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<html><head></head><body><pre style="word-wrap: break-word; white-space:
pre-wrap; "> This data was extracted from the census bureau database
found at
 http://www.census.gov/ftp/pub/DES/www/welcome.html
 Donor: Ronny Kohavi and Barry Becker,
        Data Mining and Visualization
        Silicon Graphics.
        e-mail: ronnyk@sgi.com for questions.
 Split into train-test using MLC++ GenCVFiles (2/3, 1/3 random).
 48842 instances, mix of continuous and discrete (train=32561,
test=16281)
 45222 if instances with unknown values are removed (train=30162,
test=15060)
 Duplicate or conflicting instances: 6
 Class probabilities for adult.all file
| Probability for the label '>50K' : 23.93% / 24.78% (without
unknowns)
| Probability for the label '<=50K': 76.07% / 75.22% (without
unknowns)
Extraction was done by Barry Becker from the 1994 Census database.
                                                                    Α
set of
   reasonably clean records was extracted using the following
conditions:
    ((AAGE>16) && (AGI>100) &&
(AFNLWGT>1)&& (HRSWK>0))
 Prediction task is to determine whether a person makes over 50K
 a year.
 First cited in:
  @inproceedings{kohavi-nbtree,
    author={Ron Kohavi},
    title={Scaling Up the Accuracy of Naive-Bayes Classifiers: a
           Decision-Tree Hybrid},
    booktitle={Proceedings of the Second International Conference on
               Knowledge Discovery and Data Mining },
    year = 1996,
    pages={to appear}}
 Error Accuracy reported as follows, after removal of unknowns from
    train/test sets):
    C4.5
               : 84.46+-0.30
    Naive-Bayes: 83.88+-0.30
    NBTree : 85.90+-0.28
 Following algorithms were later run with the following error rates,
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| Following algorithms were later run with the following error rates, | all after removal of unknowns and using the original train/test split.

| All these numbers are straight runs using MLC++ with default values.

		Algorithm	Error				
	1	C4.5	15.54				
İ	2	C4.5-auto	14.46				
İ	3	C4.5 rules	14.94				
İ	4	Voted ID3 (0.6)	15.64				
İ	5	Voted ID3 (0.8)	16.47				
İ	6	Т2	16.84				
ĺ	7	1R	19.54				
ĺ	8	NBTree	14.10				
ĺ	9	CN2	16.00				
	10	HOODG	14.82				
	11	FSS Naive Bayes	14.05				
	12	IDTM (Decision table)	14.46				
	13	Naive-Bayes	16.12				
	14	Nearest-neighbor (1)	21.42				
	15	Nearest-neighbor (3)	20.35				
	16	OC1	15.04				
	17	Pebls	Crashed.	Unknown	why	(bounds	WERE

Conversion of original data as follows:

- 1. Discretized agrossincome into two ranges with threshold 50,000.
- 2. Convert U.S. to US to avoid periods.
- 3. Convert Unknown to "?"
- 4. Run MLC++ GenCVFiles to generate data, test.

Description of fnlwgt (final weight)

The weights on the CPS files are controlled to independent estimates of the

civilian noninstitutional population of the US. These are prepared monthly

| for us by Population Division here at the Census Bureau. We use 3 sets of

controls.

state.

increased)

These are:

- 1. A single cell estimate of the population 16+ for each
  - 2. Controls for Hispanic Origin by age and sex.
  - 3. Controls by Race, age and sex.

We use all three sets of controls in our weighting program and "rake" through

| them 6 times so that by the end we come back to all the controls we used.

The term estimate refers to population totals derived from CPS by

creating
| "weighted tallies" of any specified socio-economic characteristics of
the
| population.

People with similar demographic characteristics should have similar weights. There is one important caveat to remember about this statement. That is that since the CPS sample is actually a collection of 51 state samples, each with its own probability of selection, the statement only applies within state.

>50K, <=50K.

age: continuous.

workclass: Private, Self-emp-not-inc, Self-emp-inc, Federal-gov, Local-gov, State-gov, Without-pay, Never-worked.

fnlwgt: continuous.

education: Bachelors, Some-college, 11th, HS-grad, Prof-school, Assocacdm, Assoc-voc, 9th, 7th-8th, 12th, Masters, 1st-4th, 10th, Doctorate, 5th-6th, Preschool.

education-num: continuous.

marital-status: Married-civ-spouse, Divorced, Never-married, Separated, Widowed, Married-spouse-absent, Married-AF-spouse.

occupation: Tech-support, Craft-repair, Other-service, Sales, Exec-managerial, Prof-specialty, Handlers-cleaners, Machine-op-inspct, Adm-clerical, Farming-fishing, Transport-moving, Priv-house-serv, Protective-serv, Armed-Forces.

relationship: Wife, Own-child, Husband, Not-in-family, Other-relative, Unmarried.

race: White, Asian-Pac-Islander, Amer-Indian-Eskimo, Other, Black.

sex: Female, Male.

capital-gain: continuous.

capital-loss: continuous.

hours-per-week: continuous.

native-country: United-States, Cambodia, England, Puerto-Rico, Canada, Germany, Outlying-US(Guam-USVI-etc), India, Japan, Greece, South, China, Cuba, Iran, Honduras, Philippines, Italy, Poland, Jamaica, Vietnam, Mexico, Portugal, Ireland, France, Dominican-Republic, Laos, Ecuador, Taiwan, Haiti, Columbia, Hungary, Guatemala, Nicaragua, Scotland, Thailand, Yugoslavia, El-Salvador, Trinadad& Tobago, Peru, Hong, Holand-Netherlands.

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