

Bayesian and Neural Networks for Motion Picture Recommendation

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4 5 6 3 * % ' 7 # 8 ' % 9 & * * % ' # : & ' / 3 % # 4 ; # 4 . 6 < ' & =
> < " # ? @ # @ BBC

Abstract

1 2 3 * # (2 & * 3 * # < DD . 3 & * # E < F 2 3 , & # . & < ' , 3 , / # (& F 2 , 3 G) & * # (% # < # 5 < (< * & (# % 9 # E % 6 3 & * # 5 & * F ' 3 ! & 5 #
! " # F % . . < ! % ' < (3 6 & # 1 % F 3 < . 1 # < , 5 # F % , (& , (# < ((' 3 !)) (& * # 3 , # % ' 5 & ' # (% # F ' & < (& # < # E 3 J & 5 # ' & F % E E & , 5 & ' #
* * * (& E # 9 % ' # E % 6 3 & * ; # + < " & * 3 < , # , & (K % ' L * A # (K % # 6 & * * 3 % , * # % 9 # , &) ' < . # , & (K % ' L * A # 5 & F 3 * 3 % , # (' & & * A #
< , 5 # * 3 E D . & # ') . & # F . < * * 3 9 3 & * * < ' & # F % E D < ' & 5 ; # M (# 3 * # 5 & (& ' E 3 , & 5 # (2 < (# + < " & * 3 < , # < , 5 # , &) ' < .
, & (K % ' L * #) (D & ' 9 % ' E # (2 & # ' & E < 3 , 3 , / # (& F 2 , 3 G) & * ; # + % (2 # (& F 2 , 3 G) & * # < F 2 3 & 6 & 5 # ' & E < ' L < ! . " #
2 3 / 2 # (% D N ? B # D ' & F 3 * 3 % , A # < , # 3 E D % ' (< , (# E & (' 3 F # % 9 # ' & F % E E & , 5 < (3 % , # G) < . 3 (' ; # 4 , # < ((& E D (# (% #
F % , (' < * (# ' & F % E E & , 5 < (3 % , # G) < . 3 (" # 9 % ' # F % , (& , (N % , . " # < , 5 # F % . . < ! % ' < (3 6 & N % , . " # 5 < (< * & (* # < * #
F % E D < ' & 5 # K 3 (2 # < # 5 < (< * & (# 5 & * F ' 3 ! & 5 # ! " # ! % (2 # F % , (& , (# < , 5 # F % . . < ! % ' < (3 6 & # < ((' 3 !) (& * # " 3 & . 5 * #
3 , F % , F .) * 3 6 & # ' & *) . (* ; # M , *) 9 9 3 F 3 & , (# F % , (& , (# 3 , 9 % ' E < (3 % , # 3 , # (2 & # F) ' ' & , (# 5 < (< * & (* # E < " # ! & # (2 & #
' & < * % , # 9 % ' # (2 3 * ;

Introduction

Definition: A *recommender system* is a system that takes data about a user's past history in a certain industry, such as products they have purchased, movies they have seen, or websites they have visited, and predicts what the user may prefer to purchase or see in the future.

: %E&#&F%E&E, 5&'#*''*(E*#<'&#collaborative*''*(E*#3, #K23F2#%(2&'#)*&'*0#D<*(# 23*(%3&#<'&#)*, #'3, /#(%#3, L#(2&#D<'(3F).<'#)*&'#(%#<#/'%)D#%9#)*&'*#K3(2#*3E3.<'# 3, (&'&*(%#&'D)'F2<*&#;##123*#/'%)D#K3..#(2&, #3, 9.)&, F&#K2<#(2&#&F%E&E, 5&'#*''*(E*#K3..# %)D)#!<*%, #K2<#(2&#/'%)D#.3L&#%#*#53*.3L&#*#P(2&'#*''*(E*#<'&#content-based *''*(E*#3, #K23F2#5&<3.*#%9#(2&#D'5)F(##E%63&#K&! *3(&A#%#%(2&'#3(&E#<'&#F%ED<'5# </<3, *(#(2%*&#%9#*3E3.<'#3(&E*#(2<#(2&#)*&'#2<#!&&, #3, #F%, (<F(##K3(2;##12&*&#*3E3.<'#3(&E*# <'&.D#(%#<)/&#K2&(2&'#(23*#)*&'#K3..#3L&#%#*#53*.3L&#(2(&E;##: %E&#&F%E&E, 5&'# *''*(E*#)*&#<#E3J)'&#%9#F%..<'%<(36&#<, 5#F%, (&, (N!<*#<DD'<F2&#; \$&F%E&E, 5&'#*''*(E*#2<6&#!&F%E&#<#D%D).<'#*)!0&F(#<'(3, /#3, #2&#.<(&#?RRB0* H\$&*, 3FL#S#T<'3<, A#?RRUI;##><, ''#%, .3, &#&'&<3.&'*A#*)F2#<#4E<=#, ;F%EA#)*&# '&F%E&E, 5&'#*''*(E*#3, #%'5&'#(%#&'&F%E&E, 5#, &K#D'5)F(*(%F)*(%E&'#3, #%'5&'#(%#(''# (%#E<J3E3=#&'D'93(*;#V%#&J<ED.&A#*)DD%*&#<#F)*(%E&'#!)''*#<#*F3&, F&N93F(3%, #!%#L;##12&# , &J(##3E&#(2&#F)*(%E&'#63*3*(2&#*3(&A#3(##E3/2(##&'&F%E&E, 5#!%#L*#!''#(2&#<E&#<)(2%'A#%# E<'#*)//&*(%#(2&'#*F3&, F&N93F(3%, #!%#L*#(2<#(2&'#F)*(%E&'*#2<6&#!%) /2(##, #<553(3%, #(%# (2&#%, &#(2&#F)*(%E&'#D)'F2<*#(2&#.<*(##3E&#%#*2?*3(&5;##+'#2<63, /#(23*#('D&#%9# *''*(E*#3, #D.<F&A#&'&<3.&'*#<'&#<!.&#(%#<..%K#F)*(%E&'#(%#E%&#&<*3.'#93, 5#D'5)F*(2<# (2&'#E<'#!, (&'&'&5#3, #2<, #2&3'##&<3.#*(%&'&#F%), (&'D<'(*; 12&'&#<'&#*#E&#&F%E&E, 5&'#*''*(E*#K23F2#<..%K#)*&'*(%#3*(&, #(%#5399&'&, (# ('D&'#%9#E)*3F#<, 5#&#(##D'9&'&, F&'#!<*%, #2&#*%, /*#(2&'#3*(&, #(%#12&*&#('D&'#%9# '&F%E&E, 5&'#*''*(E*#A#*)F2#<#W<2%#;F%E0*#X4YZ-OF<*(#&<53%A#<..%K#<#)*&'#(%#/#(## 3, *(, (#)D5<(&'#3, #&'&F%E&E, 5<(3%, *#5)&#(%#!&3, /#<!.&#(%#<(&#*%, /*#<#(2&'#<'&#D.<'3, /;## 4*#*)F2A#)*&'#<'&#<!.&#(%#3*(&, #(%#E%&#*%, /*#(2<#(2&'#F%).5#D%**3!.'#3L&#%, #<# D&'*%, <.3=&5#<53%#*(<(3%, ; ><, ''#K&!*3(&#<.*%#)*&#&F%E&E, 5&'#*''*(E*#%#D&'*%, <.3=&#(2&3'3, (&'9<F&#K3(2#)&'*#3, #%'5&'#(%#E<3, (<3, #63*3*(%#(2&3'##3(&##V%#&J<ED.&A#<(#<#, &K*#*3(&A#39#<#D&'*%, # &, (&'*#(2&3'##=3D#F%5&A#(2&#*3(&E<'#F%, (<3, #%F<#, &K*#A#K<(&'A#<, 5#*D%'(*#(2<#(2&#)*&'# E<'#D'9&'#(%#&'&<5#%6&'# , <(3%, <.#%#/'%!<#F%, (&, (;##+'#<..%K3, /#(23*#F)*(%E3=<(3%, #)*&'*# E<'#!&#(&'&#&, 0%'#(2&#*3(&#<, 5#E%&#&<*3.'#93, 5#<'(3F.&*#(2<#(2&'#<', (&'&'&5#3, ; >'#3, (&, (3%, #3*#%#93, 5#<#!&#&'&<.'%3(2E#K23F2#F%E!3, &*#F%, (&, (N!<*#<(<

Machine Learning Concepts

> <F23, &#.&<' , 3, /#3*#<#<D35. ''#/'%K3, /#93&.5#K3(23, #F%ED)(&'#*F3&, F&;

Definition: “A computer program is said to **learn** from experience E with respect to some class of tasks T and performance measure P , if its performance at tasks in T , as measured by P , improves with experience E ” (Mitchell, 1997).

Definition: A **dataset** is a group of data. It is basically an n -by- m matrix with n rows and m columns. The rows are called **instances**. Instances are basically different occurrences of a situation. The columns are called **attributes**. Attributes are certain details that were recorded during every instance.

V3/) ' &#?#3*#K2<(#<#/&, &'3F#5<(<*&(#E<''#.%L#.3L&;

Figure 17#Sample Dataset

Student	Graduation Year	Major	GPA
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Decision Trees

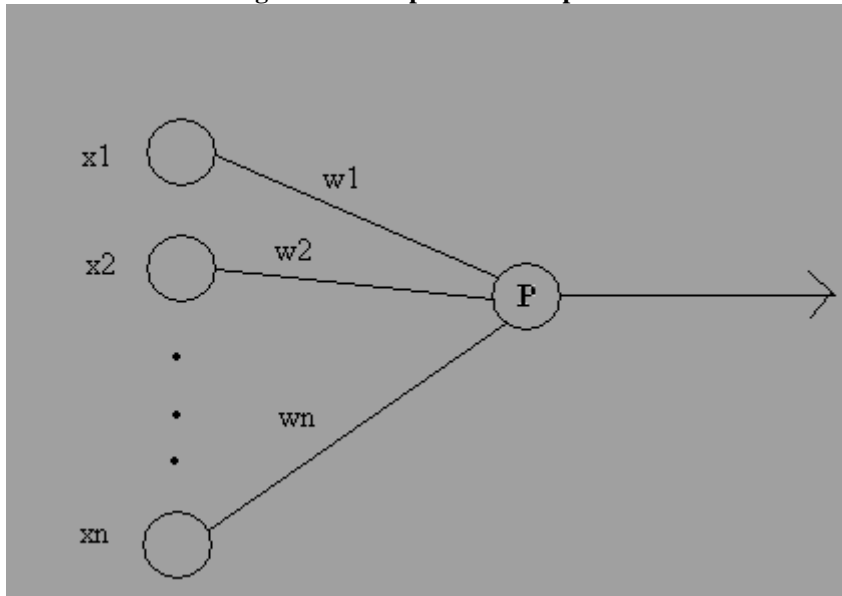
4, %(2&'#F.<***393&'#K23F2#!)3.5*#)D%, #').&#F.<***393&'*#3*#%, &#K3(2#5&F3*3%, #'&'&*&#;##
` &F3*3%, #'&'&*&#<'3F<..'#(<L&#<#.%%L#<(#E).(3D.&#<((('3!))&'&#<, 5#! '&<L#(2&E#3, (%#*#D<'<(&#
D<'3, /*#D' &53F(3, /*#5399&'&, (#6<.)&'9%#(2&#(<' /&#<((('3!))&'&#<'&5#%, #2&#D<'3, /*#9%#(2&#
6<'3<!.&'&#;#Z&K#! '<, F2&'&#<'&#E<5	%'# , &K#D<'3, /*#<, 5#<'&#<'&#*).(#<, #&, (3'&#('&'*#E<5&#&#
K23F2#(2&#F.<***393&'#! '<, F23, /*#<(#&<F2#D%3, (#5&D&, 53, /*# , #K2<#(2<.)	%#<#D<'(3F).<'#
<((('3!))&'*#;#><, ''#<./%'3(2E*#2<6&#!&&, #E<5&#&#)*3, /*#5&F3*3%, #'&'&*&#(2&#E*#(#!<'3F#<E%, /*#
(2&E#!&3, /*(2&#M` ^#<./%'3(2E#1a)3, <, #?RR^!;#O%K&6&#A#5&F3*3%, #'&'&#*#3, 3(3<..'#2<5#<#
D'#! .&E#3, #6&'93(3, /*5<(<#3, #2<(#'<, 5%E#%FF)''&, F&'&#K%).5#F'&(&# , &K#! '<, F2&'&#(%#
'&'&#<, 5#K%).5#(2&'&9%'&#.&<5#(%#<#E%'&#F%ED.&J#('&'&#&# , &#(2<(#F%).5#D' &53F(#9<.*&#
) (F%E*#5)&#(%#*#E, *(<, F&'&#K2*#&#(<' /&#<((('3!))&'<.)&#K%).5#!ᔗ&'&, (#2<, #
) (2&', *(<, F&'&#K3(2*#3E3.<'#6<.)&'9%#3*#<((('3!))&'*#<'&#*).(# , &K#<./%'3(2E*#K&'&#&
E<5&#&#)*3, /*D') , 3, /*#8') , 3, /*#K%).5#&'&E#6&#! '<, F2&'	#(2&#('&''#3(3ED' %6&5#(2&#
)6&'&#<.#<FF)'<F' #9#(2&#('&'&#bJ<ED.&'#9#5&F3*3%, #'&'&#<./%'3(2E*#3, F.)5&#- \; [1a)3, <, #
?RR^!#K23F2#!)3.(#)D%, #M` ^#! ''#<553, /*D') , 3, /*#<'&#K&.#<'&#<.%K3, /* ,)E&'3F#<((('3!))&#
6<.)&'&#;

Neural Networks

4'(393F3<.#, &)'<.#, &(K%'L*#<'&#<, %(2&'#('D	#E<F23, &#.&<' , 3, /*#&F2, 3G)&#M(#K<*#
3, *D3'&5#! ''#(2&# , &)'% , *#3, *35&#<#2)E<, #!'<3, #K23F2#F% , , &F(#%#&<F2#%(2&'&#<, 5#/#&, &'<(&
) (D)(*#!<'&5#% , *#(3E).3#9' %E#%(2&'&# , &)'% , *;##M, #<'(393F3<.#, &)'<.#, &(K%'L*#D&'F&D(' , *#
<'&#)*&5#3, *(&<5#9# , &)'% , *;##4#D&'F&D(' , #(<L&'#3, #E<, ''#3, D)(*#<, 5#<'/ , *#*%E&#
F% , *(<, (#F<..&5#(%#&<F2#% , &#F<..&5#<#K&3/2(##12&#K&3/2(#&D'&'&#&, (*#(2ED%'< , F	#(2&#
3, D)(#%#(2&#D&'F&D(' , ;##M, #2&#E<F23, &#.&<' , 3, /*#F<'&#(2, D)(*#<'&#(2ᔗ&'&, (#6<.)&'9%#
(2&#<((('3!))&'	#<, #3, *(<, F&#D&'F&D(' , #2&, #<55*#<..#9#(2, D)(*#E).(3D.3&5#! ''#
(2&3'&K&3/2(*;##M(#2&, #<DD.3&#<#(2'&'&2%.5#(%#(23*#6<.)&#M9#3(3*#<!'%6&#BA#(2&#D&'F&D(' , #K3.#
) (D)(#<#?;##M9#(2<.)*# , %(#<!'%6&#BA#(2&, #2&#D&'F&D(' , #K3.#%) (D)(#<#V?;##V%'&#J<ED.&#
V3/)'&#@#3*#<#53<'<E#9#<#D&'F&D(' , ;##

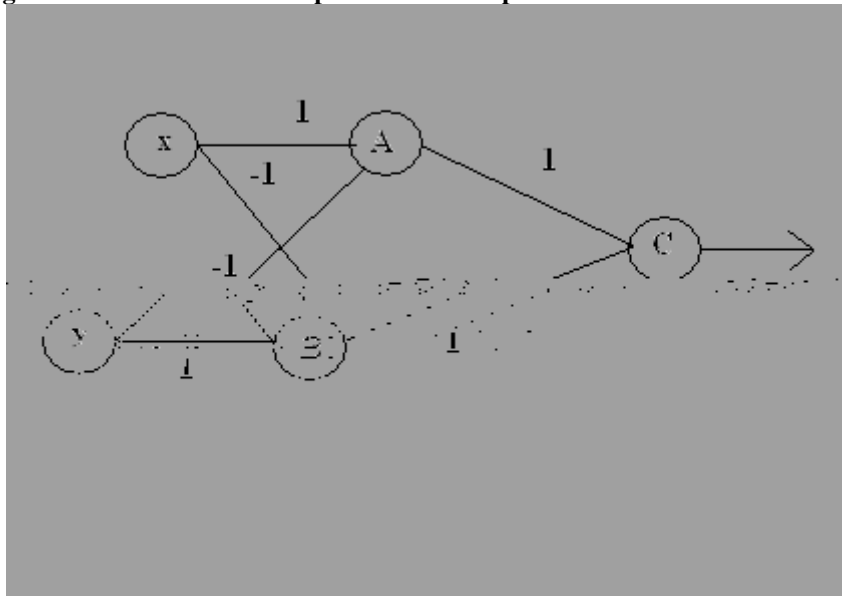
Formula: c36&, #3, D)(*#J?#J_#d #J, #<#D&'F&D(' , #K3.#<'/ , #K&3/2(*#K?#K_#d #K, #(%#
&<F2#9#(2&E;##M(#K3.#(2&, #F<F).<(&#J?K?#eJ_#e#d #eJ, K, #1#<, 5#(2&, #%) (D)(#<#*3, /.&#
6<.)&#!<'&5#% , #K2&(2&'#%# , %(#2<(#6<.)*#/'&#<(&'#2<, #B;

Figure 2: Example of a Perceptron



8&'F&D('% , *#F<, #! %%L&5#)D#K3(2#%, &#<, %(2&'#.3L&#, &)'%, *A#3, #'% '5#'(%#F'&<(&# , &(K%'L *#F%, *3*(3, /#%9#E).(3D.&#.<'&'*#%9#D&'F&D('% , *;##V%'#&J<ED.&A#*)F2#<#, &(K%'L#F<, # !&#)*&5#(%#F%ED)(&#(2&#+%%.&<, #.%/3F#9) , F(3%, #fP\$;##V3/)'&#^#3*#<#53</'<E#%9#2%K#5%3, /# J#fP\$#"#K%).5#K%'L;

Figure 3: A Network of Perceptrons that Computes the XOR Boolean Function



M, #(23*#F<*#A 4#A#+A#<, 5#-#<'&#D&'F&D('% , *;##4#K%).5#%, .''#%) (D) (#<#?#39#J#3*#?#<, 5#''# 3*#B;##+ #K%) .5#%, .''#%) (D) (#<#?#39#J#3*#B#<, 5#''#3*#?;##-#K%).5#%, .''#%) (D) (#<#?#39#4#?%'#+#3*# %) (D) ((3, /#<#?;##12&'&9%'&A#(23*#, &(K%'L#%9#D&'F&D('% , *#E%5&.*#(2&#fP\$#9) , F(3%, ;##M, # /&, &'<.A#D&'F&D('% , *#<'&#&9%'&'&5#(%#<*#, %5&*A#<, 5#D&'F&D('% , *#(2<(#%, .''#%) (D) (#6<.)&*#(2<(# /%#3, (%#E%'&#D&'F&D('% , *#<'&#F<..&5#2355&, #, %5&*

Network Training

12&#K<'<#&)'<.#, &(K%'L#.&<', *#3*#! ''#(<3, 3, /#3(*#K&3/2(#6<.)&*;##b<F2#(3E&#<, #
3(&'<(3%, #9#D<***3, /#(2, D)(*#(2%')/2#(2&#, &(K%'L#3*#5%, &A#(2&#%) (D) (#6<.)&*#<'&#
F%E D<'&5#</<3, *(#(2&#(<' /&#(<('!)&0*#6<.)&*;##12&, #<#F<.F).<(3%, #3*#D&'9%' E&5#(%#F' &<(&#
(2&.<#(2<(#E) *(!&#<55&5#(%#(2&#%'3/3, <.#K&3/2(##123*#F%, (3,)&*#), (3.#(2&#K&3/2(*#5%#
, %(#F2<, /&A#%'#), (3.#(2&#<E%), (#9#D' &N5&(&' E3, &5#3(&'<(3%, *#3*#&'<F2&5;##12&#&'&#&'6&'<.#
K<'*#(%#)D5<(&#K&3/2(*#9%'#D&'F&D('%, *;##P, &#K<'*#3*#(%#<***3/, #'<, 5%E#K&3/2(*#(%#&<F2#
3, D)(#<, 5#(2&, #3(&'<(3, /#(2&#D&'F&D('%, #)*3, /#(2	%..%K3, /#') .	%'#)D5<(3, /#&<F2#K&3/2(#
K_#9%'#3, D)(#J_7

Formula: $K_3 \leftarrow \#K_3 \#e\# \hat{A}k_3 \# \# \# \# \# K2\&' \& \# \# \# \# \hat{A}k_3 \cdot 1' \text{ fhg}\%I J_3$

M, #(2&#<! %6	%' E).<A# 'jg'hY'YUfb]b['fUY'i gY'hc'XYHfa]bY'hc'k \UhYI HYbh'hY'
K&3/2(#K3.#! &#F2<, /&5A#(#3*#(2&#(<' /&#(<('!)&0*#6<.)&A#%#3*#(2&#%) (D) (#6<.)&#/#36&, #! ''#(2&#
D&'F&D('%, A#<, 5#J_#3*#(2, D) (#6<.)	&5#3, (%#(2&#D&'F&D('%, ;##123*#E&(2%5#/) '<<, (&*#
F%, 6&' /&, F'#(2&#F.<***&'<'&#.3, &<'.'''*#D<'<! .&#H>3, *L''#S#8<D&'(A#?RC[I;#O%K&6&'A#
E%*(#5<(<*&'5%#%, %2<6&#.3, &<'.'''*#D<'<! .&#F.<***&*;
4*#%.) (3%, #(%#(23*#D%'! .&E#K<*#)*3, /#&'%'#! <FLD%'D</<(3%, ;##M, #(23*#K<'A#(2&#
) (D) (*#<'&#F<.F).<(5A#<, 5#(2&, #2&#&'%'#3*#&, (#1 <FLK<'5*#(2%')/2#(2&#, &(K%'L#(%#)D5<(&#
(2&#K&3/2(*;##b<F2#%) (D) (#D&'F&D('%, #L#F<.F).<(&*]rg'Yffcf'hYfa ' _#)*3, /#(2	%..%K3, /#
9%' E).<7

Formula: $L \leftarrow \%LH? \#g\%L IH(L g\%L I$

O&'&A#%_#3*#(2&#%) (D) (#6<.)	%'#(2&#D&'F&D('%, A#<, 5#(L#3*#(2&#(<' /&#(<('!)<.)&#;
12&*#6<.)&*#<'&#)*&5#3, #F<.F).<(3, /#(2&#&'%'#(&' E#9%'#&<F2#2355&, #, %5#)*3, /#(2&#
9%..%K3, /#9%' E).<7

Formula: $2 \leftarrow \#LH? \#g\%L \text{ k } L_2 L$

Z%(&#(2<(#K_L2#5&, %(&*#(2&#K&3/2(#9%'E#%, %5#(%#%, %5&#L;#12&#(&'E# 'k L2 L 9%'#
2355&, #, %5#3*#(2&'&9%'&#(2&#*) E#%9#(2&#K&3/2(*#(3E&*#(2&#&'%'#6<.)&*#%9#<.#(2&#%) (D) (#
, %5&*#L#(2<(#<'&#F%, , &F(&5#9%'E 2;

12&#K&3/2(*#<'&#(2&, #)D5<(&5#3, #2	%..%K3, /#K<'7

Formula: $K_{03} \leftarrow \#K_{03} \#e\# \hat{A}k_{03} \# \# \# \# \# K2\&' \& \# \# \# \# \hat{A}k_{03} \cdot 1' \quad J_{03}$

123*#)D5<(&#') .&#&'&5)F&*#(2&#E&<, *#G)<'&#&'%'#<#(2&#%) (D) (#<'&' ;##-%, , &F(3, /#
E<, ''#D&'F&D('%, *#%/&(2&'#K3(2#E<, ''#2355&, #<'&'*)*3, /#&'%'#! <FLD%'D</<(3%, #F<, #2&.D#
3ED%'6&#<FF)'<F''!)(#&<5*#(%#<#2<'D#3, F'&<*, #'<3, 3, /#(3E&##T<'''3, /#(2&#(2'&'2%.5A#
(2&#,)E!&'#%9#<'&'*A#<, 5#(2&#<E%), (#%9#(3E&#<L&, #(%#(<3, #2&#, &(K%'L#F<, #<..#2&.D#
3ED%'6&#<FF)'<F'';##1&*(#2<6&#!&&, #5%, &#)*3, /#(2&*ᔗ&'&, (#6<'3<(3%, *;H\$)E&.2<'(A#
h35%'KA#S#X&2'A#?RR\I;

Y*3, /#&'%'#! <FLD'D%</<(3%, #3, #(&#K<'<'#!%6&#) *&*<#(2'&*2%.5#3, #'5&'#(##
 F%, 6&' /&#(%##, <.)&;#O%K&6&'A#(2&'*#<#K<'#(%#E%539''#&'%'#! <FLD'D%</<(3%, #K2&'&#
 3, *(&<5#9#) *3, /#(2&#(2'&*2%.5A#<#D'#! <1.3.3(''3*#3, *(&<5#F%ED) (&5A#<, 5#(2<(#3*#) *&5#3, #
 5&(&'E3, 3, /#(2&#%) (D) (#<((!3!)) (<.)&;###, #(&23*#K<'A#(2<.)	%'#(2&#&'%'#(E'9%'#&<F2#
 %) (D) (#, %5&#L#K%) .5#(2&, #! &#F<.F) .<(&5#3, #(&2	%..%K3, /#K<'7

Formula: $L \leftarrow \#_L H(L \text{ g} \#_L I)$

12<.)	%'#(2&#&'%'#(&'E'9%'#&<F2#235&, #, %5#K%) .5#(2&, #! &#F<.F) .<(&5#3, #
 (2	%..%K3, /#K<'7

Formula: $2 \leftarrow \#_L k_{L2} L$

M, #(&23*#K<'A#(2&#(2'&*2%.5#K%) .5#! &#&E%6&5#<, 5#! &#&D.<F&5#K3(2#<#D'#! <1.3.3(''##
 T<'3%) *#K%'L#2<#*! &&, #5%, &#) *3, /#(23*#E%5393&5#9%'E#<#K&..#H+3*2%DA#?RRCI

Bayesian Techniques

4, %(2&'#("D	#F.<*393&'#) *&#*#<+'*3<, #&<*, 3, /;##<+'*3<, #(&F2, 3G) &#<'&#
 !<*5#%, #D'#! <1.3.3(''53*('3!)) (3%, *#<, 5#(2<(#) *3, /#(2&*&#D'#! <1.3.3(3&*#%, #'! *'6&5#5<(<#F<, #
 3ED'%6&#D&'9%'E<, F&;###('#3&*(%#D'%5) F&#(2&#! *#(2''D%(2&*3*9%'E#*%E&#*D<F	#
 2''D%(2&*&#O/36&, #*%E&#(<3, 3, /#5<(<#` ;#>%*(#9#<+'*3<, #.&<', 3, /#&.3&#*%, #<+'*0
 (2&%&E;#<+'*0#(2&%&E#<*) E&#*(2<(#9%'#&<F2#2''D%(2&*3*#2A#(2&'*#<#D'3%'#D'#! <1.3.3(''#
 <.'&<5''F<.F) .<(&5A#F<..&5#8H21;#8H` 1#3*#(2&#D'3%'#D'#! <1.3.3(''#(2<(#(2&#('<3, 3, /#5<(<#` #K3..#
 !&#%!' *'6&5;#8H` i21#3*#(2&#D'#! <1.3.3(''#(2<(#(2&#('<3, 3, /#5<(<#` #K3..#! &#%!' *'6&5#/36&, #(<2<(#
 (2''D%(2&*3*#2#2%.5*#4, 5#8H2i` 1#3*#(2&#D'#! <1.3.3(''#(2<(#(2''D%(2&*3*#2#K3..#2%.5#
 /36&, #(&2&#('<3, 3, /#5<(<#` ;#1%#5&(&'E3, H2i` 1A#<+'*0#(2&%&E#D'%635&*(2	%..%K3, /#
 9%'E) .<7

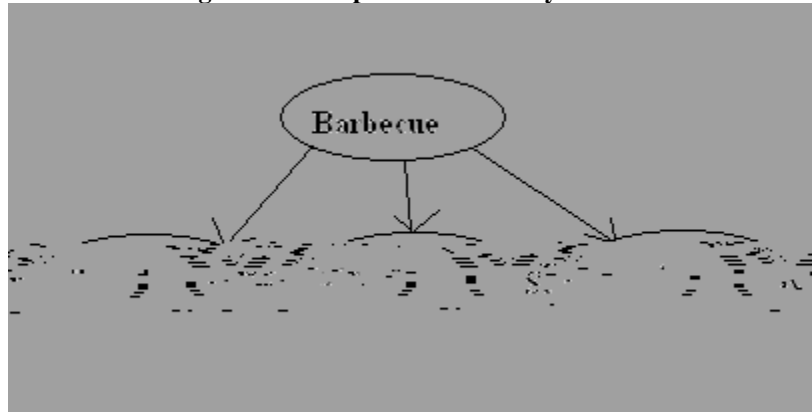
Formula: $8H2i` 1#j \#8H` i218H21 k 8H` 1$

<+'*0#(2&%&E#<..%K*#F.<*393F<(3%, #!' *'&. &F(3, /#(2&#E<J3E) E#<#D%*(&'3%'3#
 H> 481#2''D%(2&*3*#V%#&6&'''#2#3, #O#8H` i218H21#3*#F<.F) .<(&5;##12&, %E3, <(%'#8H` 1#3, #
 <+'*0#9%'E) .<#3*#&E%6&5#! &F<)*#(2&#('<3, 3, /#5<(<#` #, &6&'#F2<, /#*9%'#&<F2#F2<, /#%9#
 2#3, #O;##12''D%(2&*3*#2#K3(2#(2/2*(<.)	#8H` i218H21#3*#(2&, #) *5#(#F.<*39''#
 &<F2#3, *(<, F&;

Naïve Bayes

12&'&#<'&#(K%#*(<, 5<'5#E&(2%5*3, #K23F2#<+'*3<, #.&<', 3, /#3*5%, &;#P, *#<#
 (&F2, 3G) &#F<..&5#Z<16&#<+'*#;###, #Z<16&#<+'*A#(2&#<./%'3(2E#F'&<(&#<#*#(#<..#D%*3! .&#
 (<' /&#(<((!3!)) (&#;###(2&, #F<.F) .<(&*(2&#D'#! <1.3.3(''#(&'E*#8H21#<, 5#8H` i21#<#*#(<(&5#<'%6&;#
 O%K&6&'A#3(#! '&<L*#&<F2#5399&'&, (#<((!3!)) (<.)#9#(2&#('<3, 3, /#5<(<#` #<, 5#F<.F) .<(&*#
 &<F2#%, 	#(2&E#*#D'<(&.';###(2&, #(<L&*#(2&#E<J3E) E#6<.)	#8H21#E) .(3D.3&5#! ''#(2&#
 D'%5) F(#9#<..#(2&#D'#! <1.3.3(3&*#9#8H` i21;#V%#&J<ED.&A#F%, *35&'#V3/) '&#;##

Figure 4: Example of a Naïve Bayes model



+<'! &F) *(2&#(<' /&#(<('3!)(&A#<, 5#2<*(2<.) &*#' &*'%'#, %;#12&#F.< **393&'#K%).5#
93'*(#!)3.5#)D#(2ᔗ&' &, (#3, *(<, F&*'%'9#(2&#! <'! &F) <(<#` #<, 5#F<.F).<(&#&<F2#D' %! <! 3.3(' '#
/36&, #&3(2&' #' &*'%'#, %;#M(#K%).5#<.*%'#F<.F).<(&#(2&#%6&'<..#D' %! <! 3.3(' '#9#! <'! &F)3, /;#V%'#
&J<ED.&A#(%#F.< **39''#(2, *(<, F&#H:), #j #%) (A#\$<3, #j #, %, &A#1&ED&'<() '&#j #2%(1A#(2&#
F.< **393&'#K%).5#F<.F).<(H'' &*l#m#8H%) (i'' &*l#m#8H, %, &i'' &*l#m#8H2%(i'' &*l#<, 5#8H, %l#m#
8H%) (i, %l#m#8H, %, &i, %l#m#8H2%(i, %l#12&#&, 5#' &*) .(#K%).5#! &#(2&#F.< **393&'#* &. &F(3, /#(2&#
E<J3E)E#6<.)&#%9#(2&#(K%#<, 5#F2%* &#&3(2&' #' &*'%'#, %;#Z<16&#+<' &*'2<*'! &&, #) * &5#9%'#
(< *L *#*)F2< *#*%' (3, /#%) (#, &K *#<' (3F.& *#H] %<F23E *A#?RRCl;

Bayesian Networks

$Z \perp\!\!\!\perp \{X_1, \dots, X_m\} \mid Y_1, \dots, Y_m$

Definition: Attributes are **conditionally independent** of one another if given the value of one or more attributes $Y_1 \dots Y_m$ determines the value of attributes $X_1 \dots X_m$ independent of values of attributes $Z_1 \dots Z_m$

$P(X_i = x_i \mid Y_1 = y_1, \dots, Y_m = y_m, Z_1 = z_1, \dots, Z_m = z_m) = P(X_i = x_i \mid Y_1 = y_1, \dots, Y_m = y_m)$

Figure 5a: Example of a Bayesian Network

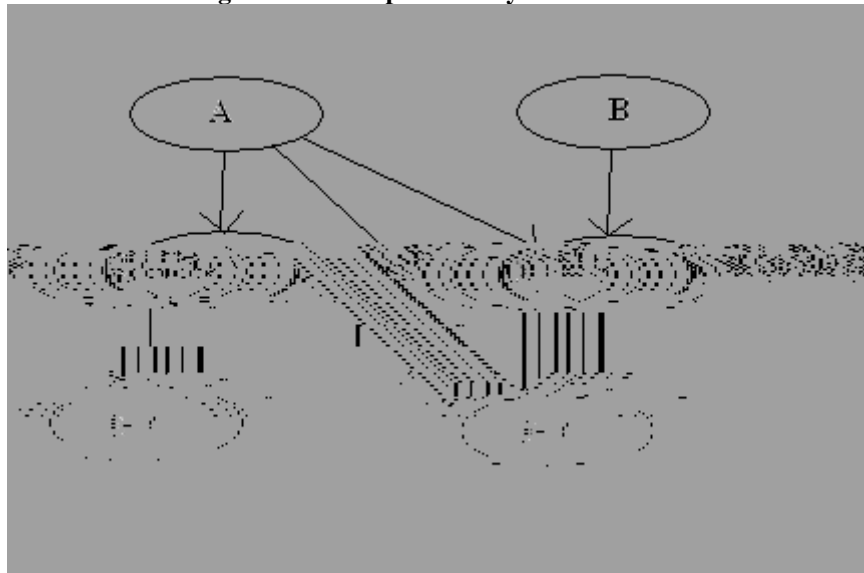


Figure 5a: Example of a Conditional Probability Table

	A=0	A=1	A=0	A=1
9	B; \	B; ?	B; 0	B; @
n9	B; C	B; R	B; @	B; 0

$P(X_i = x_i \mid Y_1 = y_1, \dots, Y_m = y_m, Z_1 = z_1, \dots, Z_m = z_m) = P(X_i = x_i \mid Y_1 = y_1, \dots, Y_m = y_m)$

K%).5#! HVj 9i4j <A` j 5l;##Y*3, /#(2&#F%, 53(3%, <.#D'#! <! 3.3(''#(<! .&#<! %6&A#(23*#6<.) &#
K%).5#(2&'&9%'&#! &#B;\;

Y, .3L&#, &)'<.#, &(K%'L*A#3, #K23F2#&'%'*#K<, (#%#! &#E3, 3E3=&5A#+<'&*3<, #, &(K%'L*

F<, #! &#('<3, &5#(%#E<J3E3=&#(2&#D'#! <! 3.3(''#%9#(2&#%!*&'6&5#5<(<#/36&, #(2&#, &(K%'L#

D<'<E&(&'*;#h &3/2(*#H/Tf<ETB.15Tf<ETW2<*%1%50#*239(<K%\)D5<1(F<)j 9i4j <A` j 5#(2&#, &(K%'L#23*#6<.)

Machine Learning Work with Recommender Systems

><, "#F%..<!%'<(36].(&'3, /#(&F2, 3G)& *#2<6&#! &&, #) * &5#! &9%'&#V%'#&J<ED.&A#
\$3, /%#K<*#<#*%F3<.#3, 9%' E<(3%, #93.(&'#) * &5#(%#('"'#<, 5#E<L&#E) *3F#' &F%EE&, 5<(3%, *#(%#
.3*(&, &'*#H: 2<'5<, <, 5#S#><&*A#?RR[I;##M, #*) F2A#&F%EE&, 5<(3%, *#<'&#D<*&5#(2'%) /2#
pK%'5#%9#E%) (2qA#<(2&'#(2<, #F%, (&, (A#*%#E) *3F#' &F%EE&, 5<(3%, *#E<"#! '<*(3F<..'#"#
5399&'&, (#(2<, #(2%*&#(2<(#<#) * &'#2<*#.3*(&, &5#(%#! &9%'&'<6&#! &&, #E<, "'#%(2&'#(&*(#

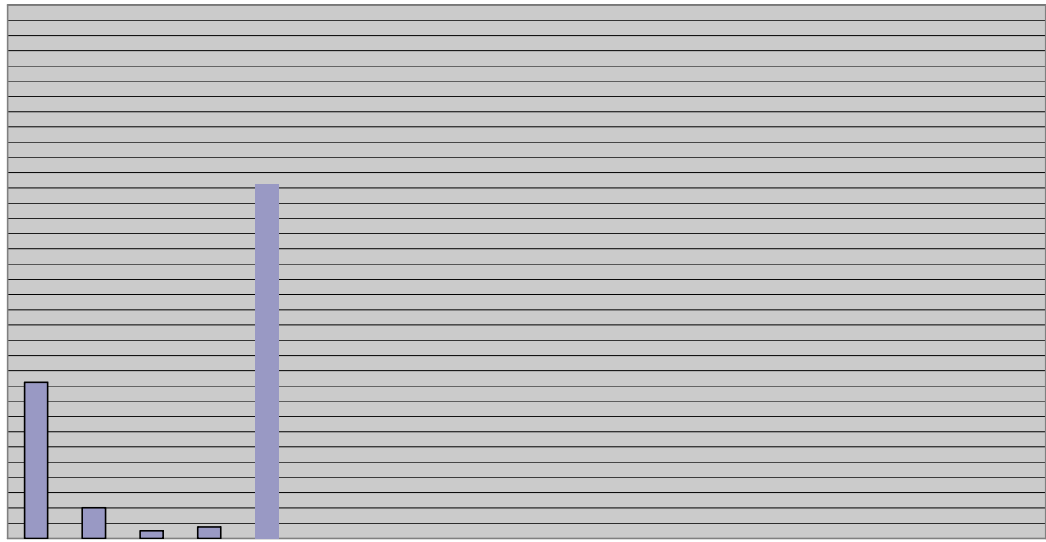
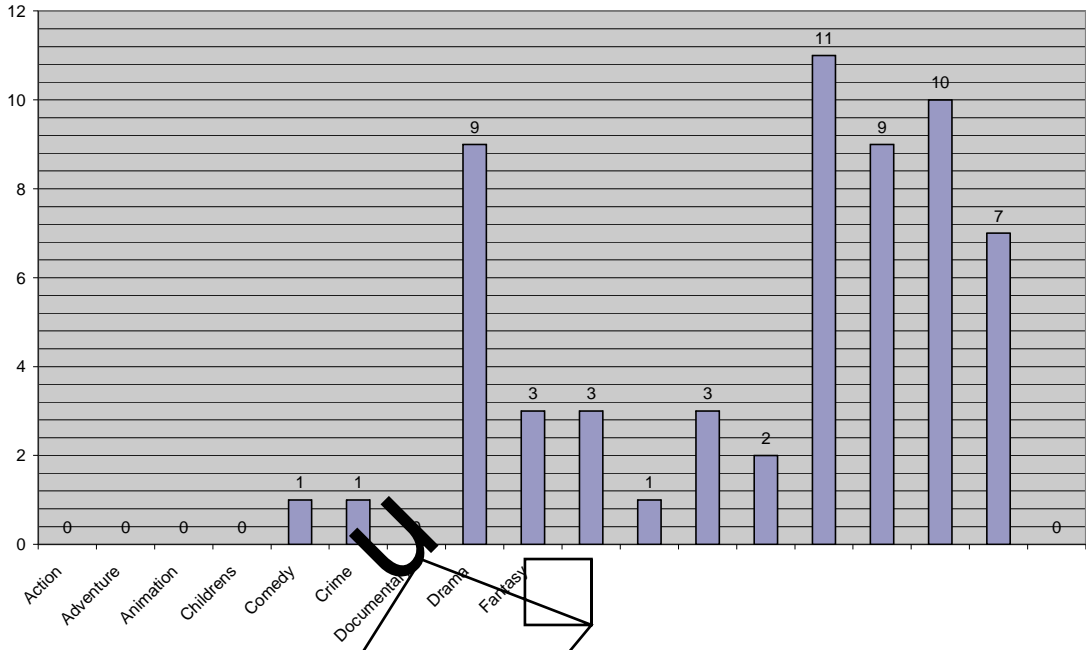
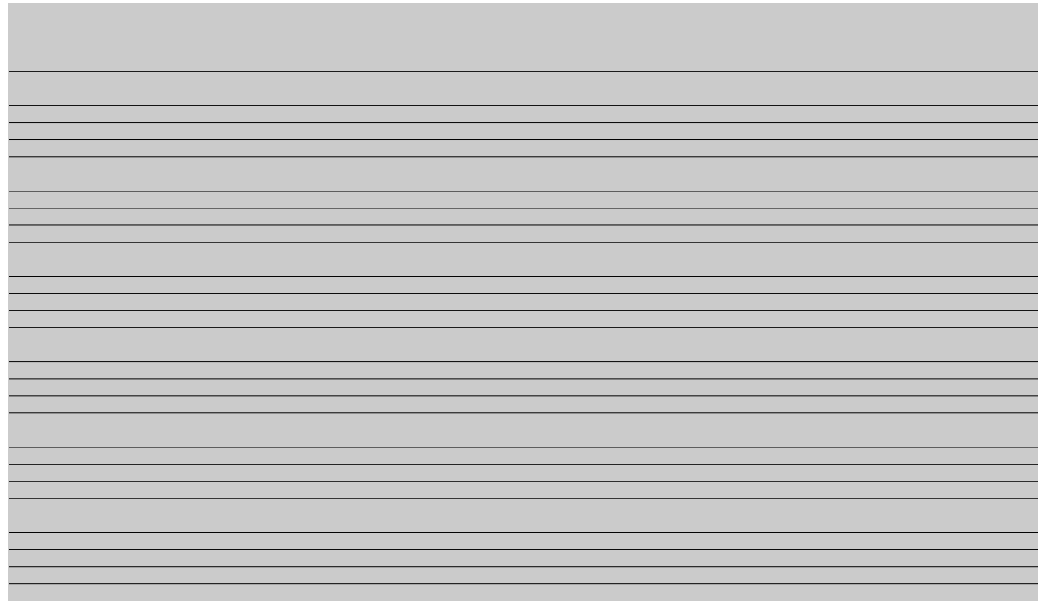


Figure 7c: Genre3 Distribution





Performance Metrics and Evaluation Protocol

$$Accuracy = \frac{\text{Number of Correctly Classified Instances}}{\text{Total Number of Instances}}$$

Definition: *Accuracy* is the percentage of instances that are correctly classified by the system.

Definition: *Precision* is the percentage of like predictions that agree with the user's taste.

$$Precision = \frac{\text{Number of True Positives}}{\text{Number of True Positives} + \text{Number of False Positives}}$$

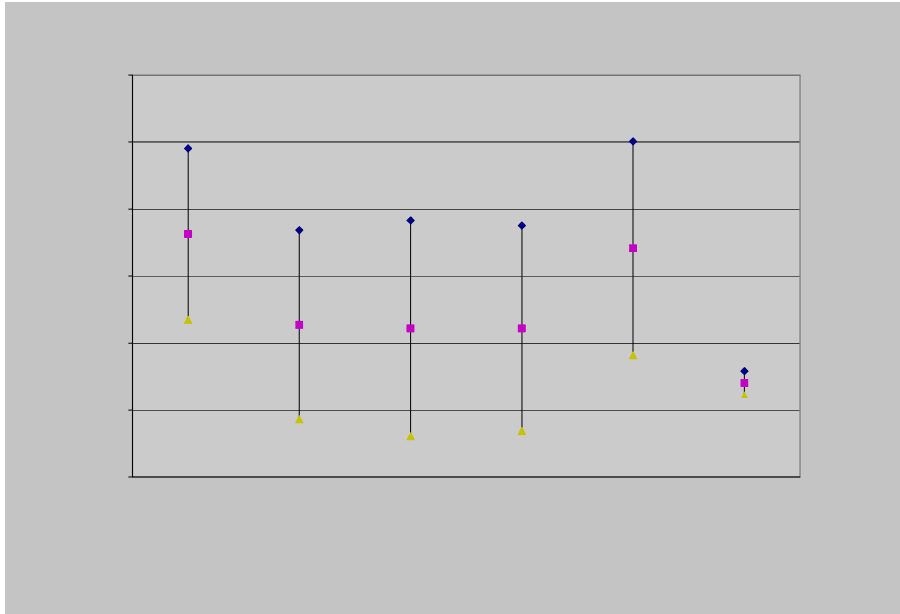
Results

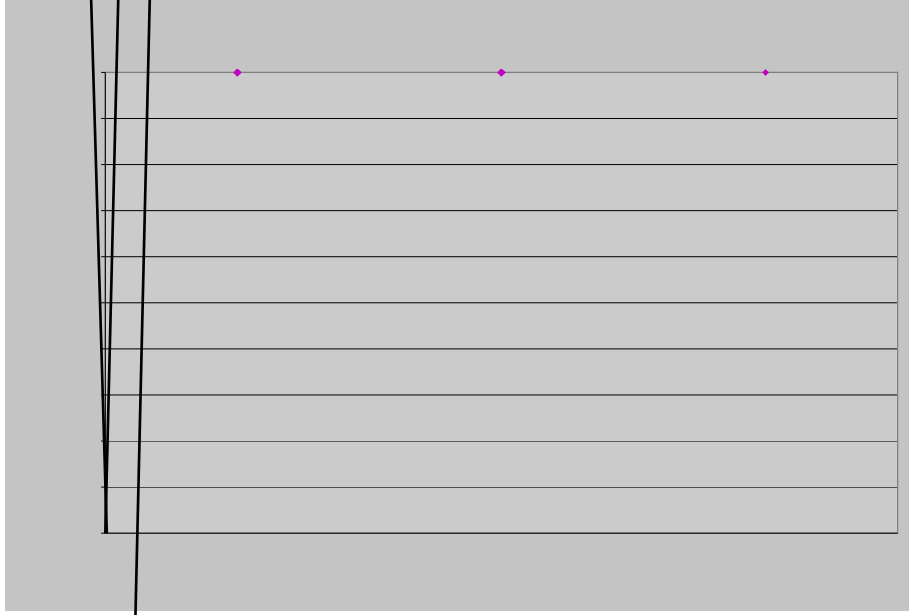
Attribute Selection

Mathematical symbols and characters: $\mathbb{M}(\&^*(\&5\#1\%(2\#D'3, F3D<.\#F\%ED\%, \&, (*\#<, 5\#+\&^*(V3'^*(\#(\%\#*\&.\&F(\#(2\&\#<((('3!)(\&^*(\%\#1\&\#9\&5\#<^*3, D)(^*\#(\%\#(2\&\#6<'3\%)*\#E<F23, \&\#.\&<', 3, /#\<./\%'3(2E^*;\#$

Principal Components

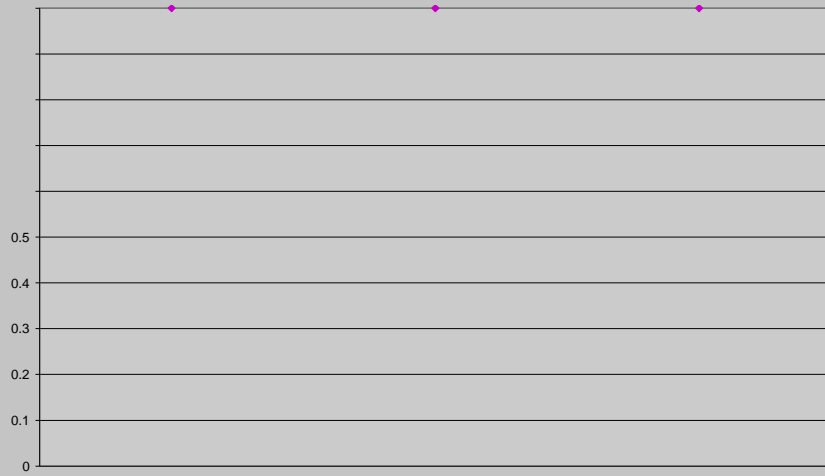
Mathematical symbols and characters: $V3/)' \&^*\#o<A\#o! A\#<, 5\#oF\#^*2\%K\#(2\&\#D\&'9\%'E<, F\&\#9\#(2\&\#5399\&' \&, (\#E<F23, \&\#.\&<', 3, /#\<./\%'3(2E^*\# \&6'\#5<(<\#(2<(\#K<^*\#D'\&D'\%F\&^*\&5\#)^*3, / \#D'3, F3D<.\#F\%ED\%, \&, (*\#<((('3!)(\&^*\&.\&F(3\%, ;$

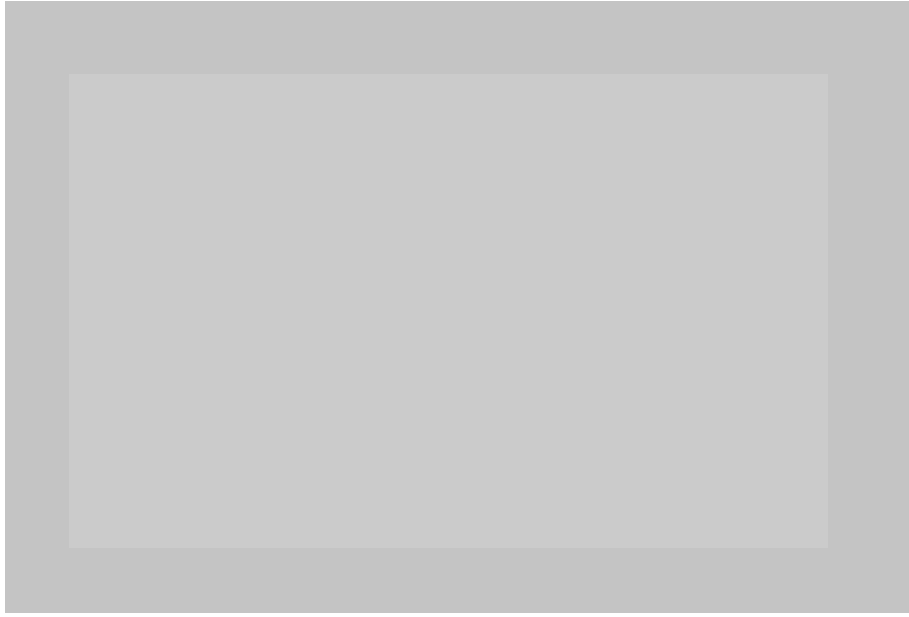




BestFirst

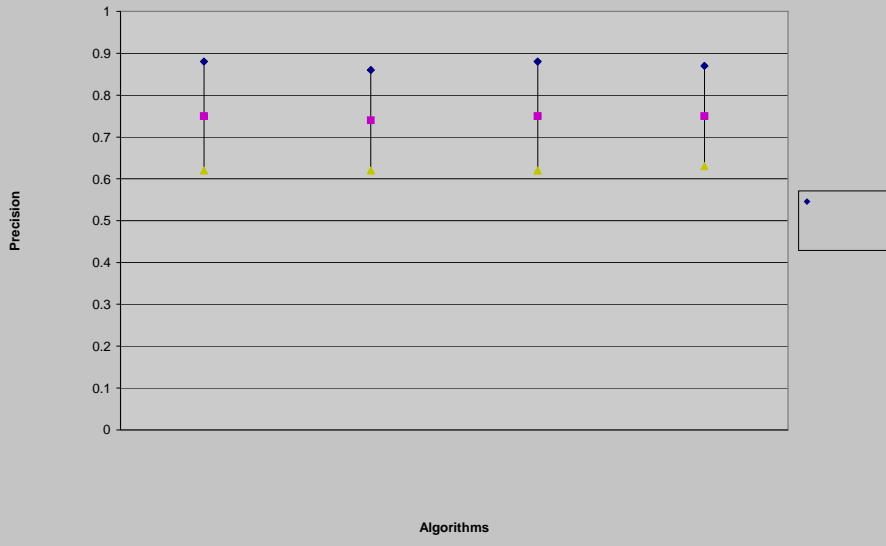
Figure 9c: Top-10 Precision Using BestFirst Attribute Selection

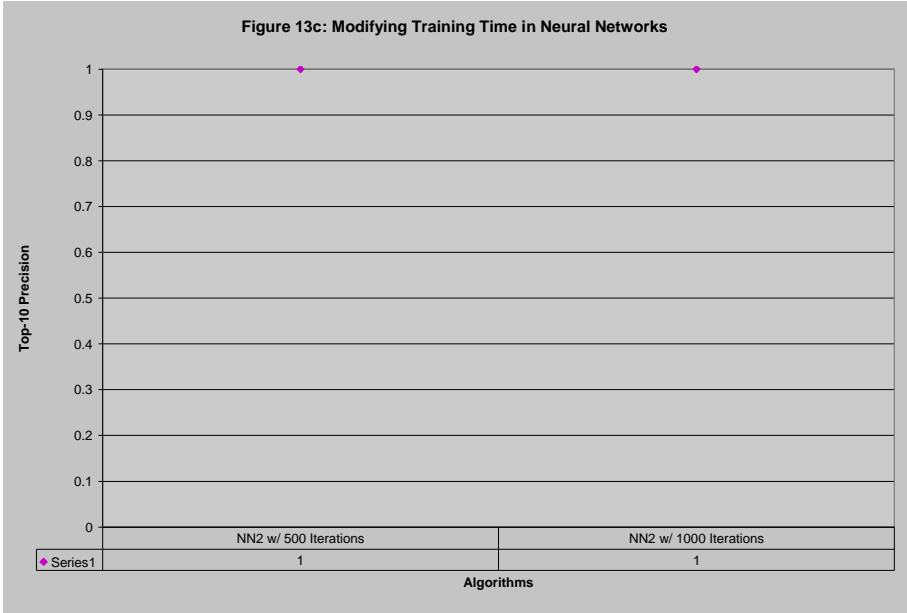




Neural Networks

Figure 12b: Modifying Hidden Nodes in Neural Networks

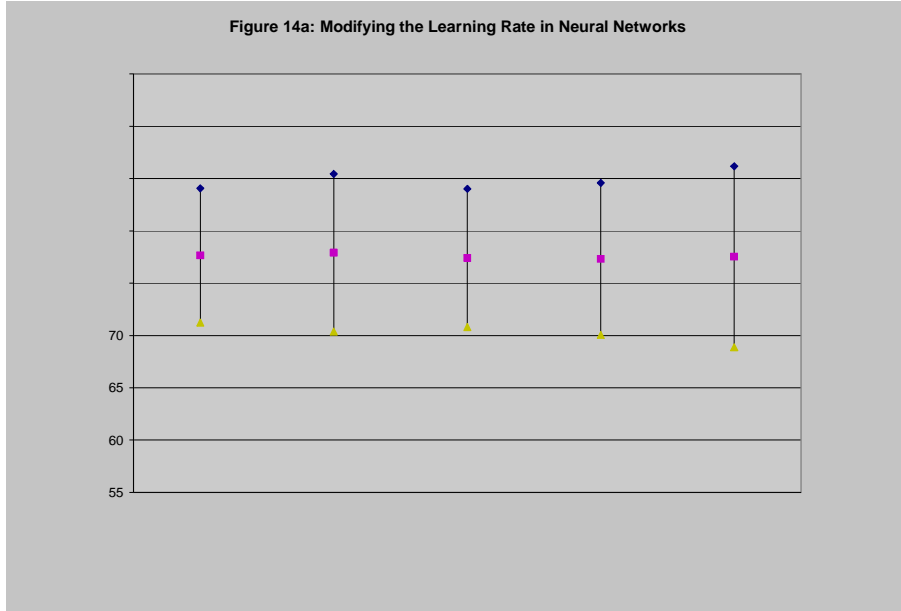




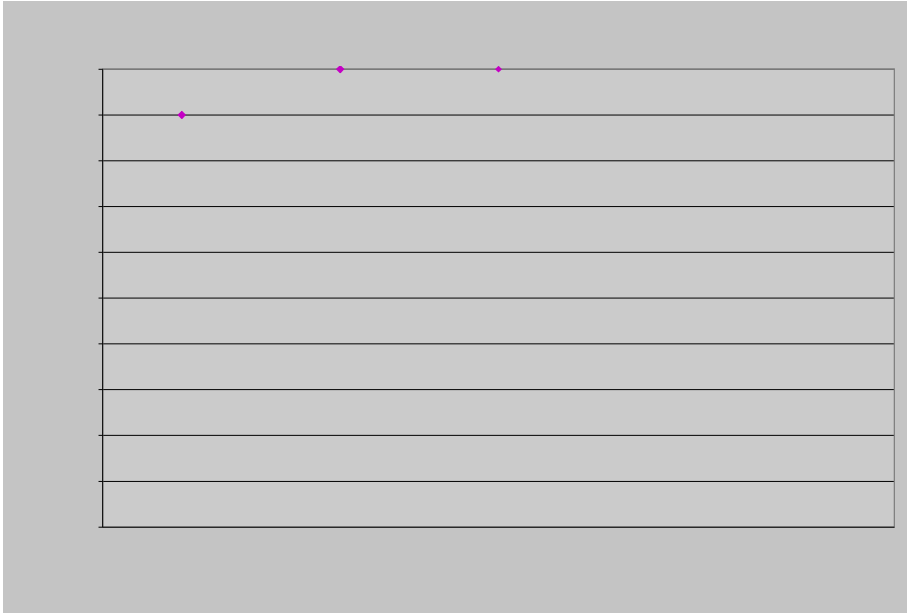
X%%L3, /#<(#(2&# ' & *).(*A#(%DN?B#D' &F3*3%, #K<*#(2&#* <E&*D3(, F' &<*3, /#(2&# ('<3, 3, /#(3E& ; #4FF)'<F''#*.3/2(.''#5&F'&<* &5A#<, 5#(2&#* (<, 5<'5#5&63<(3%, #*.3/2(.''#3, F' &<* &5;## 8'&F3*3%, #<.*%#*.3/2(.''#5&F'&<* &5;##1 2&' &9%'&A#5&*D3(, F' &<*3, /#(2&#('<3, 3, /#(3E&A#(2&# <FF)'<F''#<, 5#*(<, 5<'5#5&63<(3%, #535#, %(#2<6&#<#*3/, 393F<, (#F2<, /&##1 2&' &9%'&A##535#, %(# !%(2&'#3, F' &<*3, /#(2&#('<3, 3, /#(3E	%'#9)())'&##JD&'3E&, (*;

Learning Rate

Learning rate is a hyperparameter that controls how much the weights in a neural network are updated during training. A high learning rate can cause the model to diverge, while a low learning rate can cause it to converge very slowly. The optimal learning rate depends on the specific problem and model architecture.

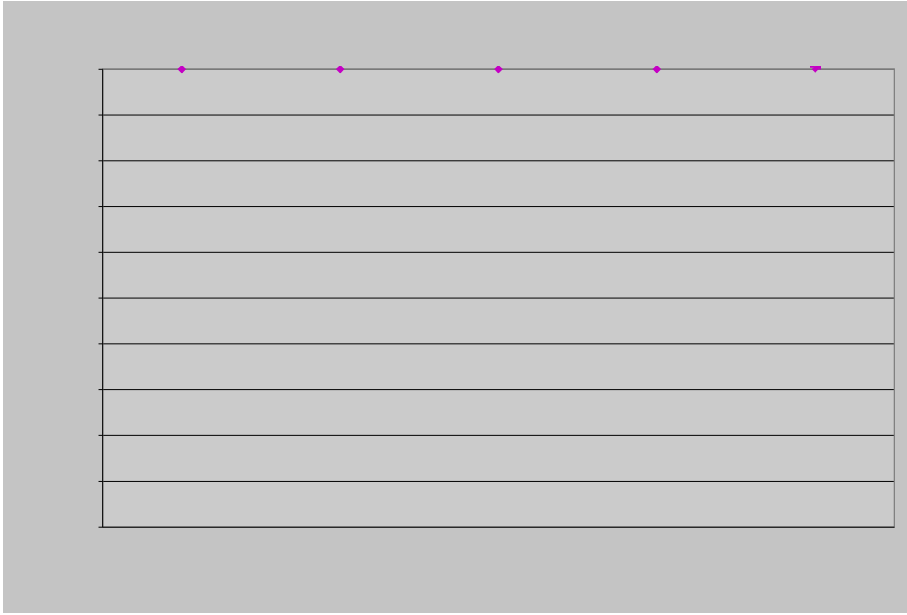


70



Bayesian Networks

W#5&F35&5#(%#(&*(#(2&#E<J3E)E#,)E! &'#%9#D<'&, (*#&<F2#, %5&#F<, #2<6&#%, #(2&#

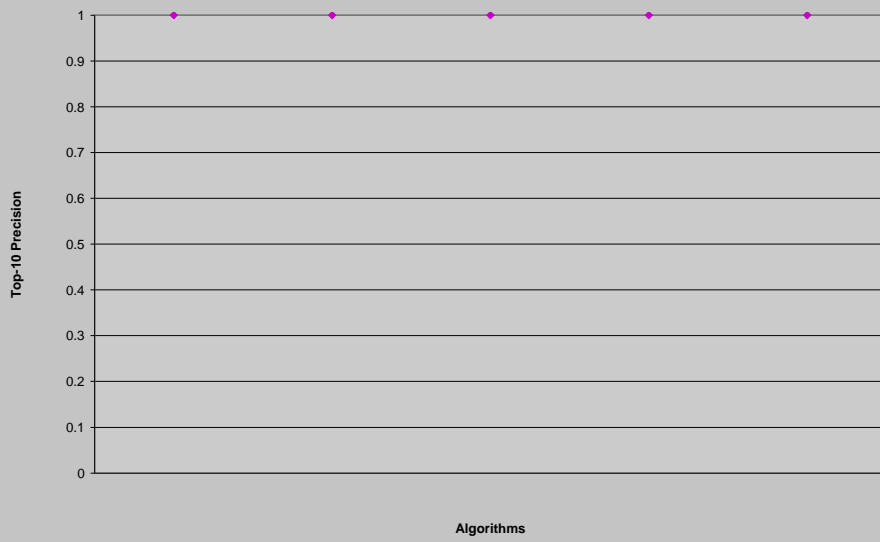


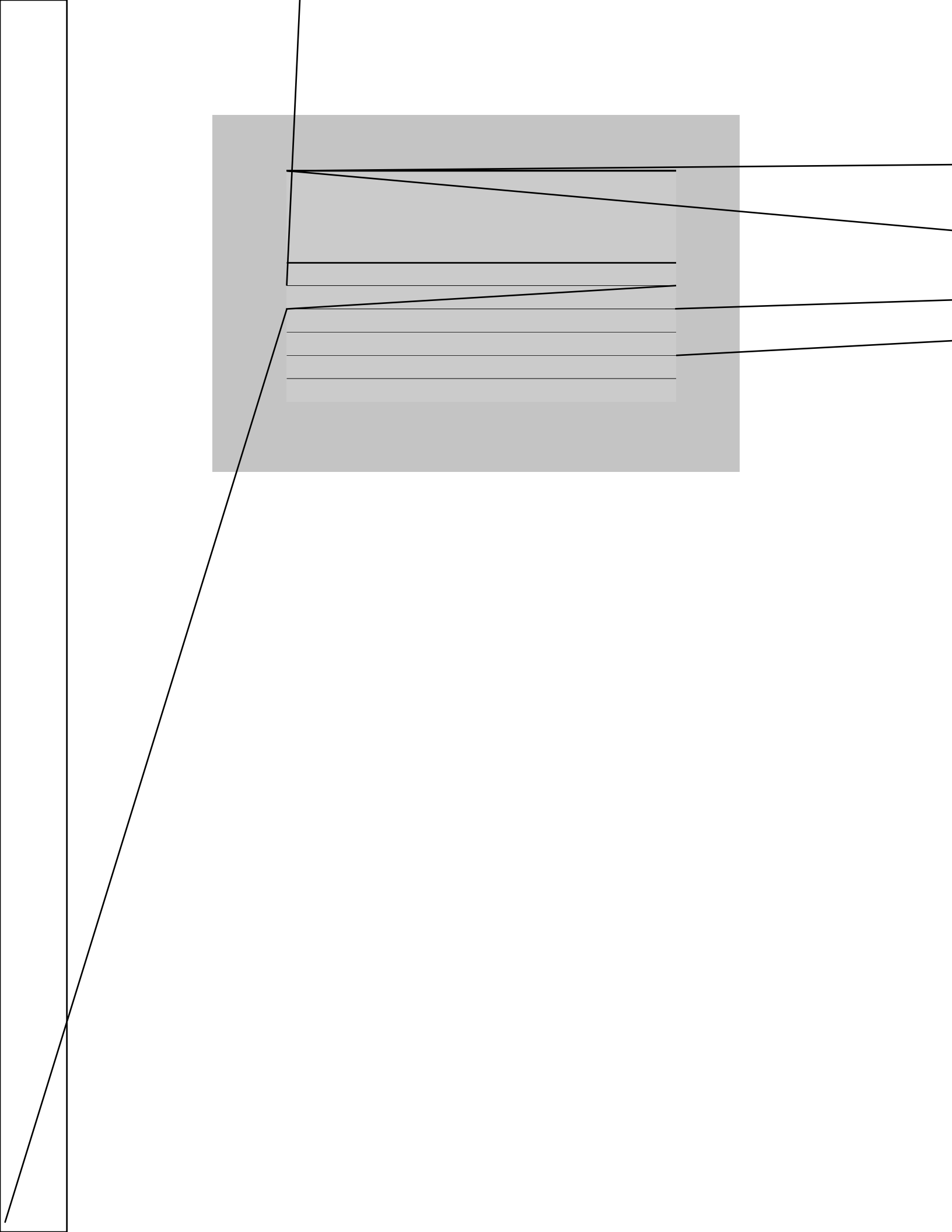
Collaborative Data Only

V3/) '&*#?C<A#?C! <, 5#?CF#*2%K#(2&#/'<D2*#9%'#F%..<! %'<(36&N%, .''#+<'&*3<, #, &(K%'L*;

The image shows a large gray rectangular area. Inside this area, there is a smaller white rectangular area. This white area contains eight horizontal lines, which typically represent redacted text or a placeholder for data in a table or document.

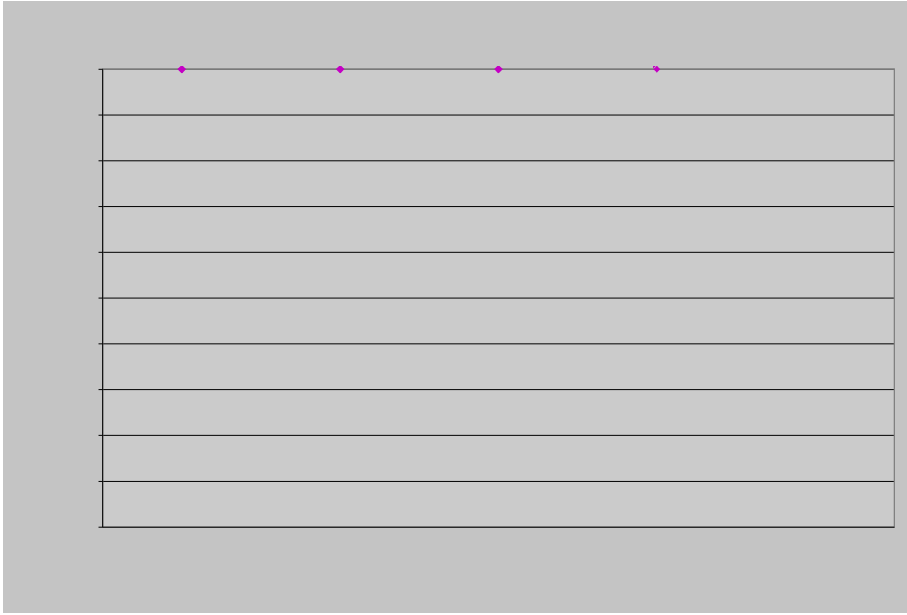
Figure 16c: Varying the Number of Parents in Bayesian Networks (Collaborative-Only)





Enhanced Content Data





Conclusions and Future Work

12&#D)'D%*	#(23*(2&*3*#K<*&#%#<, <.'=&#E<F23, &#.&<', 3, /#(&F2, 3G)&*#<, 5#<DD.'"# (2&E#(#&'&F%E E&, 5&'#*''*(E*#3, #%'5&'#%#E<L&#&'&F%E E&, 5<(3%, *#9%'#E%63& D<('%, *;#+'#<DD.'"3, /#5399&'&, (#('<(&/3&*#(%#!(2#(2<<*&#(<, 5#(2ᔗ&'&, (<./%'3(2E*A#(2%D&#K<*&# (%#93, 5#<, #D(3E3=&5#K<'#%9#D'&53F(3, /#K2&(2&'#)*&'*#K%).5#.3L&#F&'(<3, #E%63&*#!<*&5#%, #F%, (&, (#3, 9%'E<(3%, #<!)#(2&#E%63&#<, 5#F%..<!'%<(36, 9%'E<(3%, #9%'E%#(2&'# E%63&/%&'*0#<(3, /:*#V'%'E#(2&#K%'L5%, (#F<, #1&#F%, F.)5&5#(2<(#F&'<3, #E<F23, &#&<', 3, /#<./%'3(2E*#F<, #3, 5&&5#D&'9%'E#K&.#3, #E<L3, /#E%63&#&'&F%E E&, 5<(3%, *;

M, #(&'E*#%9#(%DNZ#D'&F3*3%, #, &)'<.#, &(K%'L*#<, 5#D'3, F3D<.#F%ED%, &, (*#D&'9%'E# &J('E&.'#K&.#9%'#<.#%K#6<.)&#%9#Z#1,)E!&'%9#E%63&*#&'&F%E E&, 5&5!;##123*#3*6&'# 3ED%'<,(#9%'#E%63&#&'&F%E E&, 5<(3%, *#<#&'E%*(#)*&'*#K%).5#%, .'"#K<, (#(%#*#'(2'%) /2#?B# %'#9&K&'#E%63&*#3, #%'5&'#(%#D3FL#%, &##12	<F#(2<(#(%DNZ#D'&F3*3%, #'&E<3, *#23/2#)D#(%# Zj @BB#3*#<.*%#<#/%%5#*%)'F&#%9#%(2&'*#%'(#%9#&'&F%E E&, 5<(3%, *#A#*)F2#<#&'&F%E E&, 5<(3%, *# 9%'#D&#D.&#K2#%!)'"#%'&, (# T` *#%#&'#K<(F2#E%63&*#%, #(&.&63*3%, ;##12%*#&D&#D.&#E3/2(# , &&5#<#13//&'#3*(#<*(2&'#K<, (#<#3*(#%9#E%63&*#(%#K<(F2#<(2&'#(2<, #0)*(#<#&.&F#9&K;

+<'&*3<, #, &(K%'L*#<(L&#E)F2#.&*#'(<3, 3, /#(3E&#(2<, #, &)'<.#, &(K%'L*#K23.&# 2<63, /#(2&#<E&#(%DN?B#D'&F3*3%, #*#%#3(#F%).5#!&#E%'&#)*9).#9%'#&N'<3, 3, /#K2&, #)*&'*# <55#(2&3'#.3L&*#<, 5#53*.3L&*#3, (%#<#*''*(E;##4.*#A#3, #(&'E*#%9#<FF)'<F"##+<'&*3<, #, &(K%'L*# 53*D.<'#<#3/ , 393F<, (#3ED'%'6&E&, (#%6&'# , &)'<.#, &(K%'L*#K3(2#!%(2#(2&#&'<, 5<'5#6&'*3%, # <, 5#E%5393&5#6&'*3%, #%9#&'%'#D'%'D</<(3%, #<#*#K&.#<#5&F3*3%, #('&#*#Z<16&#'+<'&*#A#?#&#<, 5# B\$F.<*#393&'*#;##+<'&*3<, #, &(K%'L*#D&'9%'E#K&.#K3(2#!%(2#D'3, F3D<.#F%ED%, &, (*#<, 5#(2&# +&'(V3'*(#<('3!)&#*#&.&F(3%, #*#%E&(23, /#(2<(#, %, &#%9#(2&#%#(2&'#<./%'3(2E*#(&'&5#5%, ##M#3*# F.*#&#K3(2#?#, #D'3, F3D<.#F%ED%, &, (*#!)(#(23*#F%).5#!)&#(%#(2&#*3=&#%9#(2<<*&#(##4# 5<<*&#(K3(2#E%6&#E%63&*#9%'#&<F2#''&'&#F%).5#D%*#3!.'"#%K&'#?#0*#D&'9%'E<, F&#

+&'(V3'*(#<DD&<'*#(%#!&#<, #&99&F(36&#(&F2, 3G)	%'#<('3!)&#*#&.&F(3%, ;##M('5)F&*# (2&# ,)E!&'%9#&#<('3!)&#*#F%, *35&'<!'##/36&*#?BB#r #(%DN?B#D'&F3*3%, #<, 5#3, F'&<'&*#(2&# E<J3E)E#<FF)'<F"##9%'E#(2&#B\$#!<*&3, &##+&'(V3'*(#3*#<.*%#9<*&'#(2<, #D'3, F3D<.# F%ED%, &, (*#3, #<(2<(#3(#*#&.&F(*#9&<)'&'&'<(2&'#(2<, #<DD.'"3, /#K&3/2(*#(%#<.#%9#(2&E#*#3(#3*# !&((#3, #(&'E*#%9#<FF)'<F"##D'&F3*3%, #<, 5#(3E&#A#!)(# , %(#3, #%6&'<.#(%DNZ#D'&F3*3%, ;

12&'%&*# , #(<DD&<'#(%#!&#<#13/F2<, /#&3, (%DN?B#D'&F3*3%, #<FF)'<F"##%'# D'&F3*3%, #K2&, #F%ED<'3, /# , &)'<.#, &(K%'L*#K3(2#(2&#&'<, 5<'5#6&'*3%, #%9#&'%'# # !<FLD'%'D</<(3%, #<, 5#(2&#E%5393&5#6&'*3%, ;##>'&#(&'(3, /#K%).5# , &&5#%#%, &#%6&'# #. , /&'#<E%), (*#%9#(3E&#(%#*#'#%, *#3/ , 393F<, (.'"#&((&'#(2<, #<2&#%#(2&'# ;

>%&#(&'(3, /#F%).5#!%, &#% , #<2&#D<'<E&#&'&'(3, /#%9# , &)'<.#, &(K%'L*#3, #%'5&'# (%#5&(&'E3, &#K2<#D<'<E&#&'#6<.)&#*#D(3E3=&#<FF)'<F"##A D'&F3*3%, #<, 5#(%DN?B#D'&F3*3%, ;##M#<DD&<'*#(2<#<50)*3, /#(2&#&'<3, 3, /#(3E&#A#&'<3, /#<(&#<, 5#%(2&'#6<'3<!'&*#5%&*# , %# 3/ , 393F<, (.'"#F2<, /#&(2&#<FF)'<F"##<, 5#(2&#D'&F3*3%, #%9#(2&#&'&F%E E&, 5<(3%, *;

12&#&'*#).(*#%9#(2&#%#D(3E<#<./%'3(2E*#&JF&&5#(2&#! , F2E<'L*#*#&'#B\$ 3/ , 393F<, (.'"##12&#!&'#D&'9%'E3, /#<./%'3(2E*#2<6&#?BB#r #(%DN?B#D'&F3*3%, #K23F2#3*#<# 6&'#3ED%'<, (#<(3*(3F;#12&'#<.*#3, F'&<'&5#(2&#<FF)'<F"##<, 5#D'&F3*3%, #9%'E#C@r #(%#o^r <, 5#C@r #(%#oBr #*#D&F(36&#.';

M#<DD&<'*#(2<#)*3, /#F%, (&, (N%, .'"##F%..<!'%<(36&N%, .'"##%#<#F%E!3, &5#5<(<*%&*# , %#(F2<, /#&(2&#(%DN?B#D'&F3*3%, #<#<#<.#(2'&<(<*&'*3%, *#/<6&#?BB#r #(%DN?B#D'&F3*3%, ;## ?BB#r #(%DN?B#D'&F3*3%, #3, #F%, (&, (N%, .'"#5<(<#3*#&'D&F3<..'"#/%%5#9%'#<# , &K#E%63	%'#K23F2# , %#F%..<!'%<(36<(<#3*#<6<3.<!.&#;##-..<!'%<(36<(<#% , .'"#D'%'6&5#E%#&)*9).#3, #

3, F' < *3, /#%6&' <..#<FF)' <F'' :# O%K&6&'A#3(#K%) .5#! , (&' &* (3, /#(%#* &&#K2<(#K%) .5#2<DD&, #39#&6&, #E%' &#F%, (&, (#5<(<#K<*#<55&5A#*) F2#<*#<F(%' *A#<*#%E&#E%63&/%&' *2<6	<6%'3(&#<F(%' *#<, 5#<' &#E%' L&.' '#(%#* &]. E *#! ''#(2%* &#<F(%' *#< ;

V3, <..'A#+<' &*3<, #<, 5#, &)' <.#, &(K%'L#(&F2, 3G) &* *2%) .5#! &#(&*(&5#%, #%(2&' #5<(<* &*(#(2<(#F%, (<3, #E%' <(<#12<(<* &#F2%* &, #K<*#<(2&' *#E<..#5) &#(%#(2&#(3E &#F%ED.&J3(3&*#%9#<#.<' /&' #5<(<* &#(##M(#*2%) .5#! &(&' E3, &5#K2&(2&' #2&#F%, F.) *3%, #(<L&, #9%'E#(23*#* &#(%9#&JD&'3E&, (*#<DD.3&*#(%#.<' /&' #5<(<* &#(##4553(3%, <.#(<' /&#) * &' *#*2%) .5#! &F%, *35&' &5 <*#K&..;

References

4.6<' &#=#: ; 4:A#)\$)3=#- ;A#_<K<(%A#1 ;A#S#_%/&.A#h ;#H@BBCI ;#Z&)' <.#bJD&'#Z&(K%'L*#9%'#V<*(&'#- %E!3, &5#-%..<!%' <(36&#<, 5#-%, (&, (N+<* &5#&F%E E&, 5<(3%, ;#M, #Journal of Computational Methods in Sciences and Engineering, #(%#<DD&<';

+<.<!<, %63FA#> ;A#S#: 2%2<EA#W ;#H?RRUI ;#-%E!3, 3, /#F%, (&, (N+<* &5#<, 5#F%..<!%' <(36&#&' &F%E E&, 5<(3%, ;#M, #Communications of the ACMA#^BH^IA#DD ;#CCNU@;

+3..*) *A#` ;A#S#8<==<, 3A#> ;] ;#H?RRoI ;#X&<' , 3, /#F%..<!%' <(36, 9%'E<(3%, #93. (&' * ;#M, #Proceedings of the Fifteenth International Conference on Machine Learning#DD ;#\CN[\;

+3*2%DA#- ;#> ;#H?RRCI ;#Neural networks for pattern recognition ;#PJ9%'5A#b, /.<, 57#PJ9%'5#Y, 36&' *3(' #8' *#* ;

+ '&*&A#] ;#: ;#O&FL&'E<, A#` ;A#S#_<53&A#- ;#H?RRoI ;#bED3'3F<.#<, <..' *3*#%9#D' &53F(36&#<./% '3(2E *#9%' #F%..<!%' <(36]. (&' 3, / ;#M, #Proceedings of the Fourteenth Conference on Uncertainty in Artificial Intelligence#DD ;#\^g[@ ;

- %D&'A#C ;A#S#O&' *L%63(*A#b ;#H?RR@I ;#4#++<' &*3<, #E&(2%5#9%'#(2, 5) F(3%, #%9#D%' !< 3.3*(3F#, &(K%'L*#9%'E#5<(< ;#M, #Machine Learning#RA#DD ;#^BRN^ \U ;

O<'D&'A#V ;#A#X3A#f ;#A#- 2&, AW ;#S#_%, *(<, A#] ;#H@BB[I ;#4, #bF%, %E3F#>%5&.#%9#Y* &'#\$<(3, /#3, #<, #P, 3, &#&F%E E&, 5&'# : ''*(&E ;#M, #Proceedings of the 10th International Conference on User Modeling ;

] %<F23E *#1 ;#H?RRCI ;#A probabilistic analysis of the Rocchio algorithm with TFIDF for text Categorization#H1- %ED) (&'# : F3&, F#1&F2, 3F<.#\$&D%' (#- > YN- : NRCN??oI ;#-<' , &/3&#>&..% , #Y, 36&' *3(' ;

>3, *L"A#> ;A#S#8<D&' (A# : #H?RCRI ;#Perceptrons ;#- <E! '35/&A#> 47#> M#b#8' *#* ;

>3(F2&..#1 ;#> ;#Machine Learning ;#H?RRUI ; +%*(%, A#> 47#12&#> Fc' <KNO3. #- %ED<, 3&*A#M, F ;

a)3, .<, A#] ;#\$;#H?RoCI ;#M, 5) F(3%, #%9#5&F3*3%, #' (&*& ;#M, #Machine Learning#H?IA#DD ;#o?N?BC ;

a)3, .<, #]#\$;#I?RR^I;#C4.5: *Programs for Machine Learning*;#: <, #><(&%A#- 47#>%'/<, #
_<)9E<, ,;

Appendix

Top-N Precision w/ NN1 Code

3ED%('0<6<;3%;+)99&'&5\$&<5&'u
3ED%('0<6<;3%;+)99&'&5h'3(&'u
3ED%('0<6<;3%;V3.&\$&<5&'u
3ED%('0<6<;3%;V3.&h'3(&'u

3ED%('K&L<;F%'&M, *(<, F&u
3ED%('K&L<;F%'&M, *(<, F&*u
3ED%('K&L<;F.<***93&'*;9), F(3%, *, >).(3.<'&'8&'F&D('% , u
3ED%('K&L<;F.<***93&'*;b6<.)<(3%, u

D)! .3F#F.<***#1%DZ8'&F3*3%, #v

D)! .3F#*(<(3F#6%35#E<3, H: ('3, /wx#<'/*lv

('"v

M, *(<, F&*#5<(<#j #, &K#M, *(<, F&*H
, &K#+)99&'&5\$&<5&'H
#, &K#V3.&\$&<5&'HyF%E! +V;<'99yIIIu

kk*&((3, /#F.<***#<('3!)(&
5<(<*&(-.<***M, 5&JH5<(<:,) E 4(('3!)(&*HI#W#?lu

kk#.%<5#), .<! &.&5#5<(<

>).(3.<'&'8&'F&D('% , #E.D#j #, &K#>).(3.<'&'8&'F&D('% , HIu
E.D;*&(O355&, X<'&'Hy^ylu #kk#*&(#(2ळ&, #, %5&*&
E.D;*&(X&<' , 3, /\$<(&HB;@lu #kk#*&(#(2&#.&<' , 3, /#<(&
E.D;*&(1'<3, 3, /13E&H[BBlu #kk#*&(#(2&#('<3, 3, /#(3E&
E.D;!)3.5-.<***93&'H5<(<lu##kk#!)3.5#F.<***93&'

M, *(<, F&*#), .<! &.&5#j #, &K#M, *(<, F&*H

, &K#+)99&'&5\$&<5&'H
#, &K#

V3.&\$&<5&'HyF%E! +V;<'99yIIIu

kk#*&(#F.<***#<('3!)(&
, .<! &.&5;*&(-.<***M, 5&JH), .<! &.&5;,) E 4(('3!)(&*HI#W#?lu
kk#F'&<(&F%D"

M, *(<, F&*#<! &.&5#j #, &K#M, *(<, F&*H), .<! &.&5lu

kk#.<! &.#3, *(<, F&*#

M, *(<, F&wx#3, *(<, F&*#j #, &K#M, *(<, F&w?Bxu

5%)! .&wx#6<.)&*#j #, &K#5%)! .&w?Bxu

9% 'H3, (#3#j #Bu#3#z#?Bu#3eelv

3, *(<, F&*w3x#j #,)..u

6<.)&*w3x#j #Bu

{

kk'<, L#3, *(<, F&*#W#%, .''#L&&D#(%D#Z

9% #H3, (#3#j #Bu#3#z#), .<! &.&5;,) EM, *(<, F&*HIu#3eelv

##5%)! .&wx#F.*X<! &.#j #E.D;53*('3!)(3%, V%'M, *(<, F&H), .<! &.&5;3, *(<, F&H3lu

```
##39H6<.)&*wBx#zj #F.*X<! &.w?x1#v#kk.3L&#D'#! <! 3.3('
#          9%'H3, (#0#j #?u#0#z#?Bu#0eelv
          39H6<.)&*w0x#zj #F.*X<! &.w?x#SS#0#j #Rlv
          6<.)&*w0N?x#j #6<.)&*w0xu
          3, *(<, F&*w0N?x#j #3, *(<, F&*w0xu
          {
          &.*#39#10#j j #Rlv
          6<.)&*w0N?x#j #6<.)&*w0xu
          3, *(<, F&*w0N?x#j #3, *(<, F&*w0xu
          6<.)&*w0x#j #F.*X<! &.w?xu
          3, *(<, F&*w0x#j #, .<! &.&5;3, *(<, F&H3lu
          {
          &.*&v
          6<.)&*w0N?x#j #F.*X<! &.w?xu
          3, *(<, F&*w0N?x#j #, .<! &.&5;3, *(<, F&H3lu
          !' &<Lu
          {
          {
##{
{
```

```
.<! &.&5;5&.&(&Hlu
9%'H3, (#3#j #Bu#3#z#?Bu#3eel
          39H3, *(<, F&*w3x#j #, )..l
          .<! &.&5;<55H3, *(<, F&*w3xlu
```

```
kk&6<.)<(&#(2&#3, *(<, F&*
b6<.)<(3%, #&6<.#j #, &K#b6<.)<(3%, H.<! &.&5lu
&6<.;F'***T<.35<(&>%5&.H
#####E.DA#.<! &.&5A#?BA#.<! &.&5;/&($<, 5%EZ)E! &'c&, &'<(%'H?llu
: ""(&E;%)(:D'3, (., H&6<.;(%: )EE<'": ('3, /H(')&llu
```

```
kkK'3(&#3, *(<, F&*#H(%#5%)! .#F2&FL#, %#5%)! .&*l
+)99&'&5h'3(&'#K'3(&'#j #, &K#+)99&'&5h'3(&'H
#####, &K#V3.&h'3(&'Hy, , ?;<'99yllu
K'3(&'K'3(&H.<! &.&5;(%: ('3, /Hllu
K'3(&' ;, &KX3, &Hlu
K'3(&'9.)*2Hlu
K'3(&'F.%*&Hlu
```

```
{F<(F2HbJF&D(3%, #&lv
&:D'3, (: (<FL1'<F&Hlu
{
```

```
{
{
```

Top-N Precision w/ NN2 Code

3ED%('#0<6<;3%;+)99&'&5\$&<5&'u
3ED%('#0<6<;3%;+)99&'&5h'3(&'u
3ED%('#0<6<;3%;V3.&\$&<5&'u
3ED%('#0<6<;3%;V3.&h'3(&'u

3ED%('#K&L<;F%'&;M,*(<,F&u
3ED%('#K&L<;F%'&;M,*(<,F&*u
3ED%('#K&L<;F.<***393&'*;9),F(3%,*;>).(3.<'&'8&'F&D('% ,Z&Ku
3ED%('#K&L<;F.<***393&'*;b6<.)<(3%,u

D)! .3F#F.<***#1%DZ8'&F3*3%,@#v

D)! .3F#*(<(3F#6%35#E<3,H:('3, /wx#<'/*lv

('"v

M,*(<,F&*#5<(<#j #,&K#M,*(<,F&*H
,&K#+)99&'&5\$&<5&'H
#,&K#V3.&\$&<5&'HyF%E! +V;<'99y11lu
kk#&((3, /#F.<***#<('3!)(&
5<(<*&(-.<***M,5&JH5<(<:,)E4('3!)(&*H1#H#?lu

kk#.%<5#), .<! &.&5#5<(<

>).(3.<'&'8&'F&D('% ,Z&K#E.D#j #,&K#>).(3.<'&'8&'F&D('% ,Z&KHlu
E.D;*&(O355&,X<'&'*Hy^ylu #kk#&#(2ळ&, #,%5&*&
E.D;*&(X&<' ,3, /\$<(&HB;@lu #kk#&#(2&#&<' ,3, /#<(&
E.D;*&(1'<3,3, /13E&H[BBlu #kk#&#(2&#('<3,3, /#(3E&
E.D;!)3.5- .<***393&'H5<(<lu##kk#!)3.5#F.<***393&'

M,*(<,F&*#), .<! &.&5#j #,&K#M,*(<,F&*H

,&K#+)99&'&5\$&<5&'H
#,&K#

V3.&\$&<5&'HyF%E! +V;<'99y11lu

kk#&#(F.<***#<('3!)(&
, .<! &.&5;*&(-.<***M,5&JH), .<! &.&5;,)E4('3!)(&*H1#H#?lu
kk#F'&<(&#F%D"

M,*(<,F&*#<! &.&5#j #,&K#M,*(<,F&*H), .<! &.&5lu

kk#.<! &.#3,*(<,F&*#

M,*(<,F&wx#3,*(<,F&*#j #,&K#M,*(<,F&w?Bxu
5%)! .&wx#6<.)&*#j #,&K#5%)! .&w?Bxu

9%#H3, (#3#j #Bu#3#z#?Bu#3eelv
3,*(<,F&*w3x#j #,)..u
6<.)&*w3x#j #Bu

{

kk'<,L#3,*(<,F&*#N#%, .''#L&&D#(%D#Z

9%#H3, (#3#j #Bu#3#z#), .<! &.&5;,)EM,*(<,F&*Hlu#3eel#v

##5%)! .&wx#F.*X<! &.#j #E.D;53*('3!)(3%,V%'M,*(<,F&H), .<! &.&5;3,*(<,F&H31lu
##39H6<.)&*wBx#zj #F.*X<! &.w?x1#v#kk.3L&#D'#! <1.3.3('

9%#H3, (#0#j #?u#0#z#?Bu#0eelv

39H6<.)&*w0x#zj #F.*X<! &.w?x#SS#0#j #Rlv

6<.)&*w0N?x#j #6<.)&*w0xu
3, *(, F&*w0N?x#j #3, *(, F&*w0xu

{
&.*#39#H0#j j #Rlv
6<.)&*w0N?x#j #6<.)&*w0xu
3, *(, F&*w0N?x#j #3, *(, F&*w0xu
6<.)&*w0x#j #F.*X<! &.w?xu
3, *(, F&*w0x#j #), .<! &.&5;3, *(, F&H3lu
{
&.*&v
6<.)&*w0N?x#j #F.*X<! &.w?xu
3, *(, F&*w0N?x#j #), .<! &.&5;3, *(, F&H3lu
!'&<Lu

{
##{
{

.<! &.&5;5&.&(&Hlu
9%'H3, (#3#j #Bu#3#z#?Bu#3eel
.<! &.&5;<55H3, *(, F&*w3xlu

kk&6<.)<(&#(2, *(, F&*
b6<.)<(3%, #&6<.#j #, &K#b6<.)<(3%, H.<! &.&5lu
&6<.;F'***T<.35<(&>%5&.H
#####E.D#.<! &.&5A#?B#.<! &.&5;/&(\$<, 5%EZ)E!'&c&, &'<(%'H?ll
: ""(&E.%)<D'3, (.H&6<.(%:)EE<'": ('3, /H(')&ll

kkK'3(, *(, F&*#H(%#5%)! .&#F2&FL#, %#5%)! .&*l
)99&'&5h'3(&'#K'3(&'#j #, &K#+)99&'&5h'3(&'H
#####, &K#V3.&h'3(&'Hy, , @,<'99yll
K'3(&'K'3(&H.<! &.&5;(%: ('3, /Hll
K'3(&'&KX3, &Hlu
K'3(&'9.)*2Hlu
K'3(&'F.%*&Hlu

{F<(F2HbJF&D(3%, #&lv
&D'3, (: (<FL1'<F&Hlu

{

{

{

Top-N Precision w/ BN Code

3ED%'(#0<6<;3%;+)99&'&5\$&<5&'u

3ED%'(#0<6<;3%;+)99&'&5h'3(&'u

3ED%'(#0<6<;3%;V3.&\$&<5&'u

3ED%'(#0<6<;3%;V3.&h'3(&'u

3ED%'(#K&L<;F%'&;M,*(,<,F&u

3ED%'(#K&L<;F%'&;M,*(,<,F&*u

3ED%'(#K&L<;F.<***393&'*;!<"&*;+<"&*Z&(u

3ED%'(#K&L<;F.<***393&'*;!<"&*;,&(*&<'F2;/.%!.<;mu

3ED%'(#K&L<;F.<***393&'*;b6<.)<(3%,u

D)! .3F#F.<***#1%DZ8'&F3*3%,^#v

D)! .3F#*(<(3F#6%35#E<3,H:('3,/wx#<'/*lv

('"v

M,*(,<,F&*#5<(<#j #,&K#M,*(,<,F&*H

,&K#+)99&'&5\$&<5&'H

#,&K#V3.&\$&<5&'HyF%E!+V;<'99y11lu

kk*&((3,/F#<***#<('3!)&

5<(<*&(-.<***M,5&JH5<(<;,)E4(('3!)&*H1#W#?lu

kk#.%<5#),.<!&.&5#5<(<

+<"&*Z&(#E.D#j #,&K#+<"&*Z&(Hlu

@#L@#j #,&K#@Hlu

L@:*&(><JZ'P98<'&,(H?lu kk#&(#E<J#}#9#D<'&,(*

E.D;*&(:&<'F24./%'3(2EHL@lu#kk#&(#*&<'F2#<./%'3(2E

E.D;!)3.5-.<***393&'H5<(<lu##kk#!)3.5#F.<***393&'

M,*(,<,F&*#),.<!&.&5#j #,&K#M,*(,<,F&*H

,&K#+)99&'&5\$&<5&'H

#,&K#

V3.&\$&<5&'HyF%E!+V;<'99y11lu

kk#*&(#F.<***#<('3!)&

),.<!&.&5;*&(-.<***M,5&JH),.<!&.&5;,)E4(('3!)&*H1#W#?lu

kk#F'&<(&F%D"

M,*(,<,F&*#<!&.&5#j #,&K#M,*(,<,F&*H),.<!&.&5lu

kk#<!&.#3,*(,<,F&*#

M,*(,<,F&wx#3,*(,<,F&*#j #,&K#M,*(,<,F&w?Bxu

5%)!.&wx#6<.)&*#j #,&K#5%)!.&w?Bxu

9%'H3,(#3#j #Bu#3#z#?Bu#3eelv

3,*(,<,F&*w3x#j #,)...u

6<.)&*w3x#j #Bu

{

kk'<,L#3,*(,<,F&*#W#%,.'#L&&D#(%D#Z

9%'#H3,(#3#j #Bu#3#z#),.<!&.&5;,)EM,*(,<,F&*Hlu#3eel#v

#5%)!.&#wF.*X<!&.#j #E.D;53*('3!)(3%,V%'M,*(,<,F&H),.<!&.&5;3,*(,<,F&H31lu

```
##39H6<.)&*wBx#zj #F.*X<! &.w?x1#v#kk.3L&#D'#! <! 3.3('
#          9%'H3, (#0#j #?u#0#z#?Bu#0eelv
          39H6<.)&*w0x#zj #F.*X<! &.w?x#SS#0#j #Rlv
          6<.)&*w0N?x#j #6<.)&*w0xu
          3, *(<, F&*w0N?x#j #3, *(<, F&*w0xu
          {
          &.*#39#H0#j j #Rlv
          6<.)&*w0N?x#j #6<.)&*w0xu
          3, *(<, F&*w0N?x#j #3, *(<, F&*w0xu
          6<.)&*w0x#j #F.*X<! &.w?xu
          3, *(<, F&*w0x#j #), .<! &.&5;3, *(<, F&H3lu
          {
          &.*&v
          6<.)&*w0N?x#j #F.*X<! &.w?xu
          3, *(<, F&*w0N?x#j #), .<! &.&5;3, *(<, F&H3lu
          !' &<Lu
          {
          {
```

```
##{
{
```

```
.<! &.&5;5&.&(&Hlu
9%'H3, (#3#j #Bu#3#z#?Bu#3eel
.<! &.&5;<55H3, *(<, F&*w3xlu
```

```
kk&6<.)<(&#(2&#3, *(<, F&*
b6<.)<(3%, #&6<.#j #, &K#b6<.)<(3%, H.<! &.&5lu
&6<.;F'***T<.35<(&>%5&.H
#####E.DA#.<! &.&5A#?BA#.<! &.&5;/&($<, 5%EZ)E!&'c&, &'<(%'H?llu
: ""*(&E.%)(:D'3, (., H&6<.;(%: )EE<'": ('3, /H(')&llu
```

```
kkK'3(&#3, *(<, F&*#H(%#5%)! .#F2&FL#, %#5%)! .&*l
+)99&'&5h'3(&'#K'3(&'#j #, &K#+)99&'&5h'3(&'H
#####, &K#V3.&h'3(&'Hy! , ;<'99yl lu
K'3(&' ;K'3(&H.<! &.&5;(%: ('3, /Hllu
K'3(&' ;, &KX3, &Hlu
K'3(&' ;9.)*2Hlu
K'3(&' ;F.%*&Hlu
```

```
{F<(F2HbJF&D(3%, #&lv
&:D'3, (: (<FL1'<F&Hlu
{
```

```
{
{
```