

# PROBABILISTIC REASONING

## Labsheet-1

### 1. What is a Bayesian classifier?

A Bayesian classifier is a type of statistical classifier based on Bayes' Theorem, which provides a way to update the probability of a hypothesis as more evidence becomes available. It's widely used in machine learning for tasks such as classification, especially with text data. The most well-known example is the Naive Bayes classifier.

### 2. Why we use a Naive Bayes Classifier? Why it is called Naive?

A Naive Bayes classifier is used because it is simple, efficient, and effective for many classification tasks, especially with high-dimensional data like text. Despite its simplicity, it can perform well in various applications such as email spam detection, sentiment analysis, and medical diagnosis.

The classifier is termed "naive" because it makes a naive assumption: it assumes that all features are conditionally independent given the class label.

### 3. What are the possible advantages in choosing Naive Bayes Classifier?

- Simple and Fast: Easy to implement and quick to train, even on large datasets.
- Effective with Small Data: Works well with limited training data.
- Performs Well with High-Dimensional Data: Great for text classification and other tasks with many features.
- Handles Multi-Class Problems: Naturally supports multi-class classification.
- Interpretable: Provides clear probability estimates for each class.
- Robust to Irrelevant Features: Can perform well even with noisy or irrelevant features.

4. Prepare a classification model using Naive Bayes for the given data.

5. Consider class Fish as Y1, class Animal as Y2, and class Bird as Y3 Compute  $P(Y1)$ ,  $P(Y2)$ ,  $P(Y3)$ .

6. Consider the test sample  $X=(\text{Slow, Rarely, No})$ . Predict the class label for the test sample, using the Naive Bayes classifier. (Hint: Find  $P(Y1/X)$ ,  $P(Y2/X)$ , and  $P(Y3/X)$ ).

[\* Executing Qns 4 to 6 in a single code \*]

#### CODE

% Data preparation

```
Swim = {'Fast', 'Fast', 'Slow', 'Fast', 'No', 'No', 'No', 'Slow', 'Slow', 'Slow', 'No', 'Fast'};
```

```
Fly = {'No', 'No', 'No', 'No', 'Short', 'Short', 'Rarely', 'No', 'No', 'No', 'Long', 'No'};
```

```
Crawl = {'No', 'Yes', 'No', 'No', 'No', 'No', 'No', 'Yes', 'No', 'Yes', 'No', 'No'};
```

```
ClassLabel = {'Fish', 'Animal', 'Animal', 'Animal', 'Bird', 'Bird', 'Animal', 'Animal', 'Fish', 'Fish', 'Bird', 'Bird'};
```

% Convert categorical data into numeric

```
categories_swim = unique(Swim);
```

```
categories_fly = unique(Fly);
```

```
categories_crawl = unique(Crawl);
```

```
categories_class = unique(ClassLabel);
```

```
Swim_int = arrayfun(@(x) find(strcmp(x, categories_swim)), Swim); Fly_int
```

```
= arrayfun(@(x) find(strcmp(x, categories_fly)), Fly); Crawl_int =
```

```
arrayfun(@(x) find(strcmp(x, categories_crawl)), Crawl);
```

```
ClassLabel_int = arrayfun(@(x) find(strcmp(x, categories_class)), ClassLabel);
```

```
% Combine data into a matrix
```

```
data = [Swim_int, Fly_int, Crawl_int, ClassLabel_int];
```

```
% Step 2: Compute Prior Probabilities  $P(Y_1)$ ,  $P(Y_2)$ ,  $P(Y_3)$ 
```

```
num_classes = length(categories_class);

priors = histc(data(:, end), 1:num_classes) / size(data, 1);
disp('Prior Probabilities:');

for c = 1:num_classes
    fprintf('P(Y%d) = %.2f\n', c, priors(c));end

% Step 3: Calculate likelihoods P(X|Y) for each feature and class
num_features = size(data, 2) - 1;

likelihoods = cell(num_classes, num_features);for
c = 1:num_classes

    class_data = data(data(:, end) == c, 1:num_features);for
        f = 1:num_features

            feature_vals = unique(data(:, f));
            likelihoods{c, f} = histc(class_data(:, f), feature_vals) / size(class_data, 1);end

end

% Test sample X = (Slow, Rarely, No)
test_sample = [find(strcmp('Slow', categories_swim)), ...
               find(strcmp('Rarely', categories_fly)), ...
               find(strcmp('No', categories_crawl))];

% Step 4: Compute posterior probabilities  $P(Y|X) = P(X|Y) * P(Y) / P(X)$ 
posteriors = zeros(num_classes, 1);
```

```
for c = 1:num_classes
    posterior = priors(c); for
        f = 1:num_features
            posterior = posterior * likelihoods{c, f}(test_sample(f));end
```

```
    posteriors(c) = posterior;
end

% Normalize posteriors to get  $P(Y1|X)$ ,  $P(Y2|X)$ ,  $P(Y3|X)$ 
posteriors = posteriors / sum(posteriors);

% Display results disp('Posterior
Probabilities:');for c =
1:num_classes

    fprintf('P(Y%d|X) = %.2f\n', c, posteriors(c));end

% Predict the class label

[~, predicted_class_idx] = max(posteriors); predicted_class
= categories_class{predicted_class_idx};

fprintf('Predicted class for the test sample is: %s\n', predicted_class);
```

```
octave:12> source("mat_lab1.m")
Prior Probabilities:
P(Y1) = 0.42
P(Y2) = 0.33
P(Y3) = 0.25
Posterior Probabilities:
P(Y1|X) = 1.00
P(Y2|X) = 0.00
P(Y3|X) = 0.00
Predicted class for the test sample is: Animal
```





LAB-2

```
# Install the required library
```

```
!pip install pgmpy
```



```
any.whl.metadata (9.1 kB)
```

```
rx in /usr/local/lib/python3.10/dist-packages (from pgmpy) (3.4.2)in
/usr/local/lib/python3.10/dist-packages (from pgmpy) (1.26.4)
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/usr/local/lib/python3.10/dist-packages (from google-generativeai->pgmpy) (2.19.2)
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g-extensions in /usr/local/lib/python3.10/dist-packages (from google-generativeai->pgmpy) (4.12.2)
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/usr/local/lib/python3.10/dist-packages (from pandas->pgmpy) (2.8.2)
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api-python-client->google-generativeai->pgmpy) (4.1.1) pSafe>=2.0 in /usr/local/lib/python3.10/dist-packages (from jinja2->torch->pgmpy) (3.0.2)
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# Install pgmpy and supporting libraries

!pip install pgmpy networkx matplotlib



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```
from pgmpy.models import BayesianNetwork
```

```
# Step 1: Define the Bayesian Network
```

```
structuremodel = BayesianNetwork([
```

```
    ('Diff', 'Grade'), # Difficulty influences Grade
```

```
    ('Intel', 'Grade'), # Intelligence influences Grade
```

```
    ('Grade', 'Letter'), # Grade influences Letter
```

```
    ('Intel', 'SAT'))]) # Intelligence influences SAT
```

```
# Step 2: Define the CPDs (Conditional Probability
```

```
Distributions)from pgmpy.factors.discrete import TabularCPD
```

```
# CPD for Difficulty
```

```
cpd_diff = TabularCPD(variable='Diff', variable_card=2,
                      values=[[0.6], [0.4]],
                      state_names={'Diff': ['Easy', 'Hard']})

# CPD for Intelligence
cpd_intel = TabularCPD(variable='Intel', variable_card=2,
                      values=[[0.7], [0.3]],
                      state_names={'Intel': ['Dumb', 'Intelligent']})

# CPD for Grade
cpd_grade = TabularCPD(variable='Grade', variable_card=3,
                      values=[
                          [0.3, 0.05, 0.9, 0.5], # Grade A (G=0)
                          [0.4, 0.25, 0.08, 0.3], # Grade B (G=1)
                          [0.3, 0.7, 0.02, 0.2] # Grade C (G=2)
                      ],
                      evidence=['Intel', 'Diff'], evidence_card=[2, 2],
                      state_names={'Grade': ['A', 'B', 'C'], 'Intel': ['Dumb', 'Intelligent'], 'Diff': ['Easy', 'Hard']})

# CPD for SAT
cpd_sat = TabularCPD(variable='SAT', variable_card=2,
                      values=[
                          [0.95, 0.2], # SAT Bad (S=0)
                          [0.05, 0.8] # SAT Good (S=1)
```

```

    ],

    evidence=['Intel'], evidence_card=[2],

    state_names={'SAT': ['Bad', 'Good'], 'Intel': ['Dumb', 'Intelligent']})

# CPD for Letter

cpd_letter = TabularCPD(variable='Letter', variable_card=2,

                        values=[

                            [0.1, 0.4, 0.99], # Letter Bad (L=0)

                            [0.9, 0.6, 0.01]   # Letter Good (L=1)

                        ],

                        evidence=['Grade'], evidence_card=[3],

                        state_names={'Letter': ['Bad', 'Good'], 'Grade': ['A', 'B', 'C']})

# Add CPDs to the
model

model.add_cpds(cpd_diff, cpd_intel, cpd_grade, cpd_sat, cpd_letter)

# Validate the model

assert model.check_model()

print("CPDs:")
print(cpd_diff)
print(cpd_intel
)
print(cpd_grad
e)
print(cpd_sat)
print(cpd_letter)

```

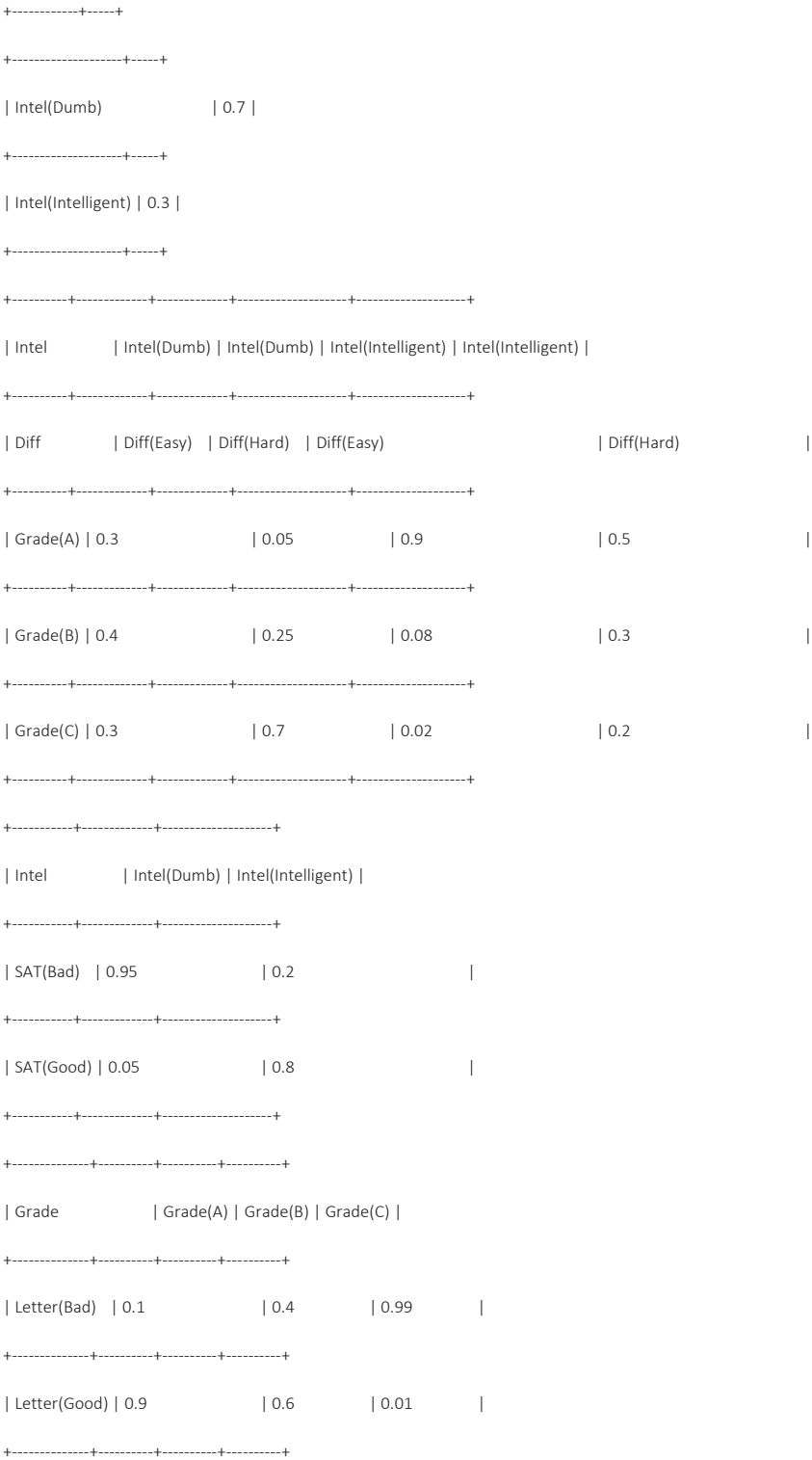


CPDs:

```

+-----+-----+
| Diff(Easy) | 0.6 |
+-----+-----+
| Diff(Hard) | 0.4 |

```



```
# Step 3: Local independencies and active trail nodes# Local
Independencies


print("Local Independencies:")

print(model.local_independencies('Diff'))
print(model.local_independencies('Intel'))
print(model.local_independencies('Grade'))
print(model.local_independencies('SAT'))
```

```
print(model.local_independencies('Letter'))
```

```
# Active trail nodes for 'Diff'
```

```
print("\nActive trail nodes for 'Diff':")
print(model.active_trail_nodes('Diff'))
```



Local Independencies: (Diff  $\perp$  SAT, Intel) (Intel  $\perp$  Diff)

(Grade  $\perp$  SAT | Diff, Intel)

(SAT  $\perp$  Diff, Grade, Letter | Intel) (Letter  $\perp$  SAT, Diff, Intel | Grade)

```
Active trail nodes for 'Diff':

{'Diff': {'Diff', 'Grade', 'Letter'}}
```



```
# Step 4: Visualize the Bayesian Network
import networkx as nx

import matplotlib.pyplot as plt

# Manually convert the BayesianNetwork into a networkx graph
graph = nx.DiGraph()
for edge in model.edges():
    graph.add_edge(edge[0], edge[1])

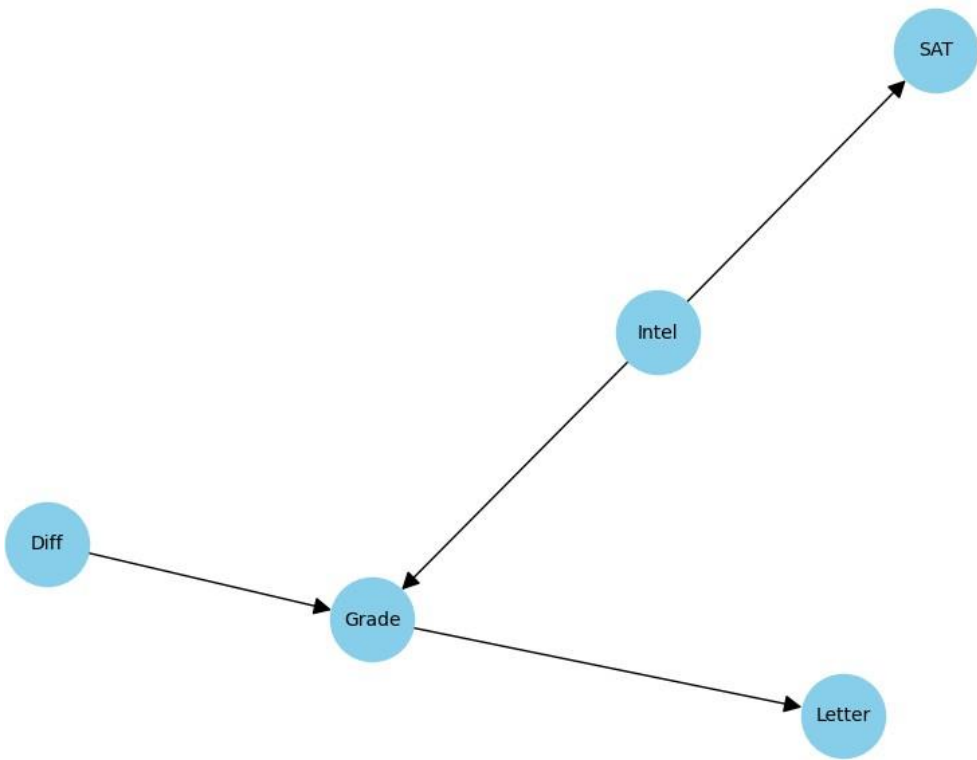
# Visualize the graph using networkx
plt.figure(figsize=(8, 6))

nx.draw(graph, with_labels=True, node_color="skyblue", font_size=10, node_size=2000, edge_color="black", arrowsize=20)
plt.title("Bayesian Network Visualization")

plt.show()
```



Bayesian Network Visualization



```
import networkx as nx

import matplotlib.pyplot as plt

# Create a directed graph using networkx
graph = nx.DiGraph()

# Add nodes and edges from the Bayesian Network
graph.add_nodes_from(model.nodes())

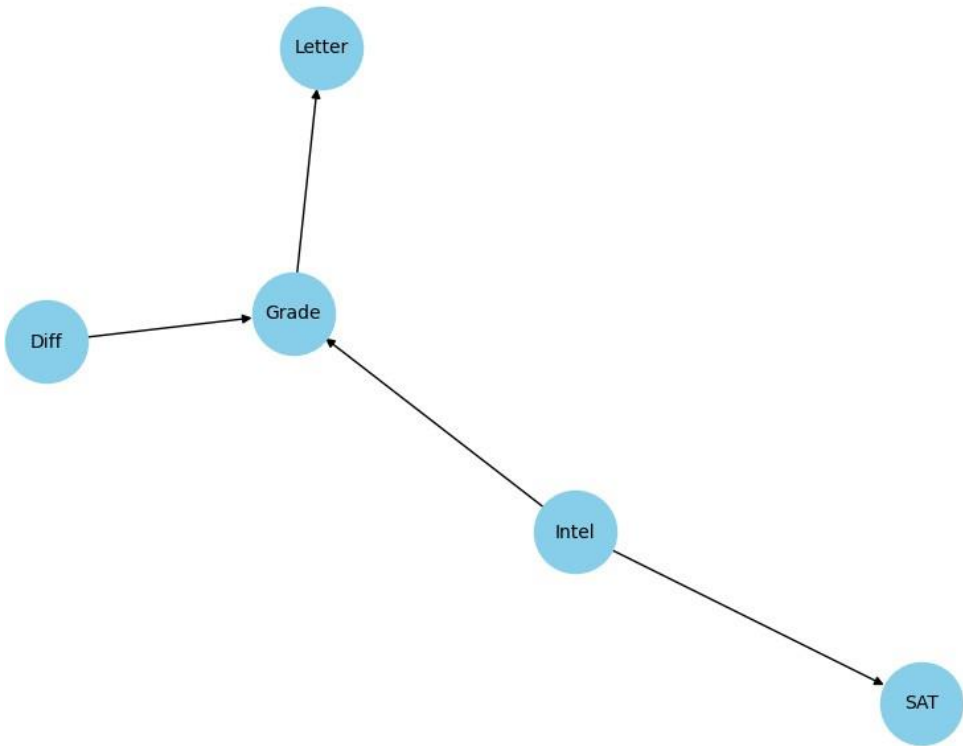
graph.add_edges_from(model.edges())
```

```
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Visualization")

plt.show()
```

Bayesian Network Visualization





## LAB-3

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Requirement already satisfied: annotated-types<=0.6.0 in /usr/local/lib/python3.10/dist-packages (from pydantic->google-generatiRequirement already satisfied: pydantic-core==2.27.1 in /usr/local/lib/python3.10/dist-packages (from pydantic->google-generativ Requirement already satisfied: grpcio<2.0dev,>=1.33.2 in /usr/local/lib/python3.10/dist-packages (from google-api-core[grpc]!=2. Requirement already satisfied: grpcio-status<2.0.dev0,>=1.33.2 in /usr/local/lib/python3.10/dist-packages (from google-api-core[ Requirement already satisfied: pyasn1<0.7.0,>=0.4.6 in /usr/local/lib/python3.10/dist-packages (from pyasn1-modules=>0.2.1->goog Requirement already satisfied: charset-normalizer<4,>=2 in /usr/local/lib/python3.10/dist-packages (from requests<3.0.0.dev0,>=2 Requirement already satisfied: idna<4,>=2.5 in /usr/local/lib/python3.10/dist-packages (from requests<3.0.0.dev0,>=2.18.0->googl Requirement already satisfied: urllib3<3,>=1.21.1 in /usr/local/lib/python3.10/dist-packages (from requests<3.0.0.dev0,>=2.18.0- Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.10/dist-packages (from requests<3.0.0.dev0,>=2.18.0- Downloading pgmpy-0.1.26-py3-none-any.whl (2.0 MB)

2.0/2.0 MB 17.6 MB/s eta 0:00:00Installing collected packages: pgmpy

# Import necessary modules

from pgmpy.models import MarkovNetwork

from pgmpy.factors.discrete import DiscreteFactorfrom  
pgmpy.inference import VariableElimination

# Step 1: Define the Markov Network# Create  
an empty Markov Network

model = MarkovNetwork()

# Add edges as per the network structure from the table

model.add\_edges\_from([('A', 'B'), ('B', 'C'), ('C', 'D'), ('D', 'A')])

# Step 2: Add the Factors# Factor  
 $\phi(A, B)$

phi\_AB = DiscreteFactor(  
variables=['A', 'B'],

```

cardinality=[2, 2], # Binary variables
values=[

    [30, 5], # A=0, B=0 | A=0, B=1 [1, 10] #
    A=1, B=0 | A=1, B=1

]

)

# Factor  $\phi(B, C)$ 

phi_BC = DiscreteFactor(
    variables=['B', 'C'],
    cardinality=[2, 2], values=[

        [100, 1], # B=0, C=0 | B=0, C=1 [1, 100]
        # B=1, C=0 | B=1, C=1

    ]

)

# Factor  $\phi(C, D)$ 

phi_CD = DiscreteFactor(
    variables=['C', 'D'],
    cardinality=[2, 2], values=[

        [1, 100], # C=0, D=0 | C=0, D=1 [100, 1]
        # C=1, D=0 | C=1, D=1

    ]

)

# Factor  $\phi(D, A)$ 

phi_DA = DiscreteFactor(
    variables=['D', 'A'],
    cardinality=[2, 2], values=[

        [100, 1], # D=0, A=0 | D=0, A=1 [1, 100]
        # D=1, A=0 | D=1, A=1

    ]

)

# Add these factors to the model

model.add_factors(phi_AB, phi_BC, phi_CD, phi_DA)

# Step 3: Perform inference

inference = VariableElimination(model)

# Example query: MAP estimation for variable 'C' given evidence A=0 and B=1
result = inference.map_query(variables=['C'], evidence={'A': 0, 'B': 1})

print("\nMAP Query result (C given A=0, B=1):")
print(result)

# Example query: Probability of 'C' given evidence

prob_result = inference.query(variables=['C'], evidence={'A': 0, 'B': 1})
print("\nProbability distribution of 'C' given A=0, B=1:")

```



print(prob\_result)



Eliminating: D: 100%

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MAP Query result (C given A=0, B=1):

{'C': 1}

Probability distribution of 'C' given A=0, B=1:

+-----+-----+	
C	phi(C)
+=====+	
C(0)	1000.0000
+-----+-----+	
C(1)	5000500.0000
+-----+-----+	



