University of Bayreuth

Institute for Computer Science

Bachelor Thesis

in Applied Computer Science

Topic:	A Constrained CYK Instances Generator: Implementation and Evaluation
Author:	Andreas Braun <www.github.com/AndreasBraun5> Matrikel-Nr. 1200197
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1. Supervisor:	Prof. Dr. Wim Martens
2. Supervisor:	M.Sc. Tina Trautner

TO MY SISTER

Abstract

The abstract of this thesis will be found here.

Zusammenfassung

Hier steht die Zusammenfassung dieser Bachelorarbeit.

Contents

1 Introduction

1.1 Motivation

The starting point of this thesis was to get a command line interface (CLI) tool to automatically generate the 4-tuples $exercise = (grammar,\ word,\ parse\ table,\ derivation\ tree)$, which are used to test if the students have understood the way of working of the CYK algorithm.

Various implementations of the Cocke-Younger-Kasami (CYK) algorithm can be found. Nevertheless none of them seemed to meet the easy to use requirements to automatically generate suitable $exercise$s, that afterwards also could be modified as wanted. Additionally the task of finding a clever algorithm to automatically generate $exercise$s with a high chance of being suitable as an exam exercise was added.

1.2 Grammar in Chromsky Normal Form

> **Definition 1. Grammar**
>
> Let there be a grammar $G = (V,\ \Sigma,\ S,\ P)$ for which the following holds:
>
> - V is a finite set of variables.
>
> - Σ is an alphabet – called terminals.
>
> - S is the start symbol and $S \in V$.
>
> - P is a finite set of rules: $P \subseteq V \times (V \cup \Sigma)^*$ – called productions.

Further it is assumed that the productions are more restricted and it holds: $P \subseteq V \times (V^2 \cup \Sigma)$. Additionally let there be a word $w \in \Sigma^*$ and a language $L(G)$ of the Grammar G.

Regarding further convenience for explaining the following default values are true:

- $V = \{A, B, ...\}$

- $(V^2 \cup\ \Sigma)^* = \{a, b, ...\} \cup \{AB, BS, AC, ...\}$

Moreover in the context of talking about sets, a set is always described beginning with an upper case letter, while one specific element of a set is described beginning with a lower case letter. Example: A "$Pyramid$" is a set consisting of multiple "$Cell$"s, which again is a subset of the set of variables "V". A "$cellElement$" is one specific element of a "$Cell$". (For further reasoning behind this example see chapter XXX "help data structure")

1.3 General approaches

Two basic approaches, that may help finding a good algorithm are explained informally.

1.3.1 Forward Problem & Backward Problem

The Forward Problem and the Backward Problem are two ways as how to determine if $w \in L(G)$.

Definition 2. Forward Problem ($G \xrightarrow{derivation} w$)

Input: Grammar G in CNF.

Output: Derivation d that shows implicitly $w \subseteq L$.

It is called Forward Problem, if you are given a grammar G and form a derivation from its root node to a final word w. The final word w is always element of $L(G)$.

Definition 3. Backward Problem = Parsing ($w \overset{?}{\subseteq} L(G)$)

Input: w and a grammar G in CNF.

Output: $w \subseteq L(G) \implies$ derivation d.

It is called Backward Problem, if you are given a word w and want to determine if it is element of $L(G)$. "This process, called parsing, is virtually always much more difficult than forming a derivation."

1.3.2 Parsing Bottom-Up & Top-Down

There are again two ways of how the approach of parsing can be classified.

Definition 4. Bottom-Up

Bottom-Up parsing means to start parsing from the leaves up to the root node.

"Bottom-Up parsing is the general method used in the Cocke-Younger-Kasami(CYK) algorithm, which fills a parse table from the "bottom up"."[Duda 8.6.3 page 426]

Definition 5. Top-Down

Top-Down parsing means to start parsing from the node down to the leaves.

Top-Down parsing means to start parsing from the node down to the leaves. "Top-Down parsing starts with the root node and successively applies productions from P, with the goal of finding a derivation of the test sentence w. Because it is rare indeed that the sentence is derived in the first production attempted, it is necessary to specify some criteria to guide the choice of which rewrite rule to apply. Such criteria could include beginning the parse at the first (left) character in the sentence (i.e., finding a small set of rewrite rules that yield the first character), then iteratively expanding the production to derive subsequent characters, or instead starting at the last (right) character in the sentence." [Duda 8.6.3 page 428]

1.4 Data Structure Pyramid

To be able to describe the way of working of the different algorithms better the help data structure *Pyramid* will be defined. But before that let there be:

Definition 6. $[i, j]$
$[i,\ j] := \{i,\ i+1, ..., j-1,\ j\} \subseteq \mathbb{N}_{\geq 0}$.

Think about if Cellij can be defined as either subset of V or (V,k)

Definition 7. $Cell_{i,j}$
$Cell_{i,j} \subseteq V$

With help of this *Pyramid* can be defined as following:

Definition 8. *Pyramid*
$Pyramid := \{Cell_{i,j} \mid i \in [0,\ i_{max}],\ j \in [0,\ j_{max} - i],\ i_{max} = j_{max} = |word| - 1\}$.

A *Pyramid* is called $EmptyPyramid \Leftrightarrow \forall i\ \forall j\ Cell_{i,j} = \emptyset$. The following is the visual representation of the *Pyramid*:
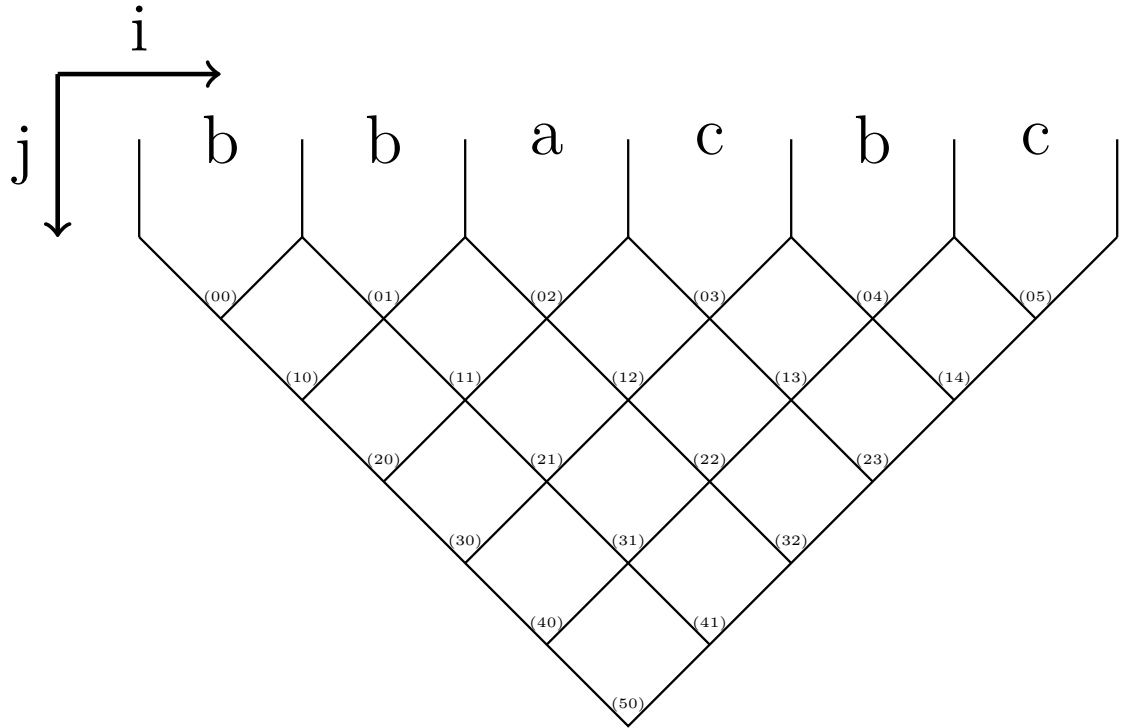
Figure 1: Visual representation of the pyramid

Definition 9. *CellDown, CellUpperLeft and CellUpperRight*
Let there be a $Cell_{i,j}$. It holds $CellDown = Cell_{i,j}$.
So there are $CellUpperLeft = Cell_{i-1,j}$ and $CellUpperRight = Cell_{i-1,j+1}$.

1.5 Cocke-Younger-Kasami CYK

Algorithm 1: CalculateSubsetForCell

    Input: $cell_{i,j} \in pyramid$

    Output: $V_{i,j} \subseteq V^2$

1 $V_{i,j} = \emptyset$;

2   **for** $k := i - 1 \rightarrow 0$ **do**

3 $V_{i,j} = V_{i,j} \cup \{X \mid X \longrightarrow YZ,\ Y \in V_{k,j},\ Z \in V_{i-k-1,k+j+1}\}$;

4   **end**

5   **return** $V_{i,j}$;

Algorithm 1 describes the magic of the CKY-algorithm. It shows what cells are taken into account while filling one cell of the parse table.

1.6 Success Rates

The Success Rates (SR) are used to compare the algorithms accounting to their performance of the different requirements. Let N be the overall count of all generated grammars of the examined algorithm.

Write down if the cellsVar or the cellsVarK are used.
Mention the basic connection between the success rates.
Write down the default values used for this.

Overall Success Rate An generated *exercise* contributes to the Overall Success Rate (SR) iff it contributes to the Success Rate Producibility (SRP), to the Success Rate Grammar Constraints (SRG) and to the Success Rate Pyramid Word Constraints ($SRPW$) at the same time.
It holds: $SR = n/N$, whereas n is the count of *exercises* that fulfil the requirements in this case.

Success Rate Producibility An generated *exercise* contributes to the SRP iff the CYK algorithm's output is true.
It holds: $SRP = p/N$, whereas p is the count of *exercises* that fulfil the requirements in this case.

Success Rate Grammar Constraints An generated *exercise* contributes to the SRG iff the following conditions are met:

- grammar has got less than a certain amount of productions.

It holds: $SRG = g/N$, whereas g is the count of *exercises* that fulfil the requirements in this case.

Success Rate Word Pyramid Constraints An generated *exercise* contributes to the $SRWP$ iff the following conditions are met:

- A certain amount of cells force a right cell combination.

- There are less than a certain amount of variables in the entire pyramid.

- There are less than a certain amount of variables in each cell of the pyramid.

It holds: $SRWP = wp/N$, whereas wp is the count of *exercises* that fulfil the requirements in this case.

Fore more detail of fore right cell combination see here:

Algorithm 2: checkForceCombinationPerCell

Input: $CellDown \subseteq V,\ CellUpperLeft \subseteq V,\ CellUpperRight \subseteq V,\ P \subseteq$
$V \times (V^2 \cup \Sigma)$

Output: $true \iff at\ least\ one\ variable\ forces$

1   $VarsForcing \subseteq V$;

2   $VarComp = \{xy \mid x \in CellUpperLeft\ \wedge\ y \in CellUpperRight\}$;

3   **foreach** $v \in CellDown$ **do**

4      $Prods = \{p \mid p \in P\ \wedge\ p = (v_1, rhse_1)\ \wedge\ v == v_1\}$;

5      $Rhses = \{rhse \mid p \in Prods\ \wedge\ p = (v_1, rhse_1)\ \wedge\ rhse == rhse_1\}$;

6   **if** $Rhses \nsubseteq VarComp$ **then**

7         $VarsForcing = VarsForcing \cup v$;

8   **end**

9   **end**

10   **return** $|VarsForcing| > 0$;

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2 Algorithms

Does the output $P \subseteq V \times (V^2 \cup \Sigma)$ imply that G is in CNF? CNF does only have useful variables [TI script Def. 8.3 page 210] vs. $P \subseteq V \times (V^2 \cup \Sigma)$.

More of a problem is that the set P is not necessarily in CNF. It is possible that there are unreachable variables – from the starting variable.

2.1 Algorithm sub modules

Sub modules are parts of the algorithms that are denoted with $\bigcirc\kern-1.0em A$, $\bigcirc\kern-1.0em B$, ... They are noteworthy procedures that need to be explained in more detail for a better understanding of the way of working of the algorithms.

2.1.1 Distribute A & B

Algorithm 3: Distribute

Input: $Rhse \subseteq (V^2 \cup \Sigma)$, V

Output: Set of productions P

1 $i \in \mathbb{N}$, $j \in \mathbb{N}$;

2 **foreach** $rhse \in Rhse$ **do**

3      $choose\ n\ uniform\ randomly\ in\ [i, j]$;

4      $V_{add} := uniform\ random\ subset\ of\ size\ n\ from\ V$;

5      $P = P \cup \{(v, rhse) \mid v \in V_{add},\ rhse \in Rhse\}$;

6 **end**

7 **return** P;

Algorithm 3 isn't needed anymore for the descriptions of the basic idea of the algorithm. It will be a module later on while tweaking the algorithms.

2.1.2 Stopping Criteria C

It is fulfilled if more than half of the pyramid cell are not empty.

2.1.3 ChooseXYDependingOnIFromRowSet D

$RowSet \subseteq \{(XY, i) \mid X, Y \in V \wedge i \in \mathbb{N}\}$

Compression of the RowSet like: (AB,3) and (AB,1) -> (AB,1) –> RowSetCompressed

rowListWeighted = add i times XY to rowListWeighted.

2.2 DiceRollOnly

2.2.1 Basic Idea

This is a very naive way of generating grammars, which will be the starting point for our algorithms to be found. Each future algorithm must have a higher score than this algorithm or otherwise it would be worse, than simple dice rolling the distribution of terminals and compound variables with removing the not contributing productions afterwards.

2.2.2 Algorithm

Algorithm 4: DiceRollOnly

    Input: Word $w \in \Sigma^*$

    Output: Set of productions P

1 $P = \emptyset$; // $P \subseteq V \times (V^2 \cup \Sigma)$

2 $P = Distribute(\Sigma,\ V)$; (A)

3 $P = P \cup Distribute(V^2,\ V)$; (B)

4 $Pyramid = CYK(G,\ w)$;

5 $P = KeepContributing(P)$;

6 **return** P;

Line 13: ?Removes all production that don't contribute, but unreachable productions still possible.? *Contributing* production iff *useful* i.e. it appears in some derivation of some terminal string from the start symbol AND *producing* i.e. it is needed for this parsing table.

As seen in table 1 the algorithm shows a relatively low success rate for producibility.

Algorithm	SR	SRP	SRG	SRWP
DiceRollOnly	09 %	24 %	87 %	53 %

Table 1: Success rates for algorithm 4

This can be explained with ... .

2.2.3 Conclusion

Something about what can be improved in another attempt or the next attempt.

2.3 **BottomUpDiceRollVar1**

2.3.1 **Basic Idea**

This algorithm uses the Bottom-Up approach where the parsing table is filled starting from the leaves. An extension compared to algorithm 4 is that productions are only added as long as the the stopping criteria isn't met.

2.3.2 **Algorithm**

Algorithm 5: BottomUpDiceRollVar1

Input: Word $w \in \Sigma^*$

Output: Set of productions P

1   $P = \emptyset$;   // $P \subseteq V \times (V^2 \cup \Sigma)$

2   $P = Distribute(\Sigma,\ V)$; $\text{\textcircled{A}}$

3   $Pyramid = CYK(G,\ w)$;

4   **for** $i := 1$ **to** i_{max} **do**

5      $J = \{0,\ ...\ ,\ j_{max} - 1\}$;   // $J \subseteq \mathbb{N}$

6      $CellSet = \emptyset$;   // $CellSet \subseteq V^2$

7   **while** $|J| > 0$ **do**

8   *choose one $j \in J$ uniform randomly*;

9          $J = J \setminus \{j\}$;

10         $CellSet = CalculateSubsetForCell(Pyramid,\ i,\ j)$;

11         $P = P \cup Distribute(CellSet,\ V)$; $\text{\textcircled{B}}$

12         $Pyramid = CYK(G,\ w)$;

13         $P = KeepContributing(P)$;

14   **if** *stopping criteria met* $\text{\textcircled{C}}$ **then**

15   **return** P;

16   **end**

17   **end**

18 **end**

19 **return** P;

Line 2: Fills the i=0 row of the pyramid.
Line 8: A cell is only visited only once.
Note: Maybe modify algorithm to also work with the threshold.

Algorithm	SR	SRP	SRG	SRWP
DiceRollOnly	09 %	24 %	87 %	53 %
BotomUpVar1	13 %	29 %	99 %	68 %

Table 2: Success rates for algorithm 5

2.3.3 Conclusion

A relatively small number of productions is already sufficient to completely fill the parsing table. This can be seen if one does take look at the log-file where the final cell that has been worked with is denoted. A good chosen stopping criteria allows a higher success rate.

2.4 BottomUpDiceRollVar2

2.4.1 Basic Idea

As seen in algorithm 5 a small number of productions is sufficient to make the parsing table quite full. If an cell is nearer to the leaves its chance to be in the set of one of the calculated sub sets for a cell is higher. Therefore you could introduce a bias that favours cells with an higher index i to allow different cell combinations.

2.4.2 Algorithm

Algorithm 6: BottomUpDiceRollVar2

Input: Word $w \in \Sigma^*$

Output: Set of productions P

1   $P = \emptyset$; // $P \subseteq V \times (V^2 \cup \Sigma)$

2   $RowSet = \emptyset$; // $RowSet \subseteq \{(XY, i) \mid X, Y \in V \wedge i \in \mathbb{N}\}$

3   $P = Distribute(\Sigma,\ V)$; (A)

4   $Pyramid = CYK(G,\ w)$;

5   **for** $i := 1$ **to** i_{max} **do**

6   **for** $j := 0$ **to** $j_{max} - i$ **do**

7         $RowSet = RowSet \cup \{(XY, i) \mid XY \in$
        $CalculateSubsetForCell(Pyramid,\ i,\ j)\}$;

8   **end**

9   **while** $threshold_i$ *not reached* **do**

10   *choose one xy out of* $(XY,\ i) \in RowSet$ *uniform randomly with*
        probability depending on i; (D)

11         $P = P \cup Distribute(xy,\ V)$; (B)

12         $Pyramid = CYK(G,\ w)$;

13         $P = KeepContributing(P)$;

14   **if** *stopping criteria met* (C) **then**

15   **return** P;

16   **end**

17   **end**

18 **end**

19 **return** P;

Line 2: Fills the i=0 row of the pyramid.
Line 7: $(AB, 1), (AB, 2), (BC, 3)... \in sub \rightarrow$ multiple occurrences of AB are allowed here yet.
Note Line 9: threshold is reached iff more than half of the cells of one row aren't empty.

Algorithm	SR	SRP	SRG	SRWP
DiceRollOnly	09 %	24 %	87 %	53 %
BotomUpVar1	13 %	29 %	99 %	68 %
BotomUpVar2	16 %	26 %	99 %	77 %

Table 3: Success rates for algorithm 6

2.4.3 Conclusion

2.5 SplitThenFill (Idea 1)

Algorithm 7: SplitThenFillPrep

Input: Word $w \in \Sigma^*$

Output: Set of productions P

1   $P = \emptyset$;   // $P \subseteq V \times (V^2 \cup \Sigma)$

2   $P = Distribute(\Sigma,\ V)$;   (A)

3   $Sol = \emptyset$;   // $Sol \subseteq \{(P_{Sol},\ Cell_{i,j}) \mid P_{Sol} \subseteq P \wedge Cell_{i,j} \in Pyramid\}$

4   $Sol = SplitThenFill(P,\ w,\ i_{max},\ 0)$;

5   **return** P_{Sol};

Line 2: Fills the i=0 row of the pyramid.

   Line **??**: $Cell_{i,j} \subseteq V \wedge Cell_{i,j} \in Pyramid$. The pyramid represents the upper part of the upper triangular matrix of the CYK. Reflection at the diagonal of the matrix and rotation of -45 degrees.

   Line **??**: Starting recursively from the tip of the pyramid.

Algorithm 8: SplitThenFill

Input: $P \subseteq V \times (V^2 \cup \Sigma)$, $w \in \Sigma^*$, $i, j \in \mathbb{N}$

Output: $(P,\ Cell_{i,j})$

1   **if** $i = 0$ **then**

2   **return** $(P,\ Cell_{i,j})$;

3   **end**

4   **if** *stopping criteria met* (C) **then**

5   **return** $(P,\ Cell_{i,j})$;

6   **end**

7   *choose one m uniform randomly in* $[j + 1,\ j + i]$;

8   $(P,\ Cell_l) = SplitThenFill(P,\ w,\ (m - j - 1),\ j)$;

9   $(P,\ Cell_r) = SplitThenFill(P,\ w,\ (j + i - m),\ m)$;

10   $Pyramid = CYK(G,\ w)$;

11   **if** $Cell_{i,j} = \emptyset$ **then**

12      $SetVc = uniform\ random\ subset\ from\ \{vc \mid v \in Cell_l\ \wedge\ c \in Cell_r\}$;

13      $P = P \cup Distribute(SetVc,\ V)$;   (B)

14   **end**

15   **return** $(P,\ Cell_{i,j})$;

Line 2: Recursion anchor that returns the up to this point modified productions P and the variables in the cell with index i and j.

   Line 8 + 15: Analogous to the CYK-algorithm a cell combination $Cell_l$ and $Cell_r$ is chosen that can generate the sub string.

   Line 5: A Recalculation is done at each recursive call because only the updated production set P is returned recursively.

The stopping criteria would be, that each marked $cell_{i,j} \neq \emptyset$ and it must be possible

to get from $cell_{m,j}$ and $cell_{i-m,m+j+1}$ to $cell_{i,j}$ through applying one of the production rules.

Algorithm **??** Idea1 uniform randomly generates a predefined structure of the derivation tree. You always update the pyramid after adding one production to the grammar. Now there are two options to fill the parse table:

1. Bottom Up: The parse table is filled relatively evenly. All information regarding the upper cells are available and can be used. Similar to the CYK Algorithm approach.

2. Top Down: The parse table is filled quiet unevenly. You don't have all information available. Think about adding a production for the node cell: You can add a production so that its producing cells fill the node cell, but you don't know what actually would be the best to fill in these producing cells because they themselves aren't looked at yet. This problem is kept until the last depth of the recursion, where the cells in row $i = 0$ are taken into account. Only starting there you know what variables actually produce the terminals.

   Maybe solution: For the Top Down approach, don't assume that the terminals are already distributed over the V. Distribute the terminals over the variables in an ideal way that fits your already generated productions best.

2.5.1 Algorithm: Idea 2, How often cells are used for subset calculations

2.6 SplitAndFill

Algorithm 9: SplitAndFillPrep

Input: Word $w \in \Sigma^*$

Output: Set of productions P

1 $P = \emptyset$; // $P \subseteq V \times (V^2 \cup \Sigma)$

2 $Pyr = EmptyPyramid$;

3 $Cell_{i_{max},0} = Cell_{i_{max},0} \cup \{S\}$; // $Cell_{i,j} \subseteq V \wedge Cell_{i,j} \in Pyramid$

4 ;

5 $Sol = \emptyset$; // $Sol \subseteq \{(P_{Sol}, \ Pyramid) \mid P_{Sol} \subseteq P\}$

6 $Sol = SplitThenFill(Pyr, \ P, \ w, \ i_{max}, \ 0)$;

7 $P = P \cup P_{Sol}$;

8 **return** P;

Line 2: $EmptyPyramid \Leftrightarrow \forall i \ \forall j \ Cell_{i,j} = \emptyset$

Line 5: $Cell_{i,j} \subseteq V \wedge Cell_{i,j} \in Pyramid$. The pyramid represents the upper part of the upper triangular matrix of the CYK. Reflection at the diagonal of the matrix and rotation of -45 degrees.

Algorithm 10: SplitAndFill

Input: $Pyramid\ Pyr,\ P \subseteq V \times (V^2 \cup \Sigma),\ w \in \Sigma^*,\ i, j \in \mathbb{N}$

Output: $(P,\ Pyr)$

1 **if** *stopping criteria met* $\textcircled{C}$ **then**

2 | **return** $(P,\ Pyr)$;

3 **end**

4 $SetAll = V \times V$;

5 $Pyramid = CYK(G,\ w)$;

6 *choose one m uniform randomly in* $[j + 1,\ j + i]$;

7 **if** $(m - j - i) \neq 0$ **then**

8 | $(P,\ Pyr) = SplitAndFill(Pyr,\ w,\ (m - j - 1),\ j)$;

9 **end**

10 else then

11 $v = choose\ one\ variable\ uniform\ randomly\ from\ Cell_{i,j}$;

12 $P = P \cup (v - - > w_j)$;

13 end

14 **if** $(j + i - m) \neq 0$ **then**

15 | $(P,\ Pyr) = SplitAndFill(Pyr,\ w,\ (j + i - m),\ m)$;

16 **end**

17 else then

18 $v = choose\ one\ variable\ uniform\ randomly\ from\ Cell_{i,j}$;

19 $P = P \cup (v - - > w_j)$;

20 end **return** $(P,\ Cell_{i,j})$;

2.7 Comparision of Algorithms

Algorithm	SR	SRP	SRG	SRWP
DiceRollOnly	09 %	24 %	87 %	53 %
BotomUpVar1	13 %	29 %	99 %	68 %
BotomUpVar2	16 %	26 %	99 %	77 %

Table 4: Comparison of the success rates of the algorithms

3 CLI Tool

Write much of this stuff in the appendix.

3.1 Scoring Model

Only valid ResultSamples are given a score. Parameters to be scored:

- RightCellCombinationsForcedCount

- maxSumOfVarsInPyramidCount

- maxNumberOfVarsPerCellCount

- maxSumOfProductionsCount

Maybe add a diversity criterion = homogeneity of the cells to the scoring matrix.

Parameter	Points					
	2	4	6	8	10	-100
cellCombinationsForced	[0,10]	[11,20]	[21,30]	[41,50]	[31,40]	>50
sumVarsInPyramid	[0,10]	[11,20]	[21,30]	[41,50]	[31,40]	>50
maxVarsPerCell	[5,5]	[4,4]	[1,1]	[3,3]	[2,2]	>5
sumProductions	[1,2]	[3,4]	[5,6]	[9,10]	[7,8]	>10

Table 5: Scoring of the different parameter values

Based on table 5 each result sample is scored. Out of the #??? best result samples one can choose.
The result will be normalized to the maximum possible points -> range 0.0 to 1.0.

3.2 Short Requirements Specification

Use Cases i $\longrightarrow$ "Lastenheft".
Input and Output parameter identification.
Here is described what the finished tool must and can do.

Generating the latex code and storing it in .tex-file. Then converting the .tex-file to .pdf-file via:
Runtime rt = Runtime.getRuntime();
Process pr = rt.exec("pdflatex mydoc.tex");
Process pr = rt.exec("pdflatex mydoc.tex");
Process pr = rt.exec("pdflatex mydoc.tex");
The triple invocation of LaTeX is to ensure that all references have been properly resolved and any page layout changes due to inserting the references have been accounted for. [http://www.arakhne.org/autolatex/]

3.2.1 Exam Exercises

An exam exercise consists out of a grammar, a word, a parsing table and a derivation tree. Creating a exam exercises must be possible. Therefore it is needed:

- Selection of a possible exam exercise out of high scoring samples $\longrightarrow$ calculateSamples.jar [Input parameter: countOfNewSamples (better scoring samples in exchange for longer computation time)], which upon execution fills samples.txt with new high scoring samples, together with its actual scoring model parameters. Out of this samples one can be selected manually that is used for an exam exercise.

- Modifying of a exam exercise candidate: Changing the grammar and changing the word. [?changing the pyramid (I think no, because of the strong interconnection between the grammar and the parsing table it is already covered through being able to change the grammar)?] $\longrightarrow$ calculateExamExercise.jar [Input parameter: examExercise.txt], that updates pre defined information for one sample upon execution.

- Predefined Information: It is a printable version of the finished exam exercise like grammar.png, parsingTable.png and derivationTable.png together with its latex code, that was used for its creation - modification later one possible. Also it is examExerciseInfo.txt, that has the information about its actual scoring model parameters.

Maybe don't add the *Tree* here. It is only needed as one part of the output of an *exercise*.
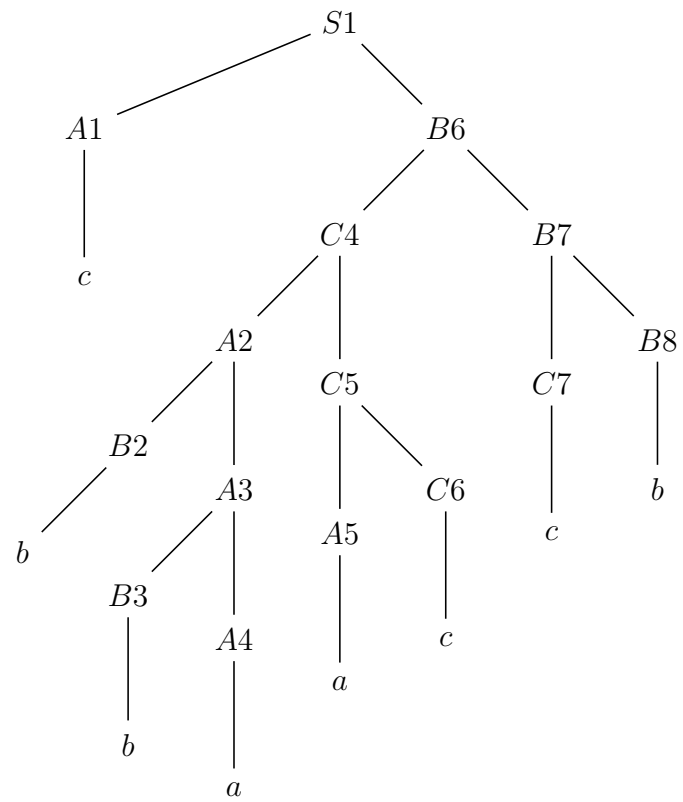
3.3 Overview - UML

UML-Diagramm showing the general idea of the implementation.

List noteworthy used libraries here, too.

Maybe some information out of the statistics tool of IntelliJ.

3.3.1 UML: More Detail 1

3.3.2 UML: More Detail 2

3.4 User Interaction

Here the specific must can do's are explained with short examples.

3.4.1 Use Case 1

3.4.2 Use Case i

References

[1] JSR 220: Enterprise Java Beans 3.0 `https://jcp.org/en/jsr/detail?id=220`, 09/09/2015