

*Online Appendix for the paper  
Exchanging Conflict Resolution in an  
Adaptable Implementation of ACT-R*

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## Appendix A Formalization of ACT-R

In this section, the fundamental concepts of ACT-R are formalized and transferred to CHR.

### **A.1 The Basic Unit of Knowledge: Chunks**

ACT-R is a *symbolic* production rule system, i.e. all declarative information is represented in form of symbols and associations of symbols and the procedural information is stored in form of production rules transforming the declarative information. Hence, the ACT-R production system is defined over a set of symbols  $\mathfrak{S}$ . The smallest unit of declarative information is a *chunk*, which basically is a structured assembly of symbols. It has a unique name and a number of labeled *slots* which can hold one single *symbol*. The chunk names and the slot labels are symbols themselves. If a chunk has a symbol naming a chunk in its slot, the two chunks are *connected*. We require the *unique-name assumption* for symbols. The concept of chunks and their connections in form of chunk stores is defined in section A.2.

### **A.2 Chunk Stores**

We extend the abstract notion of chunks given in section A.1 to a definition of chunk descriptions embedded into chunk stores which represent a network of chunks with the help of three relations.

#### *Definition 1 (Chunk Description)*

A chunk with name  $c$  and type  $t$  and corresponding slots and values can be represented as a term  $\text{chunk}(c, t, \{(s, v) \mid c \text{ has the value } v \text{ in slot } s\})$ .

#### *Definition 2 (Chunk Store)*

A *chunk-store*  $\Gamma$  over a set of symbols  $\mathfrak{S}$  is a tuple  $(\mathbb{C}, \mathbb{E}, \mathcal{T}, \text{HasSlot}, \text{Isa})$ , where  $\mathbb{C}$  is a set of chunk identifiers and  $\mathbb{E}$  a set of primitive elements both identified by unique names. The *values* of  $\Gamma$  are defined by the set  $\mathbb{V} = \mathbb{C} \cup \mathbb{E}$ .  $\mathcal{T}$  is a set of chunk-types. The set  $\mathbb{T}$  then denotes the set of all type names. A chunk-type  $T = (t, S) \in \mathcal{T}$  is a tuple with a

*unique* type name  $t \in \mathbb{T}$  and a set of slots  $S \subseteq \mathbb{S}$  where  $\mathbb{S}$  is the set of all slot names. The sets  $\mathbb{C}$ ,  $\mathbb{E}$ ,  $\mathbb{T}$  and  $\mathbb{S}$  are disjoint:  $\mathbb{C} \cup \mathbb{E} \cup \mathbb{T} \cup \mathbb{S} \subseteq \mathbb{G}$ .

$\text{HasSlot} \subseteq \mathbb{C} \times \mathbb{S} \times \mathbb{V}$  and  $\text{Isa} \subseteq \mathbb{C} \times \mathbb{T}$  are relations and are defined as follows:

$$\text{chunk}(c, t, S) \in \Gamma \Leftrightarrow c \in \mathbb{C} \wedge c \text{ Isa } t \wedge \forall (s, v) \in S : (c, s, v) \in \text{HasSlot}.$$

The Isa relation has to be right-unique and left-total, so each chunk has to have exactly one type. A chunk-store is *type-consistent*, iff the following two conditions hold:

1.  $\forall c \in \mathbb{C}, s \in \mathbb{S}, v, v' \in \mathbb{V} : (c, s, v) \in \text{HasSlot} \wedge (c, s, v') \in \text{HasSlot} \Rightarrow v = v'$
2.  $\forall c \in \mathbb{C}, s \in \mathbb{S}, v \in \mathbb{V}, t \in \mathbb{T}, S \subseteq \mathbb{S} : c \text{ Isa } t \wedge (c, s, v) \in \text{HasSlot} \wedge (t, S) \in \mathcal{T} \Rightarrow s \in S$

With this definition, a chunk store can be implemented directly in CHR by defining the sets and relations as constraints. The constraint `chunk(C, T)` is a condensed representation of the set of chunk symbols  $C$  and the Isa relation. This is possible, since each chunk in  $\mathbb{C}$  has exactly one type. The ternary HasSlot relation is represented by constraints of the form `chunk_has_slot(C, S, V)` stating that  $(c, s, v) \in \text{HasSlot}$ .

Chunk types can be represented by a constraint `chunk_type(T, S)` where  $T$  is the symbol denoting the chunk type and  $S$  is a list of symbols for the slots. Note that there can be added rules to ensure type-consistency and uniqueness of the relations as defined in definition 2.

### A.3 Buffer Systems

*Definition 3 (buffer system)*

A *buffer system* is a tuple  $(B, \Gamma, \text{Holds})$ , where  $B$  is a set of buffer names,  $\Gamma = (\mathbb{C}, \mathbb{E}, \mathcal{T}, \text{HasSlot}, \text{Isa})$  a type-consistent chunk-store and  $\text{Holds} \subseteq B \times C$  a right-unique relation that assigns every buffer at most one chunk that it holds. Buffers which do not appear in the Holds relation are called *empty*.

A buffer system is *consistent*, if every chunk that appears in Holds is a member of  $C$  and  $\Sigma$  is a type-consistent chunk-store. It is *clean*, if its chunk-store only holds chunks which appear in Holds.

In CHR, the set  $B$  and the Holds relation can be represented as a constraint `buffer/3` which holds the name of the buffer, the corresponding module (needed for requests) and the name of the chunk it holds as a reference to the chunk store. This is possible since each buffer holds at most one chunk. Empty buffers can be represented by the empty symbol `nil`. For each buffer, there must be exactly one `buffer` constraint. This transforms the Holds relation to a left-total and right-unique relation.

#### A.3.1 Production Rules

*Definition 4 (Production Rules)*

An ACT-R production rule is of the form `(p name buffer_test* ==> action*)` where `name` is a unique symbol indicating the name of the rule. Buffer tests are also denoted as the left-hand-side (LHS), actions as the right-hand-side (RHS) of a rule. A buffer test has the form `=buffer> isa t s1 v1 ... sn vn` where the symbol `buffer` references the name of the tested buffer and the rest stands for a chunk description

$\text{chunk}(c, t, \{(s_1, v_1), \dots, (s_n, v_n)\})$  for a chunk with arbitrary name  $c$ . The values  $v_i$  can be symbols or variable symbols, where variable symbols are indicated by the prefix  $=$ .

An action has the form  $\#buffer > s_1 v_1 \dots s_n v_n$  where the  $\#$  is a place-holder for the available actions  $=$ ,  $+$ ,  $-$  denoting modifications, requests and clearings respectively. The other symbols are defined as for the buffer tests. Note that for requests, the first slot symbol must be **isa** followed by a chunk type as value. The values might be variables again, but have to be bound on the left-hand-side of the rule i.e. appear on LHS.

#### *Definition 5 (Applicability of a Production Rule)*

A production rule with buffer tests  $=b_1> \text{isa } t_1 s_{1,1} v_{1,1} \dots s_{1,n_1} v_{1,n_1} \dots =b_k> \text{isa } t_k \dots s_{k,n_k} v_{k,n_k}$  is applicable in a buffer system iff  $\exists \bar{v} \in V : b_1 \in B \wedge \exists c_1 \in C \wedge c_1 \text{ Isa } t \wedge (c_1, s_{1,1}, v_{1,1}) \in \text{HasSlot} \dots \wedge \bar{v} = \bar{x}$  where  $\bar{v}$  denotes a set of values in  $V$  and  $\bar{x}$  the variable symbols used on the LHS.

#### *A.3.2 Translation of rules*

The production rules as defined in definition 4 operate on the buffer system: They match the content of the buffers and transform it with a defined set of actions. Hence, an ACT-R rule can be transferred to a CHR rule  $H ==> G \mid B$ , where the head  $H$  and guard  $G$  represent the applicability condition of the rule as defined in definition 5 and the body  $B$  contains the actions.

*Applicability Condition* The applicability of a rule in definition 5 can be translated directly to the CHR counterparts of the relations. I.e. each relational condition in the applicability condition is expressed by the respective constraint in the head of the rule. The guard is filled with the conditions from  $\bar{v} = \bar{x}$ . Note that the condition has a set-based semantics (since idempotency can be reduced in classical logic). I.e., for the special case of duplicate tests on the LHS of a production rule, additional rules have to be generated with all possible combinations of unifications of duplicate pairs to implement the set-based semantics of CHR  $\omega_{set}$  as shown in (?). In the following, we assume that the production rules are duplicate-free.

*Actions* The actions of a production rule transform the buffer system in the way as they have been defined in section ???. The transformations of the buffer system can be realized in CHR by using destructive update as described in (?), p. 32). I.e. each action has a trigger constraint **action/2** which gets the name of the buffer and the specification of the action encoded as a chunk-description (see definitions 1 and 4). The trigger constraints then use abstract methods to access the buffer system like **set\_buffer** to set the content of a buffer. This simplifies the compilation and the form of the resulting rules, since the constraints representing the relations of the buffer system only appear in the kept head of the resulting CHR rules and never in the removed head. Additionally, it simplifies extensions and adaptations of the actions, since the compiler must not be changed but only the framework implementing the actions. One adaptation of the simplest form of actions which only apply the changes to the buffers is shown in section ?? when we introduce scheduling to postpone the actual application of the changes an action performs.

## Appendix B Evaluation Results

In this appendix we list the results of our experiments as described in section ??.

### B.1 Samples

In table B 1 and B 3, the used samples of player 2 and player 3 are listed. Table B 2 and B 4 show the frequencies of rock, paper and scissors within one sample. The sum  $\Sigma$  is a control value and ensures that 20 moves have been produced per sample. The  $p$  values denote the probabilities of rock, paper and scissors respectively.

Table B 1. *Samples for Player 2*

Sample	1	r	r	r	p	r	p	p	p	r	r	r	r	p	r	p	p	p	r	p	r
2	r	p	p	p	p	p	r	r	p	r	r	r	r	p	r	p	r	r	p	p	p
3	r	p	r	p	r	r	p	p	r	p	r	p	r	p	r	r	r	r	p	p	p
4	r	r	r	r	r	r	p	r	p	r	r	p	p	p	p	p	p	r	r	p	p
5	p	p	p	p	p	r	p	p	r	p	p	r	p	r	r	p	p	r	p	p	p
6	p	p	p	p	r	p	r	r	r	r	r	r	r	r	r	p	p	p	r	p	p
7	p	p	r	p	p	p	r	r	r	r	r	r	r	p	r	r	r	r	r	r	r
8	r	r	p	r	r	p	r	p	p	r	r	r	r	p	r	r	r	p	r	r	r
9	p	p	p	p	p	r	r	p	r	p	r	r	r	r	r	r	p	r	r	r	r
10	p	p	p	p	p	p	p	p	r	r	p	r	p	r	p	r	r	p	p	p	p
11	p	r	p	p	r	p	r	r	r	r	p	p	p	r	r	p	r	p	p	p	p
12	p	p	r	r	p	r	r	p	r	r	r	p	p	p	r	p	r	r	r	r	p
13	p	r	p	r	p	r	r	p	r	p	r	p	r	p	r	p	r	p	r	p	r
14	p	r	p	p	r	r	p	p	r	p	r	p	p	r	p	p	r	p	p	p	p
15	p	r	p	p	p	p	r	r	r	r	r	r	r	r	r	p	r	p	p	r	r
16	r	r	p	r	p	p	p	r	p	r	r	r	r	r	p	r	p	r	r	r	p
17	r	r	r	r	r	p	p	p	r	r	r	r	r	p	p	p	r	r	p	p	p
18	r	r	r	r	r	r	r	p	r	p	r	r	p	r	r	p	p	p	p	p	r
19	r	p	r	r	r	p	p	p	p	r	p	r	p	r	p	r	p	r	p	r	p
20	p	p	r	r	p	p	r	r	p	r	r	p	r	r	p	r	r	p	r	p	p

Table B 2. *Samples for Player 2 – Frequencies and Probabilities*

Sample	#rock	#paper	#scissors	$\Sigma$	$p_r$	$p_p$	$p_s$
1	11	9	0	20	0.55	0.45	0
2	10	10	0	20	0.5	0.5	0
3	11	9	0	20	0.55	0.45	0
4	11	9	0	20	0.55	0.45	0
5	6	14	0	20	0.3	0.7	0
6	11	9	0	20	0.55	0.45	0
7	13	7	0	20	0.65	0.35	0
8	14	6	0	20	0.7	0.3	0
9	11	9	0	20	0.55	0.45	0
10	5	15	0	20	0.25	0.75	0
11	9	11	0	20	0.45	0.55	0
12	11	9	0	20	0.55	0.45	0
13	10	10	0	20	0.5	0.5	0
14	7	13	0	20	0.35	0.65	0
15	11	9	0	20	0.55	0.45	0
16	12	8	0	20	0.6	0.4	0
17	11	9	0	20	0.55	0.45	0
18	13	7	0	20	0.65	0.35	0
19	9	11	0	20	0.45	0.55	0
20	11	9	0	20	0.55	0.45	0
Average	10.35	9.65	0	20	0.5175	0.4825	0

Table B 3. *Samples for Player 3*

Sample	r	s	s	s	p	s	s	r	s	r	s	r	s	s	r	r	p	s	r	p
1	s	s	s	s	p	s	s	r	s	r	s	r	s	s	r	r	p	s	r	p
2	p	r	p	s	p	p	s	r	r	s	s	r	s	s	s	s	s	r	p	s
3	s	p	r	r	p	p	r	s	r	p	r	s	s	s	r	s	r	p	p	s
4	r	s	s	p	r	s	p	p	r	p	s	p	r	r	s	r	r	p	r	p
5	r	p	p	p	s	s	p	r	r	p	s	r	r	p	p	r	s	p	p	s
6	s	s	p	s	s	r	p	r	p	p	p	s	r	p	p	p	p	s	s	r
7	s	p	r	p	s	p	r	r	p	p	p	r	r	s	s	r	r	p	p	p
8	s	s	r	p	r	r	r	p	s	p	r	s	s	p	p	r	p	s	p	p
9	r	r	s	r	r	r	r	s	p	r	r	s	r	p	s	r	r	s	p	r
10	p	r	r	r	r	p	s	p	s	r	p	s	r	r	s	s	s	s	s	s
11	r	r	s	p	s	s	s	r	s	s	p	p	p	p	p	r	s	s	s	p
12	r	s	r	p	r	s	s	r	r	p	r	r	p	p	r	s	r	p	r	r
13	s	r	r	s	r	s	p	s	p	p	p	p	s	r	s	s	p	r	r	p
14	p	p	p	r	r	s	r	s	r	p	p	s	r	s	r	s	r	p	p	p
15	r	s	s	s	r	p	r	s	s	s	r	s	s	p	s	r	s	s	s	s
16	s	r	p	r	s	r	p	r	r	p	r	r	s	r	r	r	r	r	r	r
17	s	p	s	s	p	s	r	p	p	p	r	s	r	s	r	r	s	r	s	p
18	r	r	s	p	r	s	p	p	p	r	r	p	r	r	p	s	r	s	p	s
19	r	r	s	s	r	s	s	r	r	r	s	s	r	p	s	s	r	p	r	p
20	r	s	p	r	r	s	s	r	p	r	p	p	r	p	p	s	s	s	r	r

Table B 4. *Samples for Player 3 – Frequencies and Probabilities*

Sample	#rock	#paper	#scissors	$\Sigma$	$p_r$	$p_p$	$p_s$
1	6	3	11	20	0.3	0.15	0.55
2	5	5	10	20	0.25	0.25	0.5
3	7	6	7	20	0.35	0.3	0.35
4	8	7	5	20	0.4	0.35	0.25
5	6	9	5	20	0.3	0.45	0.25
6	4	9	7	20	0.2	0.45	0.35
7	7	9	4	20	0.35	0.45	0.2
8	6	8	6	20	0.3	0.4	0.3
9	12	3	5	20	0.6	0.15	0.25
10	7	4	9	20	0.35	0.2	0.45
11	4	6	10	20	0.2	0.3	0.5
12	11	5	4	20	0.55	0.25	0.2
13	6	7	7	20	0.3	0.35	0.35
14	7	8	5	20	0.35	0.4	0.25
15	5	2	13	20	0.25	0.1	0.65
16	14	3	3	20	0.7	0.15	0.15
17	6	6	8	20	0.3	0.3	0.4
18	8	7	5	20	0.4	0.35	0.25
19	8	3	9	20	0.4	0.15	0.45
20	8	6	6	20	0.4	0.3	0.3
Average	7.25	5.80	6.95	20	0.36	0.29	0.35

### B.2 Reinforcement-Learning-Based Utility Learning

Tables B 5, B 7 and B 9 show the results of the ACT-R implementation of player 1, 2 and 3 respectively. In tables B 6, B 8 and B 10 the results of our CHR implementation can be found. The  $U$  values denote the utilities for rock, paper and scissors respectively, where the other values show the performance of the model in the corresponding sample as a control of equal program flows of the two implementations.

Table B 5. Results for Player 1 – Reward-Based Utility Learning (ACT-R)

Sample	Utilities			Performance		
	$U_r$	$U_p$	$U_s$	#win	#draw	#defeat
1	0.000	1.873	-0.020	19	0	1

Table B 6. Results for Player 1 – Reward-Based Utility Learning (CHR)

Sample	Utilities			Performance		
	$U_r$	$U_p$	$U_s$	#win	#draw	#defeat
1	0.000	1.873	-0.020	19	0	1

Table B 7. Results for Player 2 – Reward-Based Utility Learning (ACT-R)

Sample	Utilities			Performance		
	$U_r$	$U_p$	$U_s$	#win	#draw	#defeat
1	0.000	1.799	-0.020	10	9	1
2	0.000	1.822	-0.020	9	10	1
3	0.000	1.833	-0.020	10	9	1
4	0.000	1.743	-0.020	10	9	1
5	0.000	0.000	1.267	14	0	6
6	0.000	0.000	0.977	9	0	11
7	0.000	0.000	0.084	7	0	13
8	0.000	1.849	-0.020	13	6	1
9	0.000	0.000	0.248	9	0	11
10	0.000	0.000	1.329	15	0	5
11	0.000	0.000	1.289	11	0	9
12	0.000	0.000	0.775	9	0	11
13	0.000	0.000	1.076	10	0	10
14	0.000	0.000	1.442	13	0	7
15	0.000	0.000	0.721	9	0	11
16	0.000	1.836	-0.020	11	8	1
17	0.000	1.763	-0.020	10	9	1
18	0.000	1.760	-0.020	12	7	1
19	0.000	1.805	-0.020	8	11	1
20	0.000	0.000	0.834	9	0	11
Average	0.000	0.811	0.493	10.400	3.900	5.700

Table B 8. Results for Player 2 – Reward-Based Utility Learning (CHR)

Sample	Utilities			Performance		
	$U_r$	$U_p$	$U_s$	#win	#draw	#defeat
1	0.000	1.799	-0.020	10	9	1
2	0.000	1.822	-0.020	9	10	1
3	0.000	1.833	-0.020	10	9	1
4	0.000	1.743	-0.020	10	9	1
5	0.000	0.000	1.267	14	0	6
6	0.000	0.000	0.977	9	0	11
7	0.000	0.000	0.084	7	0	13
8	0.000	1.849	-0.020	13	6	1
9	0.000	0.000	0.248	9	0	11
10	0.000	0.000	1.329	15	0	5
11	0.000	0.000	1.289	11	0	9
12	0.000	0.000	0.775	9	0	11
13	0.000	0.000	1.076	10	0	10
14	0.000	0.000	1.442	13	0	7
15	0.000	0.000	0.721	9	0	11
16	0.000	1.836	-0.020	11	8	1
17	0.000	1.763	-0.020	10	9	1
18	0.000	1.760	-0.020	12	7	1
19	0.000	1.805	-0.020	8	11	1
20	0.000	0.000	0.834	9	0	11
Average	0.000	0.811	0.493	10.400	3.900	5.700

Table B 9. Results for Player 3 – Reward-Based Utility Learning (ACT-R)

Sample	Utilities			Performance		
	$U_r$	$U_p$	$U_s$	#win	#draw	#defeat
1	0.000	0.533	-0.091	3	12	5
2	0.000	-0.052	-0.159	4	10	6
3	0.000	0.000	0.689	6	7	7
4	0.257	-0.020	-0.020	4	7	9
5	0.720	-0.131	-0.020	4	8	8
6	0.000	0.000	0.633	9	7	4
7	0.000	0.000	1.046	9	4	7
8	0.000	0.732	-0.090	5	10	5
9	0.000	1.254	-0.020	11	3	6
10	0.000	0.000	0.363	4	9	7
11	0.000	0.434	-0.020	3	6	11
12	-0.052	1.575	-0.075	7	6	7
13	0.274	-0.092	-0.052	4	9	7
14	0.000	0.000	0.977	8	5	7
15	1.717	-0.020	-0.020	12	4	4
16	0.000	1.764	-0.052	13	4	3
17	0.000	0.000	0.736	6	8	6
18	0.000	0.541	-0.020	7	7	6
19	0.000	1.083	-0.020	7	3	10
20	-0.020	0.918	-0.036	6	5	9
Average	0.145	0.426	0.187	6.600	6.700	6.700

Table B 10. *Results for Player 3 – Reward-Based Utility Learning (CHR)*

Sample	Utilities			Performance		
	$U_r$	$U_p$	$U_s$	#win	#draw	#defeat
1	0.000	0.533	-0.091	3	12	5
2	0.000	-0.052	-0.159	4	10	6
3	0.000	0.000	0.689	6	7	7
4	0.257	-0.020	-0.020	4	7	9
5	0.720	-0.131	-0.020	4	8	8
6	0.000	0.000	0.633	9	7	4
7	0.000	0.000	1.046	9	4	7
8	0.000	0.732	-0.090	5	10	5
9	0.000	1.254	-0.020	11	3	6
10	0.000	0.000	0.363	4	9	7
11	0.000	0.434	-0.020	3	6	11
12	-0.052	1.575	-0.075	7	6	7
13	0.274	-0.092	-0.052	4	9	7
14	0.000	0.000	0.977	8	5	7
15	1.717	-0.020	-0.020	12	4	4
16	0.000	1.764	-0.052	13	4	3
17	0.000	0.000	0.736	6	8	6
18	0.000	0.541	-0.020	7	7	6
19	0.000	1.083	-0.020	7	3	10
20	-0.020	0.918	-0.036	6	5	9
Average	0.145	0.426	0.187	6.600	6.700	6.700

### B.3 Success-/Cost-Based Utility Learning

Tables B 11, B 13 and B 15 show the results of the ACT-R implementation of player 1, 2 and 3 respectively. In tables B 12, B 14 and B 16 the results of our CHR implementation can be found. The meaning of the values corresponds to Appendix B.2.

Table B 11. Results for Player 1 – Success-/Cost-Based Utility Learning (ACT-R)

Sample	Utilities			Performance		
	$U_r$	$U_p$	$U_s$	#win	#draw	#defeat
1	19.95	19.95	19.95	0	20	0

Table B 12. Results for Player 1 – Success-/Cost-Based Utility Learning (CHR)

Sample	Utilities			Performance		
	$U_r$	$U_p$	$U_s$	#win	#draw	#defeat
1	19.95	19.95	19.95	0	20	0

Table B 13. Results for Player 2 – Success-/Cost-Based Utility Learning (ACT-R)

Sample	Utilities			Performance		
	$U_r$	$U_p$	$U_s$	#win	#draw	#defeat
1	3.790	19.830	13.250	8	10	2
2	6.550	19.822	9.925	8	10	2
3	6.550	19.863	13.250	10	8	2
4	2.144	19.755	13.250	5	13	2
5	9.925	19.790	14.913	7	11	2
6	9.925	19.832	13.250	11	7	2
7	9.925	19.890	16.575	16	2	2
8	4.838	19.868	9.925	11	7	2
9	9.925	19.803	9.925	10	8	2
10	9.925	19.653	9.925	4	14	2
11	9.925	19.846	14.913	10	8	2
12	9.925	19.847	9.925	10	8	2
13	9.925	19.859	14.913	11	7	2
14	9.925	19.830	13.250	7	11	2
15	9.925	19.825	13.250	11	7	2
16	4.838	19.868	14.913	11	7	2
17	2.550	19.796	9.925	5	13	2
18	1.817	19.805	13.250	6	12	2
19	6.550	19.791	9.925	7	11	2
20	9.925	19.858	9.925	10	8	2
Average	7.440	19.822	12.419	8.9	9.1	2

Table B 14. Results for Player 2 – Success-/Cost-Based Utility Learning (CHR)

Sample	Utilities			Performance		
	$U_r$	$U_p$	$U_s$	#win	#draw	#defeat
1	3.790	19.830	13.250	8	10	2
2	6.550	19.822	9.925	8	10	2
3	6.550	19.863	13.250	10	8	2
4	2.144	19.755	13.250	5	13	2
5	9.925	19.790	14.913	7	11	2
6	9.925	19.832	13.250	11	7	2
7	9.925	19.890	16.575	16	2	2
8	4.838	19.868	9.925	11	7	2
9	9.925	19.803	9.925	10	8	2
10	9.925	19.653	9.925	4	14	2
11	9.925	19.846	14.913	10	8	2
12	9.925	19.847	9.925	10	8	2
13	9.925	19.859	14.913	11	7	2
14	9.925	19.830	13.250	7	11	2
15	9.925	19.825	13.250	11	7	2
16	4.838	19.868	14.913	11	7	2
17	2.550	19.796	9.925	5	13	2
18	1.817	19.805	13.250	6	12	2
19	6.550	19.791	9.925	7	11	2
20	9.925	19.858	9.925	10	8	2
Average	7.440	19.822	12.419	8.9	9.1	2

Table B 15. Results for Player 3 – Success-/Cost-Based Utility Learning (ACT-R)

Sample	Utilities			Performance		
	$U_r$	$U_p$	$U_s$	#win	#draw	#defeat
1	13.985	9.925	14.887	10	6	4
2	13.228	9.908	9.881	10	3	7
3	11.307	11.304	9.925	6	6	8
4	9.268	9.925	9.913	5	6	9
5	6.550	4.838	11.883	6	7	7
6	7.381	3.850	9.865	6	7	7
7	9.858	16.558	12.394	9	7	4
8	5.536	6.583	15.258	7	6	7
9	11.279	14.182	9.925	9	4	7
10	4.888	6.557	4.888	4	4	12
11	13.210	6.550	3.790	6	7	7
12	13.208	19.871	7.850	11	7	2
13	9.875	8.450	9.925	7	5	8
14	9.925	11.116	9.925	6	7	7
15	17.658	9.925	6.550	11	5	4
16	19.888	17.394	14.913	9	9	2
17	13.875	6.550	9.875	7	7	6
18	17.356	16.214	14.887	8	9	3
19	13.539	9.925	9.925	8	7	5
20	15.439	13.835	9.925	8	9	3
Average	11.863	10.673	10.319	7.65	6.4	5.95

Table B 16. Results for Player 3 – Success-/Cost-Based Utility Learning (CHR)

Sample	Utilities			Performance		
	$U_r$	$U_p$	$U_s$	#win	#draw	#defeat
1	13.985	9.925	14.888	10	6	4
2	13.228	9.908	9.881	10	3	7
3	11.307	11.304	9.925	6	6	8
4	9.268	9.925	9.913	5	6	9
5	6.550	4.838	11.883	6	7	7
6	7.381	3.850	9.865	6	7	7
7	9.858	16.558	12.394	9	7	4
8	5.536	6.583	15.258	7	6	7
9	11.279	14.182	9.925	9	4	7
10	4.888	6.557	4.888	4	4	12
11	13.210	6.550	3.790	6	7	7
12	13.208	19.871	7.850	11	7	2
13	9.875	8.450	9.925	7	5	8
14	9.925	11.116	9.925	6	7	7
15	17.658	9.925	6.550	11	5	4
16	19.888	17.394	14.913	9	9	2
17	13.875	6.550	9.875	7	7	6
18	17.356	16.214	14.888	8	9	3
19	13.539	9.925	9.925	8	7	5
20	15.439	13.835	9.925	8	9	3
Average	11.863	10.673	10.319	7.65	6.4	5.95

**B.4 Random Estimated Costs**

Tables B 17, B 19 and B 21 show the results of the ACT-R implementation of player 1, 2 and 3 respectively. In tables B 18, B 20 and B 22 the results of our CHR implementation can be found. The results have been produced by the first sample of each player and have been run 50 times. The meaning of the values corresponds to Appendix B.2. In the tables containing the results of the CHR implementation, we added the error squares of the averages over all runs to compare them to the reference implementation.

Table B 17. Results for Player 1 – Random Estimated Costs (ACT-R)

Run	Utilities			Performance		
	$U_r$	$U_p$	$U_s$	#win	#draw	#defeat
1	19.970	19.942	9.937	12	7	1
2	19.975	19.680	9.761	11	8	1
3	9.565	19.987	9.912	18	1	1
4	9.836	19.813	9.900	18	1	1
5	19.891	19.958	9.972	14	5	1
6	19.390	19.999	9.596	13	6	1
7	4.918	19.874	9.940	16	3	1
8	19.877	19.831	9.985	11	8	1
9	19.854	19.873	9.850	13	6	1
10	19.802	19.825	9.918	11	8	1
11	3.630	19.997	9.837	16	3	1
12	19.995	19.985	9.878	11	8	1
13	6.403	19.731	9.986	17	2	1
14	19.518	19.992	9.960	9	10	1
15	9.958	19.958	9.994	18	1	1
16	19.215	19.883	9.960	9	10	1
17	19.876	19.800	9.833	14	5	1
18	19.992	19.900	9.854	9	10	1
19	19.569	19.834	9.769	15	4	1
20	19.837	19.802	9.981	10	9	1
21	9.971	19.950	9.500	18	1	1
22	19.968	19.847	9.912	11	8	1
23	19.628	19.833	9.900	13	6	1
24	9.887	19.936	9.934	18	1	1
25	9.104	19.982	9.998	18	1	1
26	19.698	19.742	9.940	12	7	1
27	19.665	19.992	9.920	14	5	1
28	13.036	19.990	9.909	17	2	1
29	19.648	19.904	9.872	14	5	1
30	19.994	19.898	9.837	12	7	1
31	19.967	19.862	9.997	11	8	1
32	19.480	19.922	9.986	13	6	1
33	19.978	19.883	9.839	10	9	1
34	19.929	19.974	9.994	14	5	1
35	19.730	19.984	9.738	11	8	1
36	19.511	19.974	9.886	13	6	1
37	9.995	19.892	9.843	18	1	1
38	19.596	19.907	9.569	13	6	1
39	19.837	19.802	9.981	10	9	1
40	9.166	19.959	9.922	18	1	1
41	19.513	19.989	9.980	9	10	1
42	19.669	19.943	9.711	15	4	1
43	3.362	19.936	9.901	15	4	1
44	19.964	19.956	9.998	14	5	1
45	3.761	19.653	9.596	16	3	1
46	9.559	19.961	9.987	18	1	1
47	5.939	19.948	9.985	17	2	1
48	19.632	19.942	9.724	11	8	1
49	19.531	19.948	9.996	9	10	1
50	19.748	19.895	9.976	14	5	1
Average	15.991	19.901	9.883	13.62	5.38	1

Table B 18. Results for Player 1 – Random Estimated Costs (CHR)

Run	Utilities			Performance		
	$U_r$	$U_p$	$U_s$	#win	#draw	#defeat
1	19.645	19.969	9.986	11	8	1
2	19.220	19.948	9.749	10	9	1
3	9.990	19.860	9.920	18	1	1
4	19.838	19.941	9.922	12	7	1
5	19.917	19.945	9.949	12	7	1
6	19.948	19.886	9.964	12	7	1
7	19.997	19.777	9.903	12	7	1
8	19.868	19.966	9.930	10	9	1
9	19.274	19.968	9.874	13	6	1
10	9.810	19.937	9.975	18	1	1
11	19.596	19.655	9.953	10	9	1
12	19.998	19.807	9.937	12	7	1
13	9.868	19.858	9.600	18	1	1
14	19.988	19.884	9.804	13	6	1
15	9.865	19.992	9.763	18	1	1
16	6.509	19.965	9.947	17	2	1
17	19.729	19.991	9.914	14	5	1
18	19.965	19.929	9.876	14	5	1
19	19.941	19.907	9.985	15	4	1
20	19.567	19.997	9.970	16	3	1
21	6.639	19.983	9.998	17	2	1
22	18.898	19.760	9.953	13	6	1
23	19.998	19.989	9.806	8	11	1
24	9.820	19.804	9.763	18	1	1
25	19.968	19.684	9.937	16	3	1
26	19.664	19.912	9.420	12	7	1
27	18.650	19.957	9.883	11	8	1
28	9.754	19.938	9.711	18	1	1
29	19.912	19.963	10.000	10	9	1
30	15.796	19.983	9.875	15	4	1
31	19.100	19.951	9.998	10	9	1
32	9.876	19.998	9.915	18	1	1
33	19.923	19.755	9.696	12	7	1
34	19.779	19.739	9.948	13	6	1
35	19.877	19.878	9.906	10	9	1
36	15.719	19.979	9.969	15	4	1
37	13.243	19.882	9.842	17	2	1
38	19.319	19.984	9.608	12	7	1
39	9.906	19.948	9.915	18	1	1
40	3.165	19.862	9.960	16	3	1
41	19.904	19.952	9.977	13	6	1
42	5.120	19.970	9.794	17	2	1
43	3.746	19.787	9.855	16	3	1
44	19.440	19.820	9.921	11	8	1
45	19.893	19.755	9.706	9	10	1
46	19.953	19.874	9.797	14	5	1
47	19.871	19.992	9.920	12	7	1
48	6.265	19.844	9.763	17	2	1
49	4.798	19.823	9.868	16	3	1
50	9.958	19.702	9.870	18	1	1
Average	15.610	19.893	9.870	13.94	5.06	1
Error Square of Average	0.145	0.000	0.000	0.10	0.10	0.00

Table B 19. Results for Player 2 – Random Estimated Costs (ACT-R)

Run	Utilities			Performance		
	$U_r$	$U_p$	$U_s$	#win	#draw	#defeat
1	2.851	14.046	6.389	6	12	2
2	15.635	17.997	16.560	11	7	2
3	9.429	18.123	9.769	8	10	2
4	13.311	19.899	9.937	9	9	2
5	9.971	19.907	9.500	9	10	1
6	13.230	18.308	14.984	11	7	2
7	12.996	19.690	9.900	9	9	2
8	9.887	19.882	9.934	9	10	1
9	9.104	19.966	9.998	9	10	1
10	4.944	19.794	14.953	10	8	2
11	9.920	19.694	9.985	10	8	2
12	16.936	19.769	15.888	8	10	2
13	6.475	19.673	9.872	9	9	2
14	13.047	15.082	13.330	7	11	2
15	6.659	19.743	9.976	9	9	2
16	12.606	19.802	9.986	6	12	2
17	6.292	19.782	9.987	9	9	2
18	13.257	18.705	9.994	9	9	2
19	7.346	17.109	9.815	8	9	3
20	13.139	19.689	15.914	12	6	2
21	6.327	19.813	9.997	9	9	2
22	3.847	18.342	9.569	8	10	2
23	6.623	19.630	9.909	9	9	2
24	9.166	19.924	9.922	9	10	1
25	4.913	19.979	9.761	8	10	2
26	11.565	19.889	9.711	7	11	2
27	3.538	19.892	14.923	9	9	2
28	7.995	19.915	13.310	8	10	2
29	13.105	19.520	9.940	9	9	2
30	9.735	19.985	9.920	10	8	2
31	13.102	19.982	9.909	9	9	2
32	6.367	19.822	9.724	9	9	2
33	6.657	19.903	9.841	9	9	2
34	4.849	15.462	6.306	7	10	3
35	5.697	19.996	9.878	9	9	2
36	9.259	16.904	9.916	8	9	3
37	9.891	19.808	9.960	9	10	1
38	7.346	17.109	9.815	8	9	3
39	6.400	19.952	9.689	9	9	2
40	9.995	19.801	9.843	9	10	1
41	3.133	19.822	9.743	8	10	2
42	9.833	18.376	9.688	8	10	2
43	3.353	17.750	8.480	6	9	5
44	13.313	19.406	9.761	9	9	2
45	9.565	19.976	9.912	9	10	1
46	9.836	19.654	9.900	9	10	1
47	6.601	19.922	9.932	9	9	2
48	9.364	9.125	13.513	7	8	5
49	9.559	19.928	9.987	9	10	1
50	5.939	19.902	9.985	8	11	1
Average	8.878	18.923	10.588	8.64	9.36	2

Table B 20. Results for Player 2 – Random Estimated Costs (CHR)

Run	Utilities			Performance		
	$U_r$	$U_p$	$U_s$	#win	#draw	#defeat
1	13.261	19.844	13.319	10	8	2
2	6.585	17.628	9.744	9	9	2
3	16.548	17.907	17.098	11	7	2
4	9.823	19.906	9.907	10	8	2
5	14.785	19.965	9.967	8	10	2
6	11.344	11.680	9.960	5	12	3
7	15.863	19.948	9.831	7	11	2
8	9.608	19.874	9.941	9	10	1
9	9.377	19.855	9.794	9	10	1
10	11.966	19.909	15.979	9	10	1
11	4.679	19.174	9.928	8	10	2
12	8.550	16.976	9.902	6	11	3
13	9.552	19.889	9.999	9	10	1
14	9.815	19.972	9.629	9	10	1
15	9.798	19.968	9.994	9	10	1
16	9.940	19.733	9.929	10	8	2
17	9.553	19.670	9.905	9	10	1
18	13.267	19.345	9.964	9	9	2
19	9.848	19.615	9.892	10	8	2
20	6.534	19.053	9.770	9	9	2
21	6.396	16.268	9.997	4	13	3
22	19.911	19.855	9.976	5	14	1
23	9.901	19.608	9.917	9	10	1
24	14.898	18.181	15.851	11	7	2
25	13.085	18.217	9.774	9	9	2
26	6.142	19.978	9.906	8	11	1
27	9.978	19.714	9.997	9	10	1
28	9.946	19.894	9.984	10	8	2
29	13.131	18.608	9.943	9	9	2
30	6.049	19.969	9.866	9	9	2
31	9.715	19.993	9.795	10	8	2
32	6.643	19.977	9.980	9	9	2
33	12.765	19.883	9.895	9	9	2
34	9.968	19.996	9.829	10	8	2
35	4.781	17.791	9.975	8	10	2
36	4.970	17.242	9.678	8	9	3
37	6.309	19.924	9.979	9	9	2
38	14.950	19.948	9.921	8	10	2
39	8.410	19.962	9.760	9	10	1
40	4.734	12.897	9.914	3	13	4
41	9.802	19.798	9.815	9	10	1
42	3.181	16.619	6.550	5	10	5
43	9.881	19.954	9.959	9	10	1
44	6.471	19.441	9.763	8	11	1
45	9.978	18.432	9.654	8	10	2
46	9.932	19.972	9.938	9	10	1
47	7.731	17.000	9.978	6	11	3
48	6.647	18.836	9.801	9	9	2
49	13.095	19.908	9.943	9	9	2
50	9.898	17.710	9.965	10	8	2
Average	9.800	18.910	10.275	8.46	9.66	1.88
Error Square of Average	0.850	0.000	0.098	0.032	0.090	0.014

Table B 21. Results for Player 3 – Random Estimated Costs (ACT-R)

Run	Utilities			Performance		
	$U_r$	$U_p$	$U_s$	#win	#draw	#defeat
1	15.827	9.996	9.755	10	6	4
2	14.019	9.976	7.990	7	8	5
3	15.242	9.986	10.701	8	7	5
4	13.066	6.292	9.963	6	10	4
5	14.952	9.935	6.643	8	8	4
6	14.255	9.815	7.259	7	8	5
7	14.236	9.689	7.676	7	8	5
8	15.102	9.843	9.995	9	7	4
9	15.215	9.569	10.978	8	7	5
10	13.902	9.909	7.947	7	8	5
11	15.931	9.922	9.250	10	6	4
12	14.264	9.761	7.943	7	8	5
13	15.192	9.739	9.518	9	7	4
14	15.892	9.934	14.901	10	6	4
15	16.847	9.998	16.631	8	8	4
16	14.757	9.804	10.722	8	7	5
17	15.647	9.987	7.338	9	7	4
18	13.311	9.802	13.220	7	9	4
19	14.091	6.207	4.188	6	9	5
20	14.902	9.841	6.651	8	8	4
21	14.084	9.976	11.592	7	8	5
22	14.281	9.878	6.822	7	8	5
23	15.985	9.658	14.928	10	6	4
24	15.826	9.960	9.902	10	6	4
25	15.287	9.975	9.691	9	7	4
26	13.946	9.689	11.942	9	7	4
27	15.586	9.843	11.759	9	7	4
28	19.997	9.743	14.043	9	8	3
29	15.750	12.970	13.227	9	7	4
30	14.977	9.922	13.238	8	8	4
31	14.717	9.983	9.602	9	7	4
32	16.228	6.057	9.794	9	8	3
33	14.325	9.900	4.573	8	7	5
34	15.218	9.932	9.953	9	7	4
35	14.997	9.596	11.532	7	8	5
36	14.717	9.987	14.909	7	10	3
37	15.076	9.985	11.289	8	7	5
38	14.174	9.850	7.219	7	8	5
39	15.861	9.841	14.918	10	6	4
40	15.995	9.784	14.837	10	6	4
41	14.637	6.661	6.218	7	8	5
42	14.900	9.928	11.405	8	7	5
43	16.652	9.960	16.463	7	9	4
44	15.219	9.994	9.958	9	7	4
45	15.168	9.960	10.971	8	7	5
46	15.347	6.324	3.722	7	10	3
47	19.997	9.743	14.043	9	8	3
48	14.882	9.308	5.265	7	10	3
49	15.189	9.981	9.894	9	7	4
50	14.575	9.500	4.925	8	7	5
Average	15.205	9.558	10.158	8.18	7.56	4.26

Table B 22. Results for Player 3 – Random Estimated Costs (CHR)

Run	Utilities			Performance		
	$U_r$	$U_p$	$U_s$	#win	#draw	#defeat
1	13.740	9.705	12.345	7	8	5
2	14.087	14.862	11.889	9	6	5
3	15.194	15.668	12.500	10	6	4
4	14.236	9.965	13.760	9	8	3
5	13.540	9.987	12.323	6	10	4
6	13.625	9.763	9.938	9	7	4
7	11.586	9.619	11.092	6	9	5
8	12.797	9.776	6.243	8	8	4
9	12.236	9.826	7.403	7	8	5
10	12.866	9.665	3.433	8	7	5
11	18.338	9.749	16.961	9	8	3
12	13.199	9.960	9.909	8	8	4
13	12.028	9.833	7.950	7	8	5
14	13.474	9.806	9.996	9	7	4
15	14.089	9.947	14.594	10	6	4
16	13.537	9.976	9.928	9	7	4
17	12.286	9.826	7.756	7	8	5
18	13.775	14.918	11.935	7	9	4
19	14.501	14.990	12.425	6	9	5
20	13.521	9.889	9.374	9	7	4
21	14.871	9.886	13.275	6	9	5
22	11.454	9.994	7.765	7	8	5
23	12.444	9.960	7.566	7	8	5
24	13.737	9.959	11.457	9	7	4
25	14.271	9.982	11.175	8	7	5
26	12.497	9.996	7.856	7	8	5
27	13.911	5.955	6.613	8	9	3
28	15.236	14.945	15.767	8	8	4
29	13.526	9.991	9.950	9	7	4
30	13.543	9.647	9.975	9	7	4
31	14.071	5.900	6.341	8	9	3
32	13.740	9.825	7.419	9	7	4
33	13.353	9.992	9.800	9	7	4
34	14.471	14.791	11.022	7	9	4
35	13.548	6.574	14.925	8	8	4
36	13.672	9.977	9.928	10	6	4
37	13.798	9.746	14.372	6	10	4
38	13.904	9.805	14.841	10	6	4
39	13.270	9.262	13.232	8	8	4
40	14.102	9.877	14.890	10	6	4
41	13.328	9.830	9.782	9	7	4
42	12.690	14.938	10.930	6	11	3
43	13.234	9.930	6.402	8	8	4
44	12.342	9.686	7.297	7	8	5
45	13.552	9.961	9.745	9	7	4
46	12.840	9.690	9.765	8	8	4
47	13.640	9.832	9.687	9	7	4
48	13.576	9.938	9.756	9	7	4
49	13.338	5.930	4.055	7	10	3
50	11.620	9.836	3.150	8	8	4
Average	13.525	10.267	10.210	8.06	7.78	4.16
Error Square of Average	2.823	0.503	0.003	0.014	0.048	0.010