

Bayesian and Neural Networks for Motion Picture Recommendation

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+ % * (% , # - % . & / # 0 % , % ' * # 1 2 & * 3 *
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)#!<*#5%, #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(&#%#%(2&'#3(&#<'&#F%ED<'&5# </<3, *(#(2%*&#%9#*3E3.<'#3(E*#(2<(#(2&#)*&'#2<#*!&&, #3, #F%, (<F(#K3(2;##12&*#&#*3E3.<'#3(E*# <'&.D&5#(%#/<)/&#K2&(2&'#(23*#)*&'#K3.#.3L&#%#*#53*.3L&#(2(&#:#: %E&#&F%E&E, 5&'# *''*(E*#)*&#<#E3J)'&#%9#F%..<'%<(36&#<, 5#F%, (&, (N!<*#5#<DD'<F2&#;

\$&F%E&E, 5&'#*''*(E*#2<6&#!&F%E&#<#D%D).<'#*)!0&F(#<'(3, /#3, #(2&#.<(&#?RRB0* H\$&*, 3FL#S#T<'3<, #?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(#3, #<553(3%, #(%# (2&#%, &#(2&#F)*(%E&'#D)'F2<*&5#(2&#.<*(#(3E&#%#*2?*3(&5;##+'#2<63, /#(23*#('D&#%9# *''*(E*#3, #D.<F&#&'<3.&'*#<'&#<!&#(%#<..%K#F)*(%E&'#(%#E%&#&<*3.'#93, 5#D'5)F*(2<(# (2&'#E<'#!, (&'&*(5, #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&'#!<*#5%, #(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*#*)F2#)*&'#<'&#<!&#(%#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*#%#93, 5#<#!&((&'#<./%3(2E#K23F2#F%#E!3, &*#F%, (&, (N!<*#5#5<(<

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

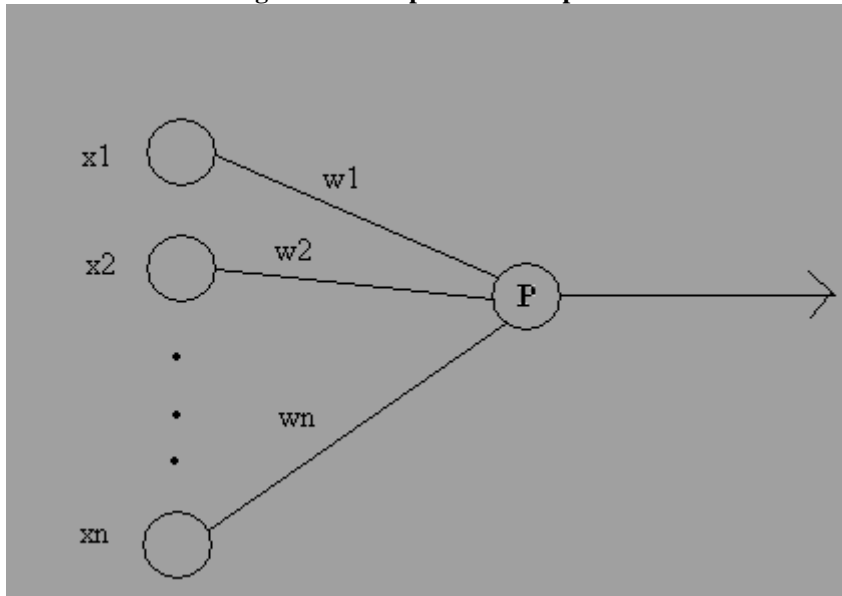
4, %(2&'#F.<***393&'#K23F2#!)3.5*#)D%, #').&#F.<***393&'*#3*#%, &#K3(2#5&F3*3%, #'&'&*&#;##
`&F3*3%, #'&'&*&#!<*3F<..'#(<L&#<#.%%L#<(#E).(3D.&#<((('3!))&#<#<, 5#!'&<L#(2&E#3, (%#*#D<'<(&#
D<'3, /*#D'&53F(3, /*#5399&'&, (#6<.)&*#9%'#(2&#(<'&#<((('3!))&#!<*#5#%, #2&#D<'3, /*#9#%(2&#
6<'3<!&*&#;#Z&K#!'<, F2&#*#<'&#E<5	%'#&, &K#D<'3, /*#<, 5#<*#<#&'&#*#).(#<, #&, (3'&#('*#E<5&#&#
K23F2#(2&#F.<***393&'#!'<, F23, /*#<(#&<F2#D%3, (#5&D&, 53, /*#%, #K2<#(2<.)&#%9#<#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<..'#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#('&#A#%, &#(2<(#F%).5#D'&53F(#9<.*#&
) (F%E*#5)&#(%#*#E, *(<, F&#*#K2*#&#(<'&#<((('3!))<.)&#K%).5#!ᔗ&'&, (#2<, #
%(2&'#3, *(<, F&#*#K3(2*#3E3.<#6<.)&*#9%'#3*#<((('3!))*#<#&'&#*#).(#, &K#<./%'3(2E*#K&'&#&
E<5&#&#)*3, /*#D'), 3, /*#8'), 3, /*#K%).5#&E#6&#&#!'<, F2&#*#9#(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&#%9#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#<***3/, *#*#E&#
F%, *(<, (#F<..&5#(%#&<F2#%, &#F<..&5#<#K&3/2(##12&#K&3/2(#&D'&#&, (*#(2ED%'<(<, F&#%9#(2&#
3, D)(#%#(2&#D&'F&D('%, ;##M, #2&#E<F23, &#&<'3, /*#F<*#&#(2, D)(*#<'&#(2ᔗ&'&, (#6<.)&*#%9#
(2&#<((('3!))&#*#%9#<, #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/)'&#*#<#53</'<E#%9#<#D&'F&D('%, ;##

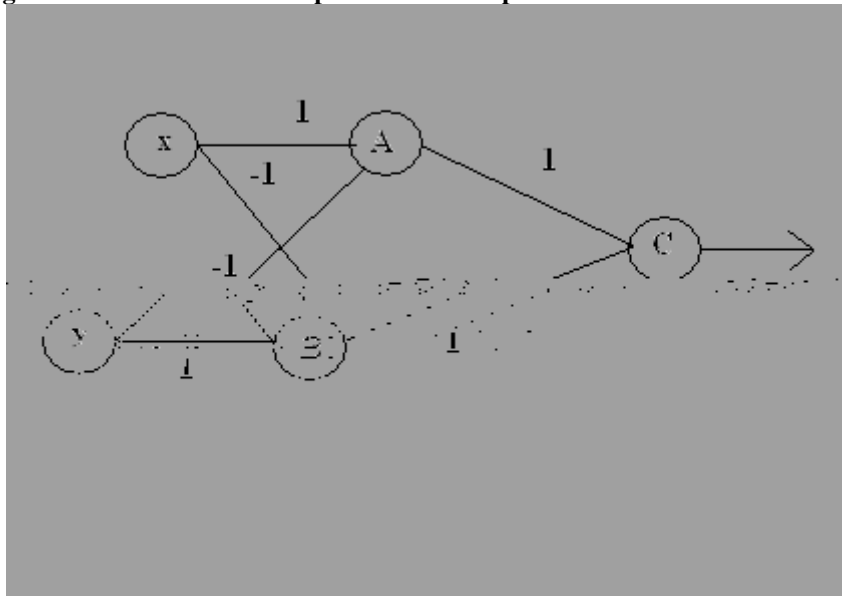
Formula: c36&, #3, D)(*#J?#J_#d #J, #<#D&'F&D('%, #K3.#<***3/, #K&3/2(*#K?#K_#d #K, #(%#
&<F2#%9#(2&E;##M(#K3.#(2&, #F<F).<(&#J?K?#eJ_#e#d #eJ, K, #1#<, 5#(2&, #%) (D)(#<#*3, /.&#
6<.)&#!<*#5#%, #K2&(2&'#%#&, %(#2<(#6<.)*#/'&<(&'#2<, #B;

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 4A#+#<, 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.<***&'<', &<'.'''*#D<'<!.&#H>3, *L''#S#8<D&'(A#?RC[I;#O%K&6&'A#
E%*(#5<(<*&'5%# , %2<6, &<'.'''*#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_L_2#5&, %(&*#(2&#K&3/2(#9%'E# , %5#(%# , %5&#L;#12&#(&'E# 'k_L_2_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&*#5399&'&, (#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'#! <! 3.3(''3*#3, *(&<5#F%ED) (&5A#<, 5#(2<(#3*#) *&5#3, #
 5&(&'E3, 3, /#(2&#%) (D) (#<((!3!)) (<.)&#;##M, #(&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'#! <! 3.3('';#
 T<'3%) *#K%'L#2<#*! &&, #5%, &#) *3, /#(23*#E%5393&5#9%'E#<#K&..#H+3*2%D#?RRCI

Bayesian Techniques

4, %(2&'#("D&#%9#F.<*393&'#) *&#*#<+'*3<, #&<*, 3, /;##<+'*3<, #(&F2, 3G) &#*#<'&#
 !<*5#%, #D'#! <! 3.3(''53*('3!)) (3%, *#<, 5#(2<(#) *3, /#(2&*&#D'#! <! 3.3(3&*#%, #'! *'6&5#5<(<#F<, #
 3ED'%6&#D&'9%'E<, F&#;##M#('3&*(%#D'%5) F&#(2&#! *#(2''D%(2&*3*#9%'E#*%E&#*D<F&#%9#
 2''D%(2&*#O#/#36&, #*%E&#('<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'#! <! 3.3(''
 <.'&<5''F<.F) .<(&5A#F<..&5#8H2I;##8H` 1#3*#(2&#D'3%'#D'#! <! 3.3(''2<(#(2&#('<3, 3, /#5<(<#` #K3..#
 !&#%!' *'6&5;##8H` i21#3*#(2&#D'#! <! 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'#! <! 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

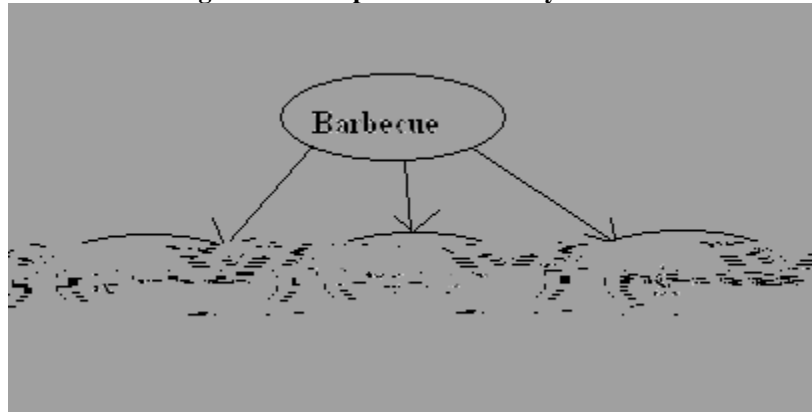
Formula: $8H2i` 1#j \#8H` i218H2I k 8H` I$

<+'*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` i218H2I#3*#F<.F) .<(&5;##12&, %E3, <(%'#8H` 1#3, #
 <+'*0#9%'E) .<#3*#&E%6&5#! &F<)*#(2&#('<3, 3, /#5<(<#` #, &6&'#F2<, /	%'#&<F2#F2<, /&#%9#
 2#3, #O;##12''D%(2&*3*#2#K3(2#(2/2*#6<.)&#%9#8H` i218H2I#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&#<+'*#;##M, #Z<16&#<+'*A#(2&#<./%'3(2E#F'&<(&#*#<#&(#%9#<..#D%*3! .&#
 (<' /&#<((!3!)) (&#;##M#(2&, #F<.F) .<(&*(2&#D'#! <! 3.3(''(&'E*#8H2I#<, 5#8H` i21#<#*#(<(&5#<'%6&;#
 O%K&6&'A#3(#! '&<L*#&<F2#5399&'&, (#<((!3!)) (<.)#%9#(2&#('<3, 3, /#5<(<#` #<, 5#F<.F) .<(&*#
 &<F2#%, &#%9#(2&E#*#D'<(&.';##M#(2&, #(<L&*#(2&#E<J3E) E#6<.)&#%9#8H2I#E) .(3D.3&5#! ''#(2&#
 D'%5) F(#%9#<..#(2&#D'#! <! 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''&*1#m#8H%) (i''&*1#m#8H, %, &i''&*1#m#8H2%(i''&*1#<, 5#8H, %1#m#
8H%) (i, %1#m#8H, %, &i, %1#m#8H2%(i, %1#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

$Z \perp\!\!\!\perp \{A, B\} \mid \{X, Y\}$

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 $Z_1 \dots Z_m$

$M, \#(2^3 \# K \lt \text{"A"} \# \#, \&(K \% 'L \# 3^* \# F' \& \lt (\& 5 \# \text{"} \# F \% , \& F(3, / \# \lt ((\text{'} 3!) (\&^* \# (2 \lt (\# \lt \& \#, \% F \% , 53(3 \% , \lt \dots \text{"} \# 3, 5 \& D \& , 5 \& , (\# \% 9 \# \% , \& \# \lt , \% (2 \& \# \# \lt , 5 \# F \lt F) . \lt (3, / \# (2 \& 3^* \# F \% , 53(3 \% , \lt \# D' \% ! \lt ! 3.3(\text{"} \# 12 \& \# \& 9 \% \& \# \lt \#^* \text{"} \# (\& E \# \% 9 \#, \% 5 \& \# 3^* \# F' \& \lt (\& 5 \# 3 E 3 . \lt \# (\% \# (2 \& \#^* \text{"} \# (\& E \# F' \& \lt (\& 5 \# \text{"} \# , \&) \text{'} \lt \# . \& \&(K \% 'L \# ; \# M, \#(2^3 \# 3) \lt (3 \% , \text{A} \# \%) \# F \lt , \# 3, 9 \& \# (\text{'} / \& (\# 6 \lt .) \&^* \# 9 \% \# \lt ((\text{'} 3!) (\&^* \# \text{"} \# F \lt F) . \lt (3, / \# (2 \& \# D' 3 \% \# D' \% ! \lt ! 3.3(\text{"} \# \% 9 \# \& \lt F 2 \# 6 \lt .) \& \# / 36 \& , \# (2 \& \# D \lt \& , (\text{*} \# \% 9 \# (2 \lt (\# , \% 5 \& ; \# M, \#(2^3 \# K \lt \text{"} \# \%) \# F \lt , \# F \% , \& F(\# , \% 5 \& \# (\% / \& (2 \& \# \# \lt , 5 \# 2 \lt 6 \& \# \# \# , \&(K \% 'L \# \% 9 \#, \% 5 \& \# 2 \lt 6 3, / \# E \lt , \text{"} \# D \lt \& , (\text{*} ; \# 4, \# \& J \lt E D . \& \# \% 9 \# \# \# \lt \text{"} \&^* 3 \lt , \# , \&(K \% 'L \# 3^* \# 2 \% K, \# 3, \# \sqrt{3} /) \text{'} \&^* \# [\# \lt , 5 \# [! ;$

Figure 5a: Example of a Bayesian Network

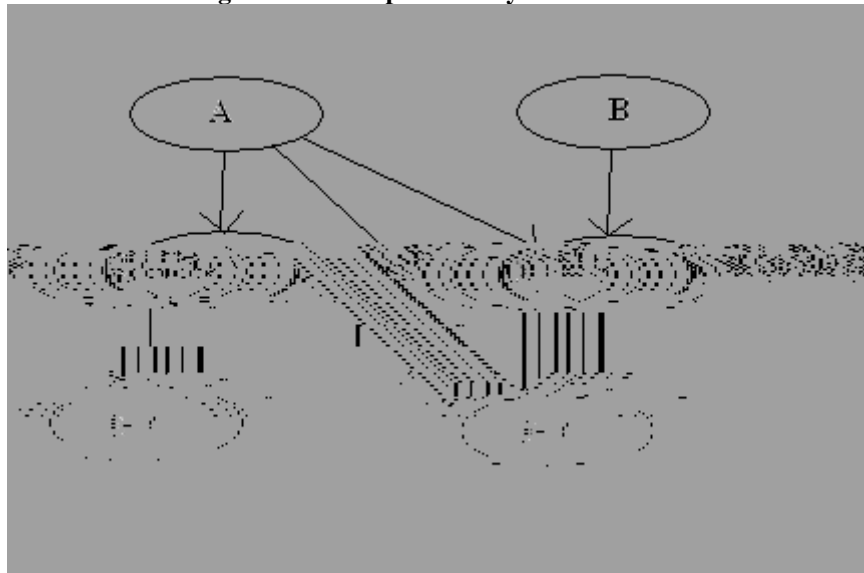


Figure 5a: Example of a Conditional Probability Table

	A=5	A≠5	A=5	A≠5
9	B; \	B; ?	B; 0	B; @
n9	B; C	B; R	B; @	B; 0

$M, \#(2^3 \# J \lt E D . \& \# - \# 2 \lt \# D \lt \& , (\# \text{ } \# b \# 2 \lt \# D \lt \& , (\# - \text{A} \# \# 2 \lt \# D \lt \& , (\text{*} \# 4 \# \lt , 5 \# + \text{A} \# \lt , 5 \# \text{V} \# (2 \& \# (\text{'} / \& (\# \lt ((\text{'} 3!) (\& \text{A} \# 2 \lt \# D \lt \& , (\text{*} \# 4 \# \lt , 5 \# \text{ } ; \# M(\# 3^* \#, \% (\# 0 \% 3, \& 5 \# 3, \# \lt \text{"} \&^* \# \lt \# 3, \# , \&) \text{'} \lt \# . \&(K \% 'L \# \text{A} \# \lt \# (2 \& \# D \lt \& , (\text{*} \# \% 9 \#, \% 5 \& \# F \lt , \# \lt . \text{*} \# ! \& \# (2 \& \# D \lt \& , (\text{*} \# \% 9 \# (2 \& 3^* \# F 23.5 \text{' } \& ; \# 12 \& \# 9 \% \& \# F \lt F) . \lt (3, / \# (2 \& \# D' \% ! \lt ! 3.3(\text{"} \# \% 9 \# \text{V} \# ! \& 3, / \# 9 \# / 36 \& , \# (2 \& \# 5 \lt (\# (2 \lt (\# 4 \# 3^* \# \# \lt , 5 \# \text{ } \# 3^* \# 5 \# K \%) . 5 \# ! \& \text{A} \# 9 \% \text{' } \# J \lt E D . \& \#$

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%E E&, 5<(3%, *#(%#
.3*(&, &'*#H: 2<'5<, <, 5#S#><&*A#?RR[I;##M, #*)F2A#&F%E E&, 5<(3%, *#<'&#D<*&5#(2'%) /2#
pK%'5#%9#E%) (2qA#<(2&'#(2<, #F%, (&, (A#*%#E)*3F#'&F%E E&, 5<(3%, *#E<''#! '<*(3F<..'#
5399&'&, (#(2<, #(2%*&#(2<(#<#)*&'#2<*.3*(&, &5#(%#! &9%'&'<6&#! &&, #E<, ''#%(2&'#(&*(#

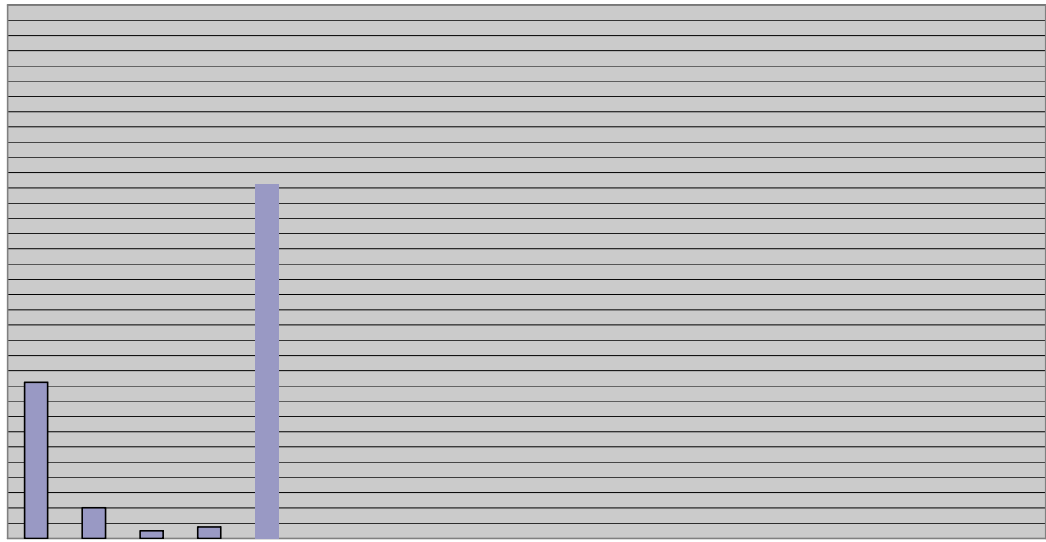
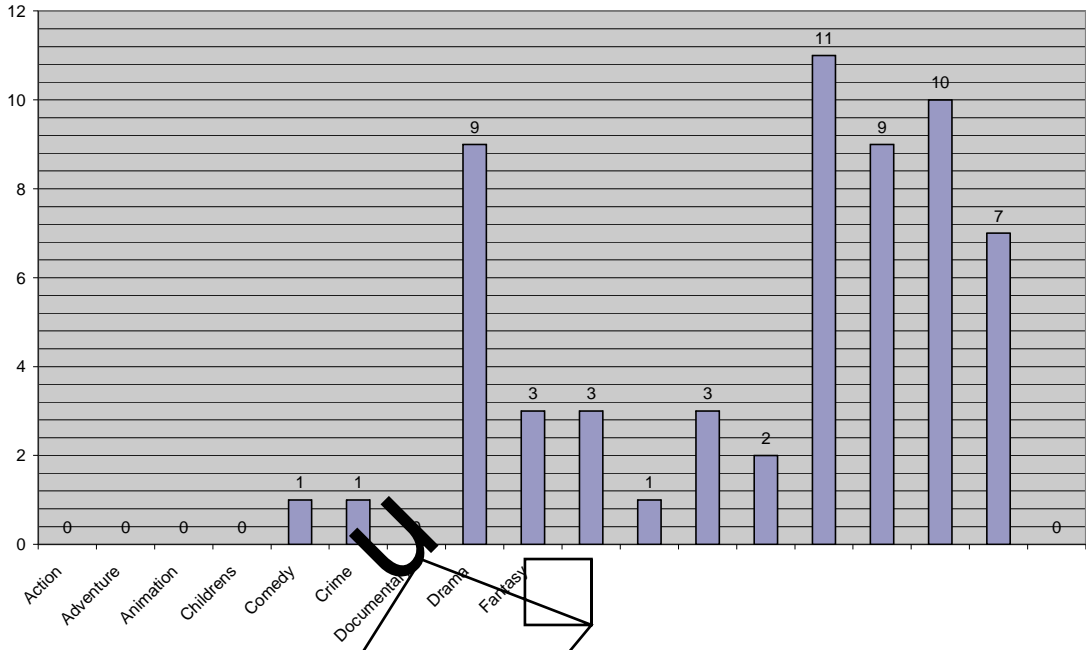
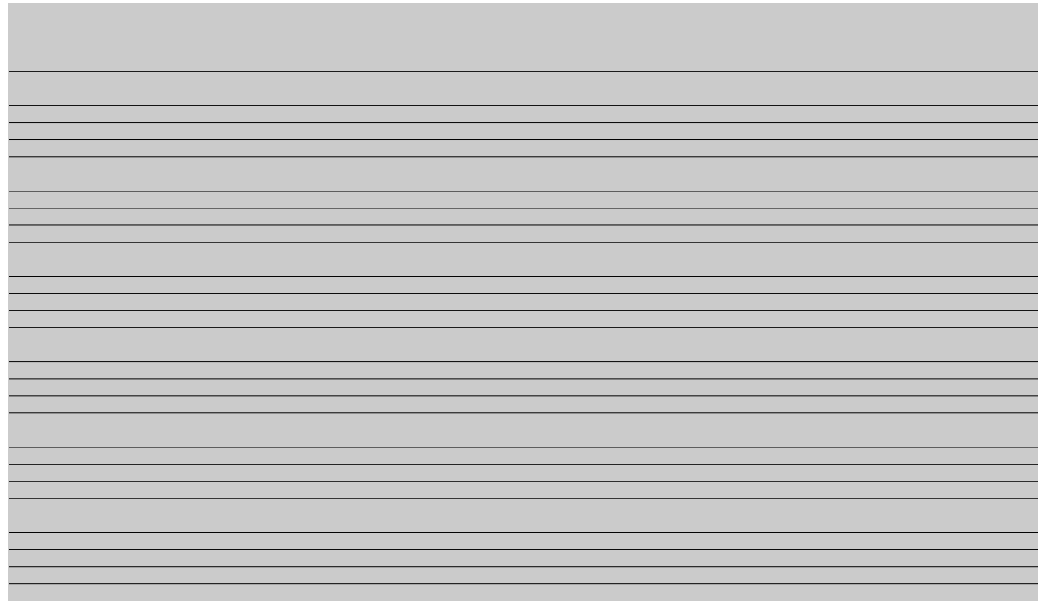


Figure 7c: Genre3 Distribution





Performance Metrics and Evaluation Protocol

$$Accuracy = \frac{TP + TN}{TP + FP + FN + TN}$$

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.

$$Precision = \frac{TP}{TP + FP}$$

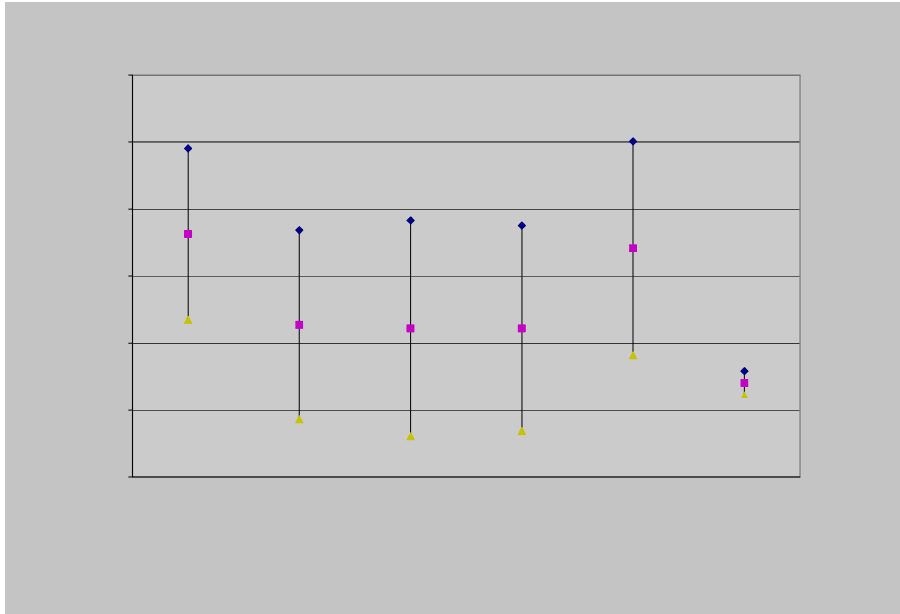
Results

Attribute Selection

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

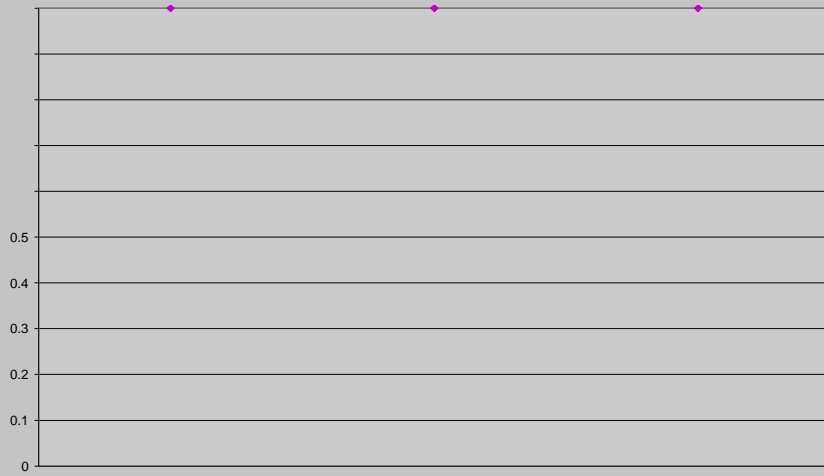
Principal Components

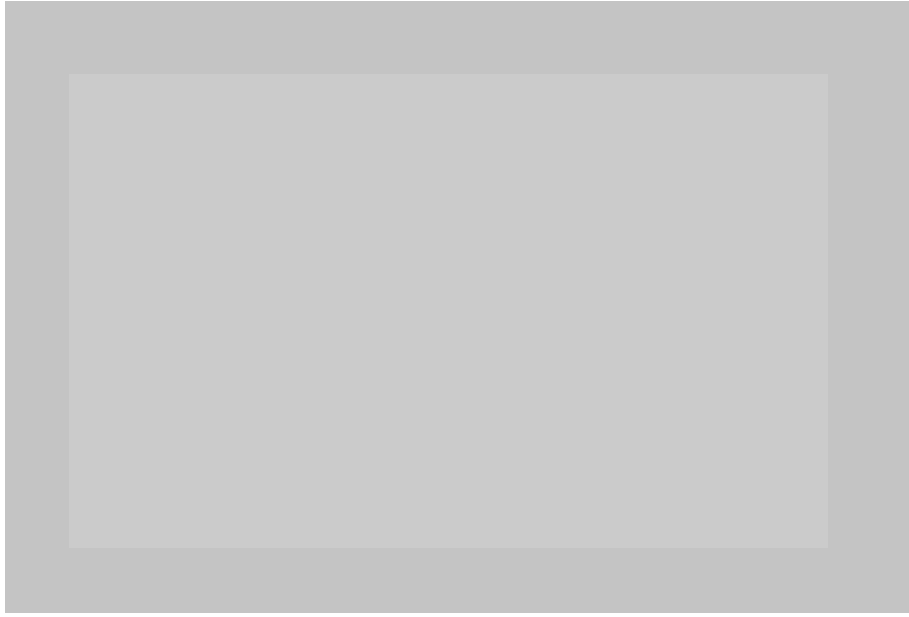
Mathematical symbols and characters: $V3/)' \&^*\#0<\#0! \#<, 5\#oF\#^*2\%K\#(2\&\#D\&'9\%'E<, F\&\#9\#(2\&\#5399\&' \&, (\#E<F23, \&\#.\&<', 3, /#\<./\%'3(2E^*\# \#6\&' \#5<(<\#(2<(\#K<^*\#D'\&D'\%F\&^*\&5\#)^*3, / \#D'3, F3D<.\#F\%ED\%, \&, (*\#<((('3!)(\&^*\&.\&F(3\%, ;$



BestFirst

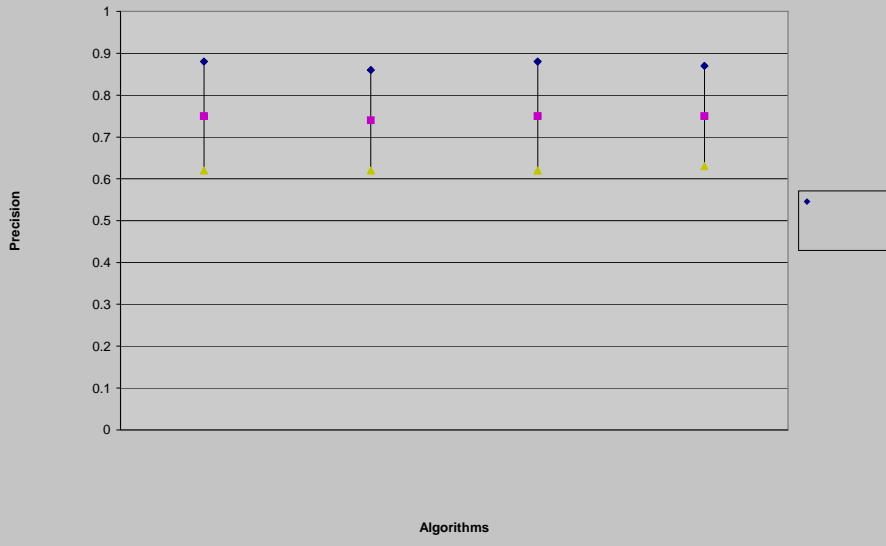
Figure 9c: Top-10 Precision Using BestFirst Attribute Selection





Neural Networks

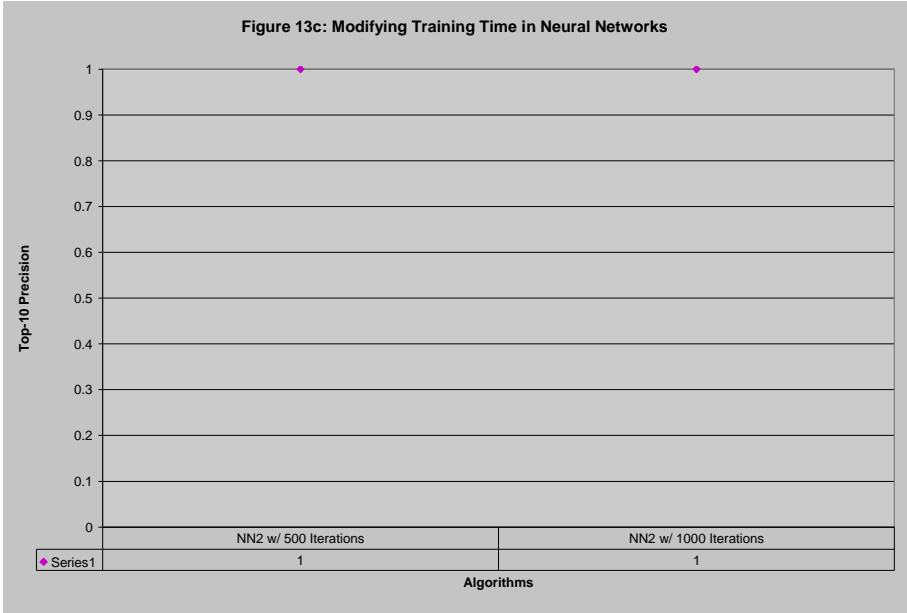
Figure 12b: Modifying Hidden Nodes in Neural Networks



Number of Training Epochs

W(2&, #5&F35&5#(%#3, F'&<*&#(2&#*(<, 5<'5#(' <3, 3, /#(3E&#%, #(2<(<#9'%E#[BB#
3(&'<(3%, *#(%#?BBB#3(&'<(3%, *;#12&#' &*) .(*#<'&#*2%K, #3, #V3/) '& *#?^<A ?^! #<, 5#?^F;

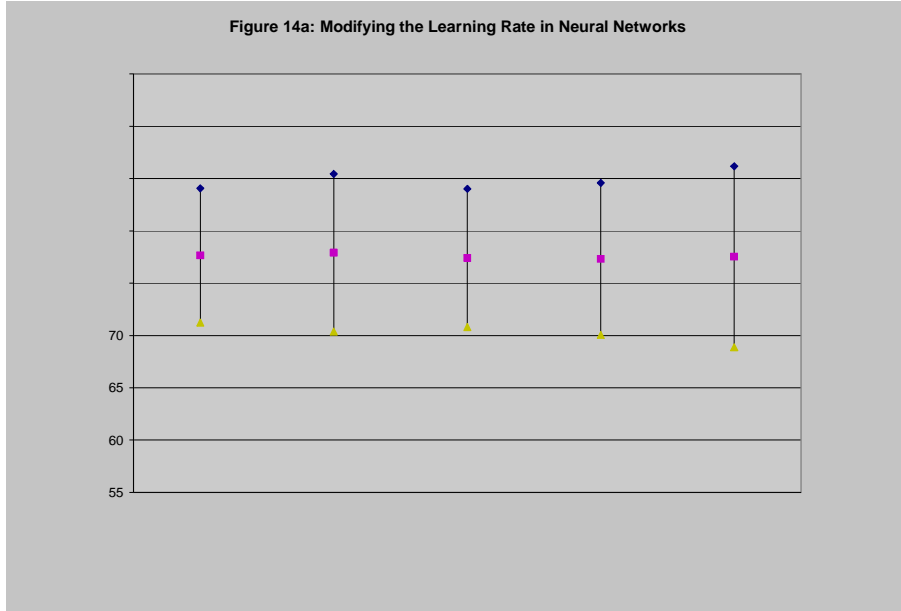




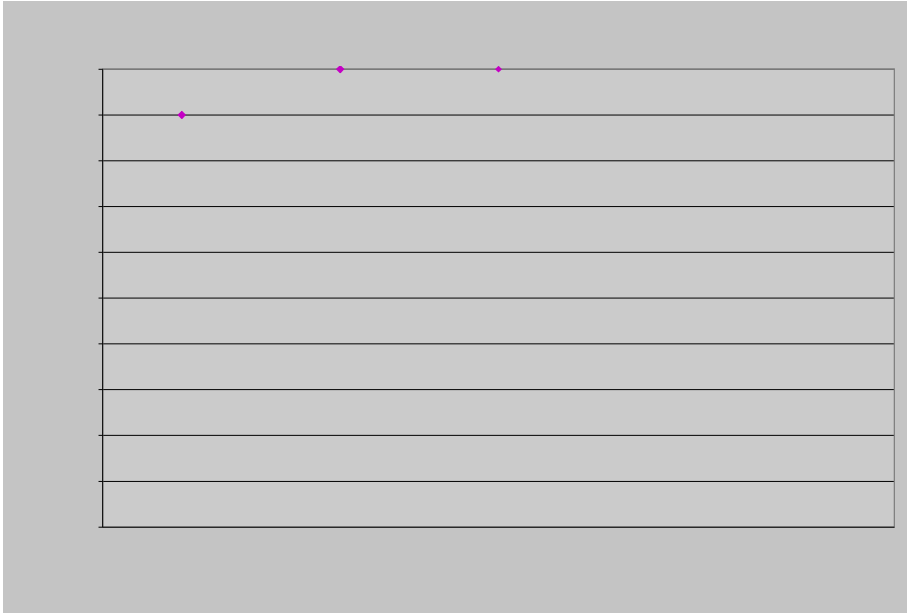
X%%L3, /#<(#(2&# ' & *).(*A#(%DN?B#D' &F3*3%, #K<*(2&#* <E&*D3(, F' &<*3, /#(2&# ('<3, 3, /#(3E&; #4FF)'<F''#.3/2(''5&F'&<* &5A#<, 5#(2&#*(<, 5<'5#5&63<(3%, #*.3/2(''3, F' &<* &5;# 8'&F3*3%, #<. *%#*.3/2(''5&F'&<* &5;##1 2&' &9%'&A#5&*D3(, F' &<*3, /#(2&#('<3, 3, /#(3E &A#(2&# <FF)'<F''#<, 5#*(<, 5<'5#5&63<(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 is often found through trial and error, and can vary significantly between different models and datasets.



70



Bayesian Networks

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

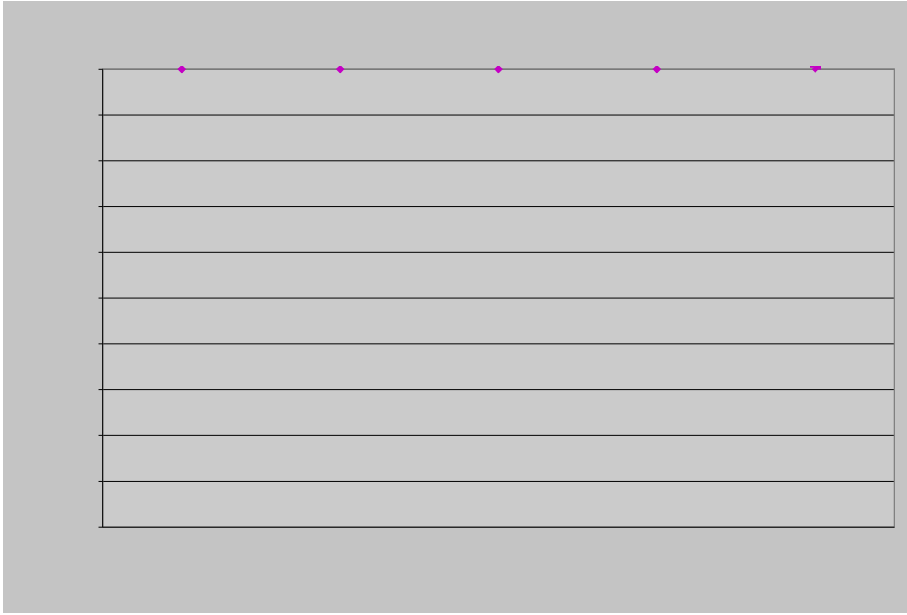
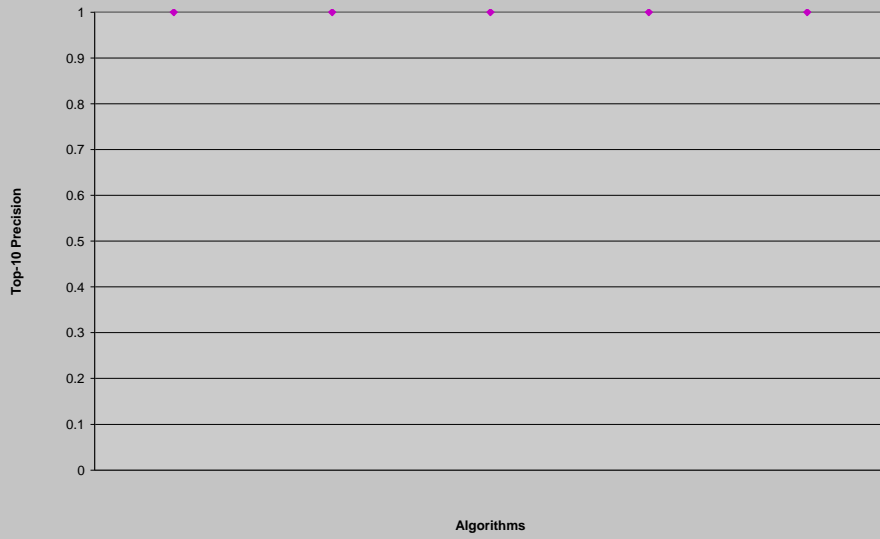
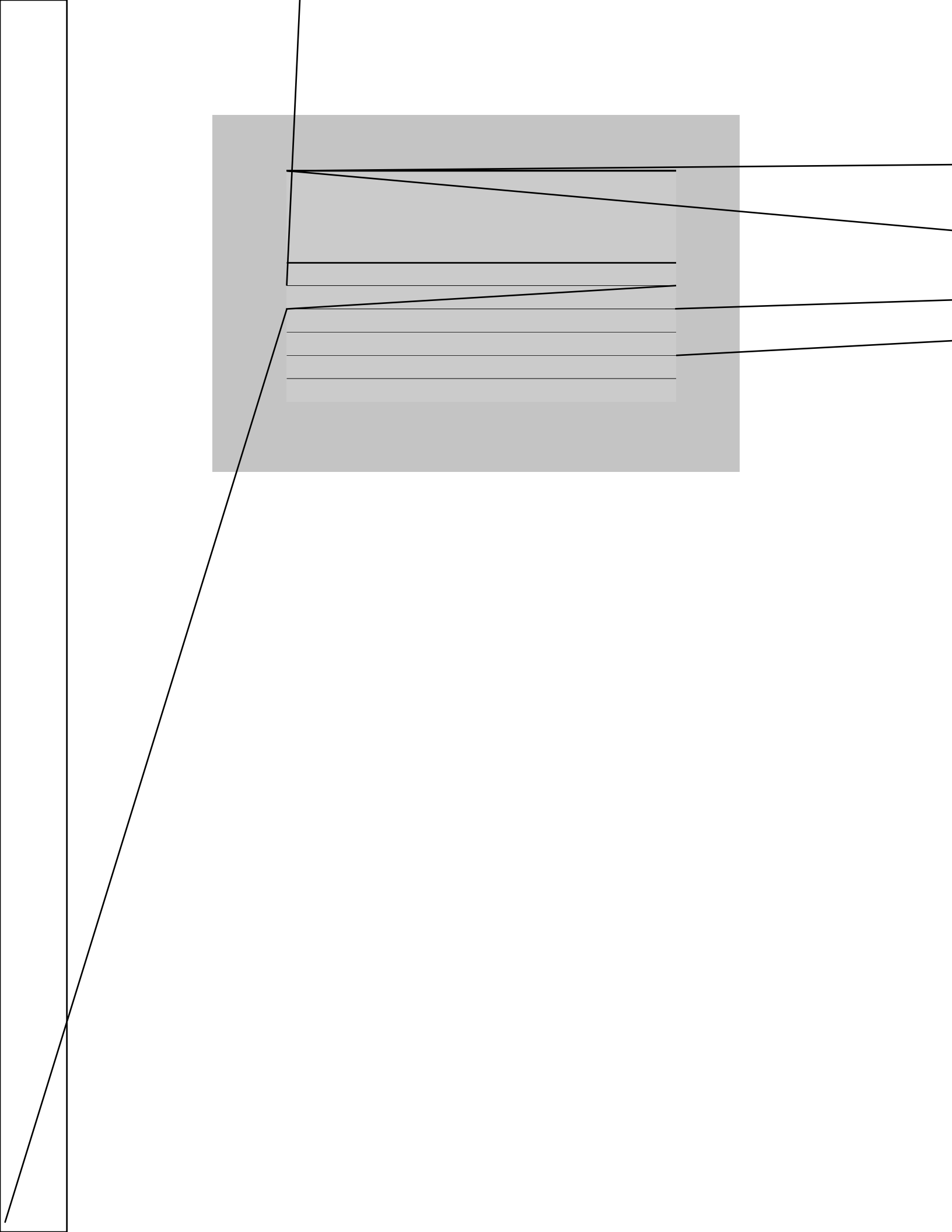


Figure 16c: Varying the Number of Parents in Bayesian Networks (Collaborative-Only)





Enhanced Content Data



Figure 18e: Varying the Number of Parents in Bayesian Networks (More Content)



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<, #!&#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&5I;##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<I6&#'+<'&*#?#&#<, 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&*#?BBR #(%DN?B#D'&F3*3%, #<, 5#3, F'&<'&*#(2&# E<J3E)E#<FF)'<F"##9%'E#(2&#B\$#!<*&3, &##+&'(V3*'#(##3*#	<*&#'(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&#&'5%&*#%, #(<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"##D'&F3*3%, #<, 5#(%DN?B#D'&F3*3%, ;##M#<DD&<'*#(2<#<50)*3, /#(2&#<'<3, 3, /#(3E&#A#&'<'3, /#<(&#<, 5#%(2<'3<!'%&*#%, %# 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&#?BBR #(%DN?B#D'&F3*3%, #K23F2#3*#<# 6&'#3ED%'<,(#(<(3*(3F;#12&'&', 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&#?BBR #(%DN?B#D'&F3*3%, ;## ?BBR #(%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%EE&, 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%EE&, 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 ;#X3A#f ;#A# - 2&, AW ;#S#_ %, *(<, A#] ;#H@BB [I ;#4, #bF%, %E3F#> %5&.#%9#Y* &' #\$(3, /#3, #
<, #P, 3, &#&F%EE&, 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#> Mb#8' &*#* ;

>3(F2&..A#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<(<:,)E4(('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;,)E4(('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, , ?;<'99ylu
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;<'99y1ll
kk#&((3,/F#<***#<('3!)&
5<(<*&(-.<***M,5&JH5<(<:,)E4('3!)&*Hl#?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;<'99y1ll

kk#&#(F#<***#<('3!)&
,.<!&.&5;*&(-.<***M,5&JH),.<!&.&5;,)E4('3!)&*Hl#?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&H3ll
##39H6<.)&*wBx#zj #F.*X<!&.w?x1#v#kk.3L&#D'#!<13.3('

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<(<;,) E 4(('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;,) E 4(('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
```

```
{
```

```
{
```

```
{
```