le 30) and
  (student))
OR
  (age=31..40)
OR
  ((age>40) and
  (credit_rating=fair))
THEN
 BELI PC=YES
```

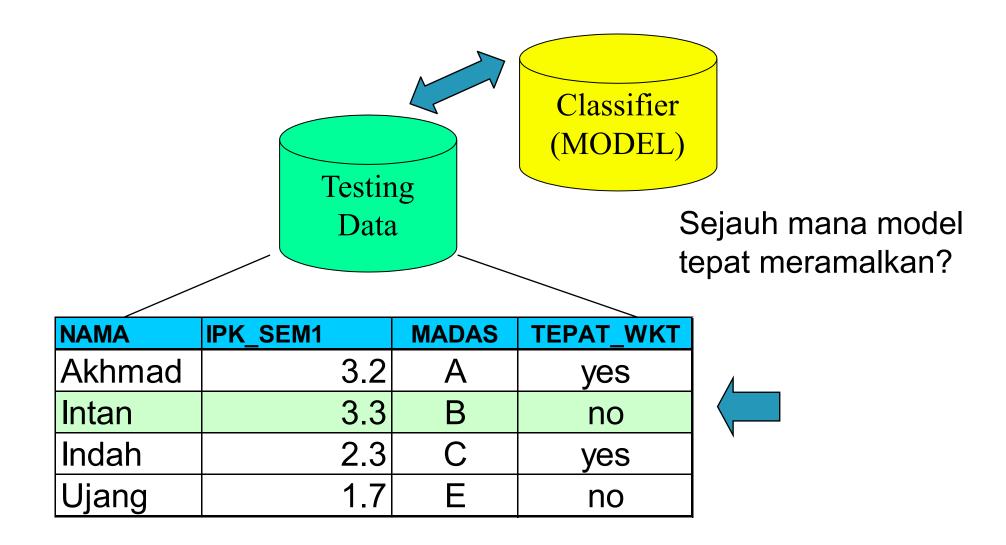
Contoh Kasus Lain

- Input: data mahasiswa
- Output: dua kelas (lulus_tepat_waktu dan lulus_terlambat)
- ✓ Jika diberikan data input mahasiswa, sistem secara otomatis menentukan mahasiswa tersebut akan lulus tepat waktu atau terlambat.

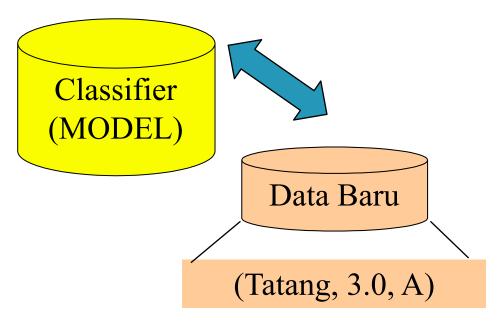
Pembuatan Model



Proses Testing Model



Proses Klasifikasi



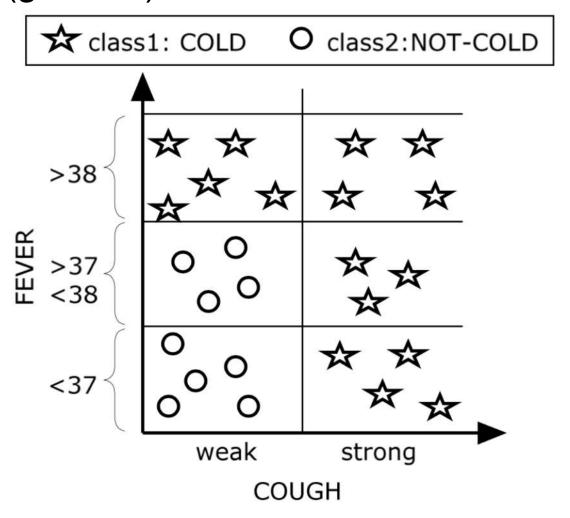
Lulus tepat waktu?



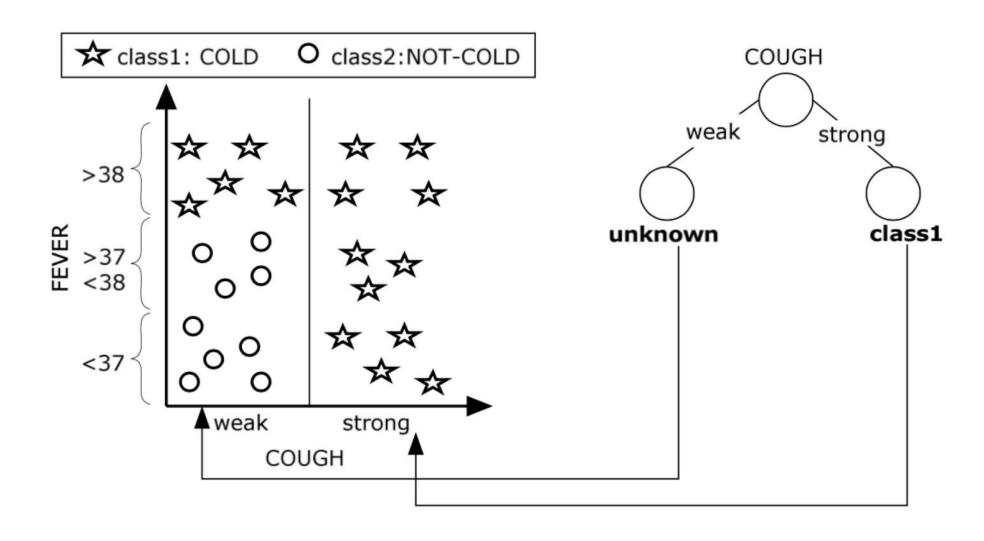


Contoh: Cold (flue) atau Tidak?

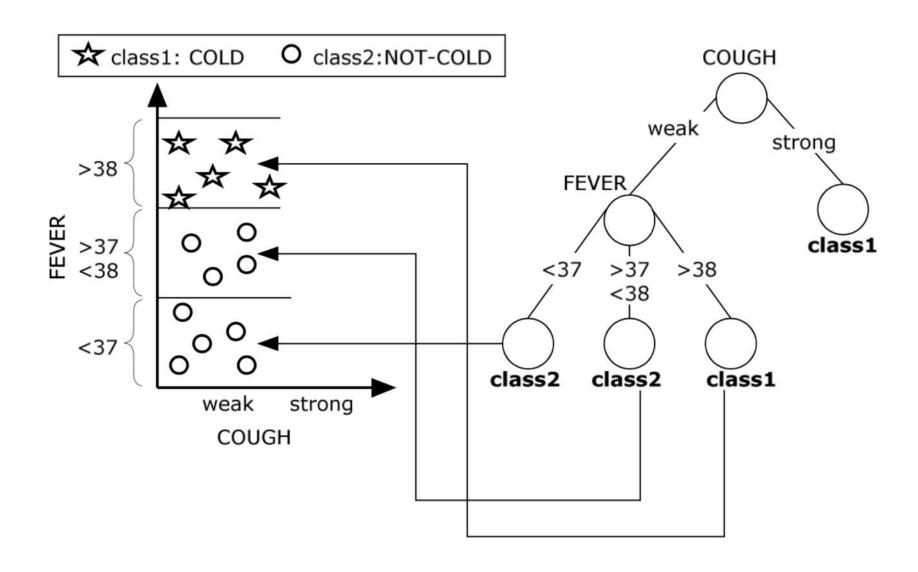
Data Training (gambar):







Bentuk Pohon Akhir

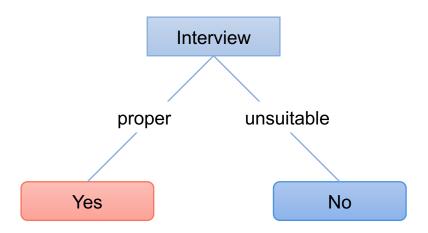


Relasi antar Atribut

Contoh: Data Rekrutmen Pegawai

| Applicants | GPA | Psychology | Interview | Accepted |
|------------|---------|------------|------------|----------|
| P1 | Good | Strong | Proper | Yes |
| P2 | Good | Moderate | Proper | Yes |
| P3 | Good | Moderate | Unsuitable | No |
| P4 | Good | Weak | Unsuitable | No |
| P5 | Average | Strong | Proper | Yes |
| P6 | Average | Moderate | Proper | Yes |
| P7 | Average | Moderate | Unsuitable | No |
| P8 | Average | Weak | Unsuitable | No |
| P9 | Poor | Strong | Proper | Yes |
| P10 | Poor | Moderate | Proper | Yes |
| P11 | Poor | Moderate | Unsuitable | No |
| P12 | Poor | Weak | Unsuitable | No |

Bentuk Decision Tree



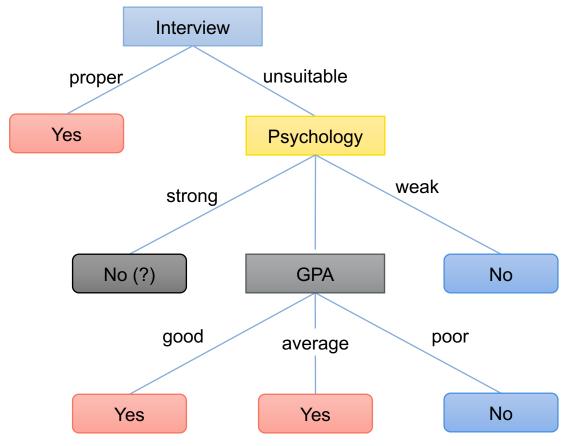
if Interview='Proper' then Accepted='Yes'

Relasi antar Atribut

Data Rekrutmen yang lain

| Applicants | GPA | Psychology | Interview | Accepted |
|------------|---------|------------|------------|----------|
| P1 | Good | Strong | Proper | Yes |
| P2 | Good | Moderate | Proper | Yes |
| P3 | Good | Moderate | Unsuitable | Yes |
| P4 | Good | Weak | Unsuitable | No |
| P5 | Average | Strong | Proper | Yes |
| P6 | Average | Moderate | Proper | Yes |
| P7 | Average | Moderate | Unsuitable | Yes |
| P8 | Average | Weak | Unsuitable | No |
| P9 | Poor | Strong | Proper | Yes |
| P10 | Poor | Moderate | Unsuitable | No |
| P11 | Poor | Weak | Proper | Yes |

Bentuk Decision Tree

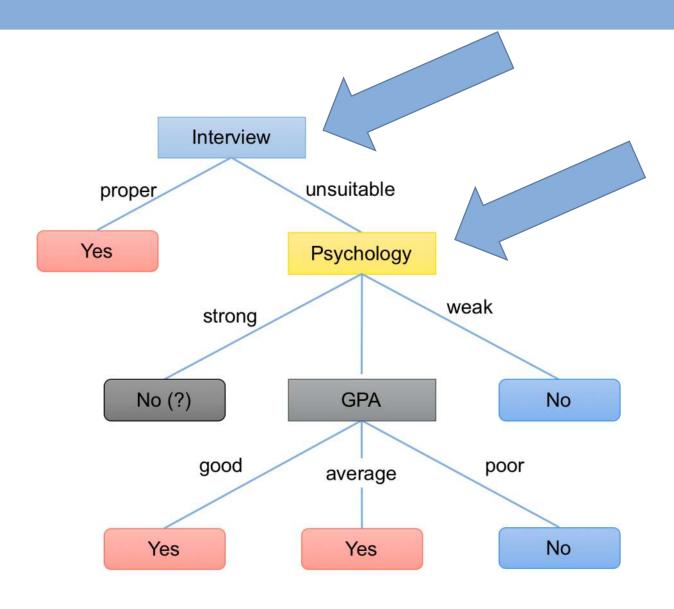


```
if (Interview='Proper') or
  ((Interview='Unsuitable')and(Psychology='Moderate')and(GPA='Good') ) or
  ((Interview='Unsuitable')and(Psychology='Moderate')and(GPA='Average') )
  then Accepted='Yes'
```

Induction of Decision Trees (ID3)

- A adalah atribut keputusan "terbaik" bagi node berikutnya
- Jadikan A sebagai atribut keputusan
- Untuk setiap nilai A, buat keturunan baru
- Urutkan sample pada node-node daun
- Jika sample telah terklasifikasi secara sempurna, STOP.
- Jika tidak, lakukan iterasi terhadap node daun baru.

Cara Pemilihan Urutan Atribut Terbaik?



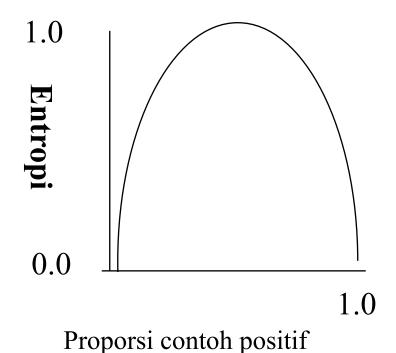
Metode Pemilihan Atribut Decision Tree

• Entropi: Ukuran kemurnian, semakin murni, semakin homogen, semakin rendah nilainya.

- Information Gain (IG): pengurangan entropi disebabkan oleh partisi berdasarkan suatu atribut.
- Semakin besar IG → atribut itu semakin membuat homogen → semakin bagus
- Ide Utama → pilih atribut dengan info gain yang paling besar

Entrophy untuk dua kelas: (+) dan (-)

Entropy(S) =
$$-p_{\oplus} \log_2 p_{\oplus} - p_{\Theta} \log_2 p_{\Theta}$$



Entropy([9+,5-] ((9 positif, 5 neg)) = $-(9/14) \log_2(9/14) - (5/14) \log_2(5/14)$ = 0.940 Entropy([9+,5-]) = 0.940 Entropy([7+,7-]) = 1 Entropy([14+,0]) = 0

Entroy([0+,14-]) = 0

Nilai Entropi

 Entropi(S) = 0, jika semua contoh pada S berada dalam kelas yang sama.

 Entroiy(S) = 1, jika jumlah contoh positif dan jumlah contoh negatif dalam S adalah sama.

 0 < Entropi(S) < 1, jika jumlah contoh positif dan negatif dalam S tidak sama.

Contoh Entropy

Contoh Data:

| Data | Class(Y/N) |
|------|------------|
| 1 | Υ |
| 2 | Υ |
| 3 | Υ |
| 4 | Υ |

$$Entropy(S) = -p_y \log_2 p_y - p_n \log_2 p_n$$
$$= -\frac{4}{4} \log_2 \left(\frac{4}{4}\right) - \frac{0}{4} \log_2 \left(\frac{0}{4}\right)$$
$$= 0$$

$$\begin{split} Entropy(S) &= -p_y \log_2 p_y - p_n \log_2 p_n \\ &= -\frac{2}{4} \log_2 \left(\frac{2}{4}\right) - \frac{2}{4} \log_2 \left(\frac{2}{4}\right) \\ &= 1 \end{split}$$

Entrophy untuk kelas > 2

$$Info(D) = -\sum_{i=1}^{m} p_i \log_2(p_i)$$

Info (D) = Entrophy suatu data 'D' untuk kelas > 2

Information Gain

$$Info_A(D) = \sum_{j=1}^{\nu} \frac{|D_j|}{|D|} \times I(D_j)$$

$$Gain(A) = Info(D) - Info_A(D)$$

Gain(A) menyatakan seberapa besar entropi berkurang akibat atribut A. Semakin besar semakin baik.

Contoh Pemilihan Urutan Atribut

- Class P: buys_computer = "yes"
- Class N: buys_computer = "no"

$$Info(D) = I(9,5) = -\frac{9}{14}\log_2(\frac{9}{14}) - \frac{5}{14}\log_2(\frac{5}{14}) = 0.940$$

| age | p _i | n _i | I(p _i , n _i) |
|------|----------------|----------------|-------------------------------------|
| <=30 | 2 | 3 | 0.971 |
| 3140 | 4 | 0 | 0 |
| >40 | 3 | 2 | 0.971 |

| age | income | student | credit_rating | buys_computer |
|------|--------|---------|---------------|---------------|
| <=30 | high | no | fair | no |
| <=30 | high | no | excellent | no |
| 3140 | high | no | fair | yes |
| >40 | medium | no | fair | yes |
| >40 | low | yes | fair | yes |
| >40 | low | yes | excellent | no |
| 3140 | low | yes | excellent | yes |
| <=30 | medium | no | fair | no |
| <=30 | low | yes | fair | yes |
| >40 | medium | yes | fair | yes |
| <=30 | medium | yes | excellent | yes |
| 3140 | medium | no | excellent | yes |
| 3140 | high | yes | fair | yes |
| >40 | medium | no | excellent | no |

$$Info_{age}(D) = \frac{5}{14}I(2,3) + \frac{4}{14}I(4,0) + \frac{5}{14}I(3,2) = 0.694$$

 $\frac{5}{14}I(2,3)$ berarti ada 5 dari 14 "age <=30" dengan 2 yes dan 3 no.

Gain (Age) = Info(D) - Info age (D) = 0.940 - 0.694 = 0.246

Gain(income) = 0.029

Gain(student) = 0.151

 $Gain(credit_rating) = 0.048$

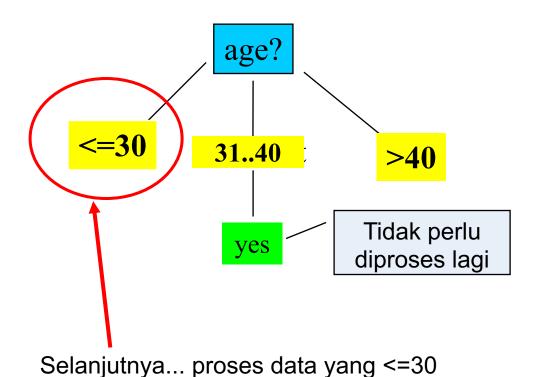
Tahap Pemilihan Atribut (lanj)

Gain (Age) = 0.246 ← yang terbesar, dipilih

Gain (income)=0.029

Gain(student)=0.151

Gain(credit_rating) = 0.048



Setelah AGE, atribut apa selanjutnya?

Diproses untuk setiap cabang selama masih ada > 1 kelas

| age | income | student | credit_rating | buys_computer |
|------|--------|---------|---------------|---------------|
| <=30 | high | no | fair | no |
| <=30 | high | no | excellent | no |
| <=30 | medium | no | fair | no |
| <=30 | low | yes | fair | yes |
| <=30 | medium | yes | excellent | yes |
| >40 | medium | no | fair | yes |
| >40 | low | yes | fair | yes |
| >40 | low | yes | excellent | no |
| >40 | medium | yes | fair | yes |
| >40 | medium | no | excellent | no |
| 3140 | high | no | fair | yes |
| 3140 | low | yes | excellent | yes |
| 3140 | medium | no | excellent | yes |
| 3140 | high | yes | fair | yes |

Tahap Pemilihan Atribut (lanj)

Selanjutnya... proses data age<=30

| age | income | student | credit_rating | buys_computer |
|------|--------|---------|---------------|---------------|
| <=30 | high | no | fair | no |
| <=30 | high | no | excellent | no |
| <=30 | medium | no | fair | no |
| <=30 | low | yes | fair | yes |
| <=30 | medium | yes | excellent | yes |

Info(D) =
$$I(2,3) = -\frac{2}{5}\log_2(\frac{2}{5}) - \frac{3}{5}\log_2(\frac{3}{5}) = 0.97$$

Gain(age) tidak perlu dihitung lagi, hitung gain(student), gain(credit_rating), gain(income)

$$Info_{student}(D) = \frac{3}{5}I(0,3) + \frac{2}{5}I(2,0) = 0$$

Gain (student) =
$$Info(D) - Info_{student}(D)$$

= $0.97 - 0 = 0.97$

Pemilihan Atribut (lanj)

| age | income | student | credit_rating | buys_computer |
|------|--------|---------|---------------|---------------|
| <=30 | high | no | fair | no |
| <=30 | high | no | excellent | no |
| <=30 | medium | no | fair | no |
| <=30 | low | yes | fair | yes |
| <=30 | medium | yes | excellent | yes |

hitung gain(credit_rating)

$$Info(D) = I(2,3) = -\frac{2}{5}\log_2(\frac{2}{5}) - \frac{3}{5}\log_2(\frac{3}{5}) = 0.97$$

$$Info_{credit_rating}(D) = \frac{3}{5}I(1,2) + \frac{2}{5}I(1,1) = 0.95$$

Gain (credit_rating) =
$$Info(D) - Info_{credit_rating}(D)$$

= $0.97 - 0.95 = 0.02$

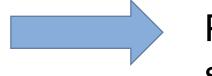
$$Info_{income}(D) = \frac{2}{5}I(0,2) + \frac{2}{5}I(1,1) + \frac{1}{5}I(1,0) = 0.4$$

Gain (income) = Info(D) - Info_{income}(D)
=
$$0.97 - 0.4 = 0.37$$

Pilihan Atribut (lanj)

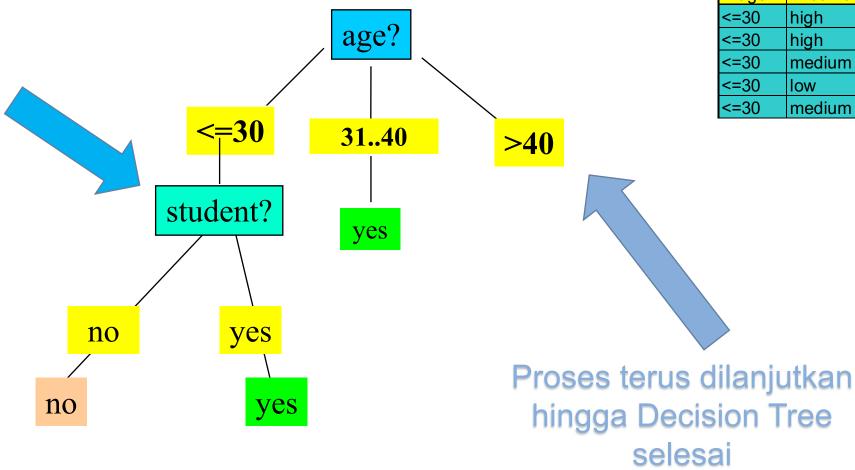
Bandingkan semua gain, ambil yang paling besar

```
Gain (student) = 0.97
Gain (credit_rating = 0.02
Gain (income) = 0.37
```



Paling besar: student

Pemilihan Atribut (lanj)



| age | income | student | credit_rating | buys_computer |
|------|--------|---------|---------------|---------------|
| <=30 | high | no fair | | no |
| <=30 | high | no | excellent | no |
| <=30 | medium | no | fair | no |
| <=30 | low | yes | fair | yes |
| <=30 | medium | yes | excellent | yes |

Contoh lainnya

Contoh: Data Rekrutmen Pegawai

| Applicants | GPA | Psychology | Interview | Accepted |
|------------|---------|------------|------------|----------|
| P1 | Good | Strong | Proper | Yes |
| P2 | Good | Moderate | Proper | Yes |
| P3 | Good | Moderate | Unsuitable | Yes |
| P4 | Good | Weak | Unsuitable | No |
| P5 | Average | Strong | Proper | Yes |
| P6 | Average | Moderate | Proper | Yes |
| P7 | Average | Moderate | Unsuitable | Yes |
| P8 | Average | Weak | Unsuitable | No |
| P9 | Poor | Strong | Proper | Yes |
| P10 | Poor | Moderate | Unsuitable | No |
| P11 | Poor | Weak | Proper | Yes |

1 Hitung Entropy Kelas

| Accepted |
|----------|
| Yes |
| Yes |
| Yes |
| No |
| Yes |
| Yes |
| Yes |
| No |
| Yes |
| No |
| Yes |
| |

•
$$Entropy(A) = -p_y \log_2 p_y - p_n \log_2 p_n$$

•
$$|A| = 11$$
, $|A_y| = 8$, $|A_n| = 3$

•
$$p_y = \frac{8}{11}$$
, $p_n = \frac{3}{11}$

•
$$Entropy(A) = -p_y \log_2 p_y - p_n \log_2 p_n$$

• $|A| = 11$, $|A_y| = 8$, $|A_n| = 3$
• $p_y = \frac{8}{11}$, $p_n = \frac{3}{11}$
• $Entropy(A) = -\frac{8}{11} \log_2 \left(\frac{8}{11}\right) - \frac{3}{11} \log_2 \left(\frac{3}{11}\right)$

$$= 0.8454$$

Information Gain - GPA

| GPA | Accepted |
|---------|----------|
| Good | Yes |
| Good | Yes |
| Good | Yes |
| Good | No |
| Average | Yes |
| Average | Yes |
| Average | Yes |
| Average | No |
| Poor | Yes |
| Poor | No |
| Poor | Yes |
| - | |

$$Gain(Accepted, GPA)$$

$$= Ent(A) - \frac{|A_g|}{|A|} Ent(A_g) - \frac{|A_a|}{|A|} Ent(A_a) - \frac{|A_p|}{|A|} Ent(A_p)$$

$$|A_g| = [3y, 1n] = 4,$$

$$|A_a| = [3y, 1n] = 4,$$

$$|A_p| = [2y, 1n] = 3$$

$$Ent(A_g) = -p_{g,y} \log_2 p_{g,y} - p_{g,n} \log_2 p_{g,n}$$

$$Ent(A_a) = -p_{a,y} \log_2 p_{a,y} - p_{a,n} \log_2 p_{a,n}$$

$$Ent(A_p) = -p_{p,y} \log_2 p_{p,y} - p_{p,n} \log_2 p_{p,n}$$

Information Gain - GPA

| GPA | Accepted |
|---------|----------|
| Good | Yes |
| Good | Yes |
| Good | Yes |
| Good | No |
| Average | Yes |
| Average | Yes |
| Average | Yes |
| Average | No |
| Poor | Yes |
| Poor | No |
| Poor | Yes |
| - | |

$$|A_g| = [3y, 1n] = 4, |A_a| = [3y, 1n] = 4, |A_p| = [2y, 1n] = 3$$

$$Ent(A_g) = -p_{g,y} \log_2 p_{g,y} - p_{g,n} \log_2 p_{g,n}$$

$$= -\frac{3}{4} \log_2 \left(\frac{3}{4}\right) - \frac{1}{4} \log_2 \left(\frac{1}{4}\right) = 0.8113$$

$$Ent(A_a) = -p_{a,y} \log_2 p_{a,y} - p_{a,n} \log_2 p_{a,n}$$

$$= -\frac{3}{4} \log_2 \left(\frac{3}{4}\right) - \frac{1}{4} \log_2 \left(\frac{1}{4}\right) = 0.8113$$

$$Ent(A_p) = -p_{p,y} \log_2 p_{p,y} - p_{p,n} \log_2 p_{p,n}$$

$$= -\frac{2}{3} \log_2 \left(\frac{2}{3}\right) - \frac{1}{3} \log_2 \left(\frac{1}{3}\right) = 0.9183$$

Information Gain - GPA

| GPA | Accepted | | |
|---------|----------|--|--|
| Good | Yes | | |
| Good | Yes | | |
| Good | Yes | | |
| Good | No | | |
| Average | Yes | | |
| Average | Yes | | |
| Average | Yes | | |
| Average | No | | |
| Poor | Yes | | |
| Poor | No | | |
| Poor | Yes | | |
| | | | |

$$|A_g| = 4$$
, $|A_a| = 4$, $|A_p| = 3$
 $Ent(A_g) = 0.8113$
 $Ent(A_a) = 0.8113$
 $Ent(A_p) = 0.9183$

Gain(Accepted, GPA)
$$= Ent(A) - \frac{|A_g|}{|A|} Ent(A_g) - \frac{|A_a|}{|A|} Ent(A_a) - \frac{|A_p|}{|A|} Ent(A_p)$$

$$= 0.8454 - \frac{4}{11} 0.8113 - \frac{4}{11} 0.8113 - \frac{3}{11} 0.9183$$

$$= 0.0049$$

Information Gain - Psychology

| Psy | Accepted | | |
|--------|----------|--|--|
| Strong | Yes | | |
| Mod | Yes | | |
| Mod | Yes | | |
| Weak | No | | |
| Strong | Yes | | |
| Mod | Yes | | |
| Mod | Yes | | |
| Weak | No | | |
| Strong | Yes | | |
| Mod | No | | |
| Weak | Yes | | |

$$Gain(Accepted, Psychology) = Ent(A) - \frac{|A_s|}{|A|} Ent(A_s) - \frac{|A_m|}{|A|} Ent(A_m) - \frac{|A_w|}{|A|} Ent(A_w)$$

$$|A_s| = [3y, 0n] = 3,$$

$$|A_m| = [4y, 1n] = 5,$$

$$|A_w| = [1y, 2n] = 3$$

$$Ent(A_s) = -\frac{3}{3} \log_2\left(\frac{3}{3}\right) - \frac{0}{3} \log_2\left(\frac{0}{3}\right) = 0$$

$$Ent(A_m) = -\frac{4}{5} \log_2\left(\frac{4}{5}\right) - \frac{1}{5} \log_2\left(\frac{1}{5}\right) = 0.7219$$

$$Ent(A_w) = -\frac{1}{3} \log_2\left(\frac{1}{3}\right) - \frac{2}{3} \log_2\left(\frac{2}{3}\right) = 0.9183$$

Information Gain - Psychology

= 0.2669

| Psy | Accepted | | | |
|--------|----------|--|--|--|
| Strong | Yes | | | |
| Mod | Yes | | | |
| Mod | Yes | | | |
| Weak | No | | | |
| Strong | Yes | | | |
| Mod | Yes | | | |
| Mod | Yes | | | |
| Weak | No | | | |
| Strong | Yes | | | |
| Mod | No | | | |
| Weak | Yes | | | |

$$|A_s| = 3, |A_m| = 5, |A_w| = 3$$

$$Ent(A_s) = 0$$

$$Ent(A_m) = 0.7219$$

$$Ent(A_w) = 0.9183$$

$$Gain(Accepted, Psychology)$$

$$= Ent(A) - \frac{|A_s|}{|A|} Ent(A_s) - \frac{|A_m|}{|A|} Ent(A_m) - \frac{|A_w|}{|A|} Ent(A_w)$$

$$= 0.8454 - \frac{3}{11}0 - \frac{5}{11}0.7219 - \frac{3}{11}0.9183$$

Information Gain - Interview

| Intv | Accepted | | | |
|--------|----------|--|--|--|
| Proper | Yes | | | |
| Proper | Yes | | | |
| Unsuit | Yes | | | |
| Unsuit | No | | | |
| Proper | Yes | | | |
| Proper | Yes | | | |
| Unsuit | Yes | | | |
| Unsuit | No | | | |
| Proper | Yes | | | |
| Unsuit | No | | | |
| Proper | Yes | | | |

Gain(Accepted, Interview)

$$= Ent(A) - \frac{|A_p|}{|A|} Ent(A_p) - \frac{|A_u|}{|A|} Ent(A_u)$$

$$|A_p| = [6y, 0n] = 6,$$

 $|A_u| = [2y, 3n] = 5,$

$$|A_u| = [2y, 3n] = 5,$$

$$Ent(A_p) = -\frac{6}{6}\log_2\left(\frac{6}{6}\right) - \frac{0}{6}\log_2\left(\frac{0}{6}\right) = 0$$

$$Ent(A_p) = -\frac{6}{6}\log_2\left(\frac{6}{6}\right) - \frac{0}{6}\log_2\left(\frac{0}{6}\right) = 0$$

$$Ent(A_u) = -\frac{2}{5}\log_2\left(\frac{2}{5}\right) - \frac{3}{5}\log_2\left(\frac{3}{5}\right) = 0.9710$$

Information Gain - Interview

| Intv | Accepted | |
|--------|----------|--|
| Proper | Yes | |
| Proper | Yes | |
| Unsuit | Yes | |
| Unsuit | No | |
| Proper | Yes | |
| Proper | Yes | |
| Unsuit | Yes | |
| Unsuit | No | |
| Proper | Yes | |
| Unsuit | No | |
| Proper | Yes | |
| | | |

$$|A_p| = [6y, 0n] = 6, |A_u| = [2y, 3n] = 5,$$

$$Ent(A_p)=0$$

$$Ent(A_u) = 0.9710$$

Gain(Accepted, Interview)

$$= Ent(A) - \frac{|A_p|}{|A|} Ent(A_p) - \frac{|A_u|}{|A|} Ent(A_u)$$
$$= 0.8454 - \frac{6}{11}0 - \frac{5}{11}0.9710$$

$$= 0.8454 - \frac{6}{11}0 - \frac{5}{11}0.9710$$

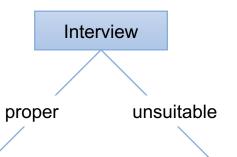
$$= 0.4040$$

Hasil Perhitungan Atribut untuk pemilihan Root

- Gain(Accepted, GPA) = 0.0049
- Gain(Accepted, Psychology) = 0.2669
- Gain(Accepted, Interview) = 0.4040

Pilih Interview sebagai Root

Bentuk Awal Decision Tree



| GPA | Psychology | Accepted | |
|---------|---------------|----------|--|
| Good | Strong | Yes | |
| Good | Good Moderate | | |
| Average | Strong | Yes | |
| Average | Moderate | Yes | |
| Poor | Strong | Yes | |
| Poor | Weak | Yes | |

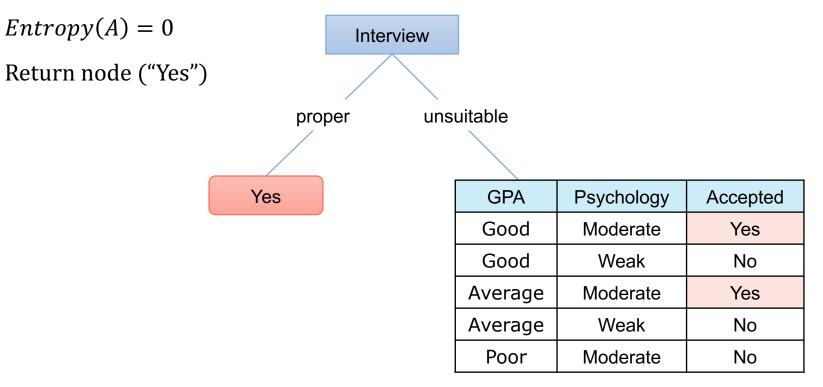
| GPA | Psychology | Accepted |
|---------|------------|----------|
| Good | Moderate | Yes |
| Good | Weak | No |
| Average | Moderate | Yes |
| Average | Weak | No |
| Poor | Moderate | No |



Pemilihan Atribut setelah Atribut Interview - Proper

Proper Interview

| GPA | Psychology | Accepted |
|---------|------------|----------|
| Good | Strong | Yes |
| Good | Moderate | Yes |
| Average | Strong | Yes |
| Average | Moderate | Yes |
| Poor | Strong | Yes |
| Poor | Weak | Yes |



Unsuitable Interview

| GPA | Psychology | Accepted |
|---------|------------|----------|
| Good | Moderate | Yes |
| Good | Weak | No |
| Average | Moderate | Yes |
| Average | Weak | No |
| Poor | Moderate | No |

$$Ent(A) = -\frac{2}{5}\log_2\left(\frac{2}{5}\right) - \frac{3}{5}\log_2\left(\frac{3}{5}\right) = 0.9709$$

$$Ent(A_m) = -\frac{2}{3}\log_2\left(\frac{2}{3}\right) - \frac{1}{3}\log_2\left(\frac{1}{3}\right) = 0.9182$$

$$Ent(A_w) = 0$$

$$Ent(A_w) = 0$$

$$Ent(A_g) = 1 \qquad Ent(A_a) = 1 \qquad Ent(A_p) = 0$$

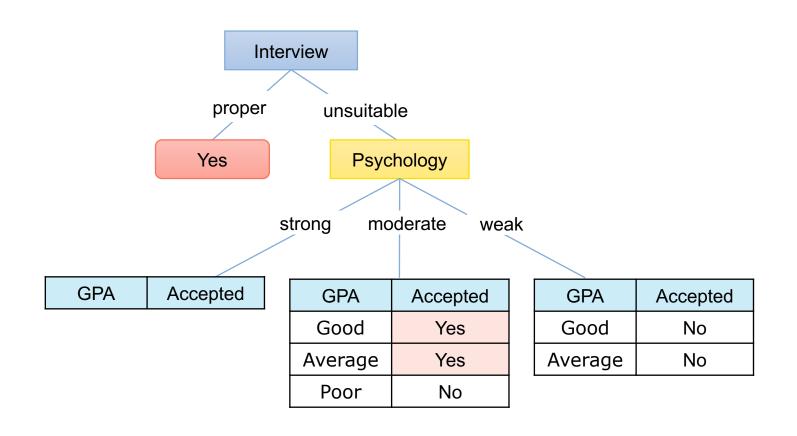
$$Gain(Accepted, GPA) = Ent(A) - \frac{|A_g|}{|A|} Ent(A_g) - \frac{|A_a|}{|A|} Ent(A_a) - \frac{|A_p|}{|A|} Ent(A_p)$$

$$= 0.9709 - \frac{2}{5}1 - \frac{2}{5}1 - \frac{1}{5}0 = 0.1709$$

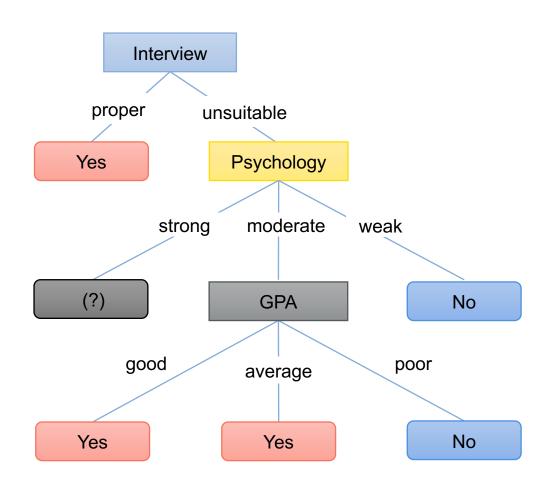
$$Gain(Accepted, Psi) = Ent(A) - \frac{|A_m|}{|A|} Ent(A_m) - \frac{|A_w|}{|A|} Ent(A_w)$$

$$= 0.9709 - \frac{3}{5}0.9182 - \frac{2}{5}0 = \mathbf{0}.2822$$

Hasil Decision Tree setelah Atribut Psycology Dipilih



Hasil Akhir Decision Tree



Mengapa Decision Tree?

- Mudah diimplementasikan
- Hipotesis yang dihasilkan mudah dipahami
- Efisien
- Daerah pengambilan dapat diubah menjadi simple dan spesifik.
- Dapat Eliminasi perhitungan-perhitungan yang tidak diperlukan

Decision Tree Cocok untuk Masalah:

- Data dalam bentuk atribut-nilai. Kondisi ideal adalah jika isi nilai jumlahnya sedikit. Misalnya: "panas", "sedang", "dingin".
- Output diskrit.

1 Latihan

Buat Decision Tree untuk menentukan kelas dari suatu buah

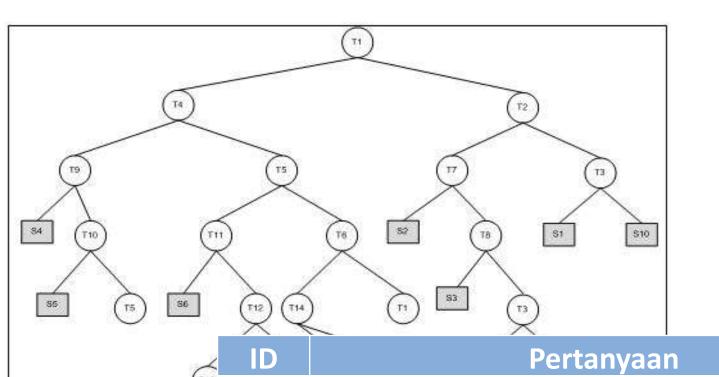
| No | Kulit Buah | Warna | Ukuran | Bau | Kelas |
|----|------------|--------|--------|-------|-----------|
| 1 | Kasar | Coklat | Besar | keras | Aman |
| 2 | Kasar | Hijau | Besar | keras | Aman |
| 3 | Halus | Merah | Besar | Lunak | Berbahaya |
| 4 | Kasar | Hijau | Besar | Lunak | Aman |
| 5 | Kasar | Merah | Kecil | Keras | Aman |
| 6 | Halus | Merah | Kecil | Keras | Aman |
| 7 | Halus | Coklat | Kecil | Keras | Aman |
| 8 | Kasar | Hijau | Kecil | Lunak | Berbahaya |
| 9 | Halus | Hijau | Kecil | Keras | Berbahaya |
| 10 | Kasar | Merah | Besar | Keras | Aman |
| 11 | Halus | Coklat | Besar | Lunak | Aman |
| 12 | Halus | Hijau | Kecil | Keras | Berbahaya |
| 13 | Kasar | Merah | Kecil | Lunak | Aman |
| 14 | Halus | Merah | Besar | Keras | Berbahaya |
| 15 | Halus | Merah | Kecil | Keras | Aman |
| 16 | Kasar | Hijau | Kecil | Keras | Berbahaya |

Contoh Konversi Decision Tree



Contoh Wujud Decision Tree

Apakah kamu suka hal yang berhubungan dengan perancangan?



T1

T2

T3

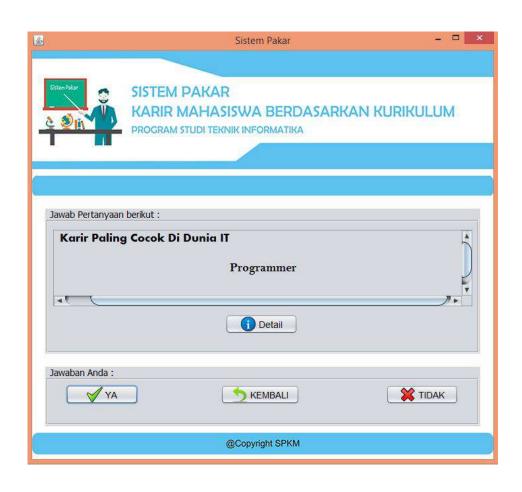
T4

T5

. . .

| | Pertanyaan | Ya | Tidak |
|-------------------------------------------------|--------------|--------------------------------------------------------|----------------|
| (11) | T1 | T4 | T2 |
| | T2 | T7 | T3 |
| Ť2 | Т3 | S1 | S10 |
| | T4 | Т9 | T5 |
| Te SS TS Pertanyaan | Jawab Pertan | SISTEM PAKAR KARIR MAHASISWA PROGRAM STUDI TEKNIK INFO | 100 |
| Apakah kamu suka hal yang berhubungan dengan so | ftware? | • | Detail |
| Apakah kamu suka hal yang berhubungan dengan ha | ardware? | 12 | |
| Apakah kamu suka hal yang berhubungan dengan ko | mputer? | YA SKE | EMBALI X TIDAK |
| Apakah kamu suka hal yang berhubungan dengan co | ding? | @Copyrl | ght SPKM |

Contoh Tabel Kesimpulan



| ID | Solusi | Detail |
|------------|----------------|-------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|
| S 5 | Program mer | Programmer adalah orang yang membuat suatu aplikasi untuk kclient atau server baik untuk perusahaan, istansi atupun perorangan. Tugas: Membuat perogram baik aplikasi maupun sistem oprasi dengan menggunakan bahasa pemrograman yang ada. Kualifikasi: Menguasai logika dan algoritma pemrograman seperti HTML, Ajax, CSS, JavaScript, C++, VB, PHP, Java, Ruby dll. Memahami SQL dan menguasai Bahasa Inggris IT. |

