#### **RandomNumberGeneratorforMicrocontrollers**

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## Abstract

Theuseofrandomnumbersinsmallmicrocontrollerscanbeveryusefulandfunin roboticprogrammingtoyielddifferentbeh aviorsfromthesamerobotic -controlprogram. Duetolimitedmemory,thelackoffloating -pointmathcapabilities,andthelimitedsize ofintegersinsomemicrocontrollers,thepracticalimplementationofasoftware -based randomnumbergeneratorforamic rocontrollerisdifficult.Thispaperlooksatthe generationofrandomnumbers,andcomparesavarietyofimplementationsonthe 68HC11microcontroller.Agoodimplementationischosenandthe68HC11sourcecode isprovidedthatimplementsit.

## Keywords

- Randomnumbergenerator
- Microcontroller
- 68HC11

## Introduction

Theuseofrandomnumbersinroboticsprojects(andothermicrocontrollerprojects)can -followingrobotgetsstuckinacornerat beverybeneficial.I'veseencaseswhereawall justtherightang le. This is due to the algorithm hard -coredintoitsbrainproducingthe exactsameback -up,turn -a-bit,go -forwardmotionthatjustoscillatesback -and-forthin thecorner(whatadisgrace!).Whilethoseofuswhoknowactuallywhatisgoingon inside of the robot can understand this behavior, to the general public this makes the robotappeartobequitebrainlessanddim -witted;thesillythingcan'tevenfinditsway outofacorner.Iftheprogramintherobotusedrandomnumberstoslightlyalterthe back-uptimeandalsothedegreesturnedinthesamealgorithm, this robot would not havetheexactsameback -and-forthbehaviorthatcausedittobestuckinthecorner.It wouldveryquicklymaneuveroutofthecornerandcontinueonitsway, makingitsel f (andyou)lookquiteintelligent. The availability of random numbers is also critical for manyalgorithmsincorporatingartificialintelligence(AI)techniquessuchasgenetic algorithmsandgeneticprogramming, neural networks, and fuzzylogic.

Thetop icofrandomnumbergeneration(RNG)canbedividedintotwocategories:true randomnumbergenerationandpseudorandomnumbergeneration.Withtruerandom numbergenerationthenextrandomnumbergeneratedisnotknown,andthesequenceof randomnumbers cannotbere -generated.Withpseudorandomnumbergeneration,a sequenceof"randomnumbers"isgeneratedusingaknownalgorithm,andtheexactsame sequencecanbere -generated;hencetheclassificationofpseudo.Forsoftware -based systemsitisveryd esirabletouseapseudorandomnumbergeneratortobeabletotest thesoftwaresystemwitharepeatablesetofrandomnumbers;youcanre -runatestwith

the exacts a merandom numbers being used. Also, with a pseudor and omnumber generator that has been formally evaluated, you can be sure of the attributes of the resulting sequence of random numbers.

Randomnumbergeneratorsprovidedwithworkstation -classcomputersreturna32 -bitor a64 -bitintegernumber,orafloating -pointnumberbasedonthatvalu e.Insmall microcontrollersthedesiredvaluefromarandomnumbergeneratorisgenerallyan8 -bit or16 -bitnumber.Iwillfocusmydiscussiononan8 -bitrandomnumbergenerator.

InmyrandomnumbergeneratorstudyIfirstdefinedasetofgradingcrit eriaasattributes IwantedinaRNG.IthenwroteaJavaprogramonaPCtotestvariousRNGalgorithms againstthesecriteria,andwrotedifferentalgorithmsinthisprogramtogenerate sequencesofrandomnumberstotest.Ithenusedtheknowledgegaine dfromthe successfulalgorithmsanddatatowrite68HC11 -specificassemblylanguageto implementtheRNG.LastlyIranthealgorithmonthe68HC11andverifiedthatit generatedthesamesequenceseenintheJavacode.

## HardwareMethods

Atruerandomnumbe rgeneratorcanbeimplementedinhardwareusingawhite -noise[1] source.In [2]and [3]ahardwarecircuitisdetailedwhichisusedtogeneraterandom numbersusingapai roftransistors.In [4]anotherhardware -basedrandomnumber generatorisdiscussed.Whilequiteusefulinmanyroboticapplications,ahardware -based randomnumbergeneratorisoutofthescopeofthefocusofthispaper.

## Requirementsofa"Good"RandomNumberGenerator

Two famous quotes on random numbers help us to set the stage for our discussion:

"Anyonewhoconsidersarithmeticalmethodsofproducingrandomdigitsis, ofcourse,inastateofsin." JohnvonNeumann(1903 -1957).

"Thegenerationofrandomnumbersistooimportanttobelefttochance." RobertR.Coveyou,OakRidgeNationalLaboratoryinTennessee(Foundin IvarsPeterson,'TheJunglesofRandomness'pp178).

In the "art" of generating random numbers with a software-based algorithm, you must determine the attribute syou wanting random number generator, and then test candidate random number generators against the segrading criteria. It is difficult to define a set of attributes that are simpleto define, eas ily test able, and complete enough to avoid sequences of numbers that follows implepattern, such as a simple sequence (0, 1, 2, 3, 4, 5...) or an increment -by-Nsequence (0, 21, 42, 63, 84...). Table 1 shows the list of attributes I used intesting candidate random number generators target edfort the 68 HC11 microcontroller.

Attribute	Parameters	Rational					
Numberrange	Numbersfrom0to255are	An8 -bitpro cessortypically					
	generated.	workswith8 -bitintegers. *					
Numberrange	Theaverageofthisnumberrangeis	Definitionusedinsubsequent					
average	127.5.	attributes.(255 -0)/2					
Cyclelength	Thelengthofthepseudorandom	ThelongertheRNGcycleis					
	sequenceshallbeatleast16,384	themorevariedtheresulting					
	$(2^{1})$ to 32,76 8(2 <sup>-16</sup> ), but preferably will be 65,536(2 <sup>-16</sup> ).	behavioris.					
Numbercoverage	Allnumbersintherangewillbe	ThroughthecycleIwantall					
	includedthesamenumberoftimes	numberstooccuranequal					
	inthecycle.Forexample,fora	numberoftimessothe					
	cyclelengthof65,536 ,allnumbers	averageofthecycleisthe					
	0through255willoccur256times.	averageoftherange,andto					
		ensurealipossiblevaluescan					
Largowindow	Theoverage of 128 sequential	Tomakasurathatany					
average	valuesinallselectionsfrom the	sequenceofthissizewill					
average	cyclewillyaryfromtheaverageby	coveragoodrangeofyalues					
	atmost20% (i.e. 102to153).	coveragoourangeorvaraes.					
Largewindow	Theaverageof128sequent ial	Tomakesurethewindow					
min/max	valuesinatleast1selectionwill	averagesarenotallthesame					
	varyfromtheaveragebyatleast	ortoosimilar.					
	10% above and by another						
	selectionbyatleast10% below						
	(i.e.114to140).						
Mediumwindow	Theaverageo f32sequentialvalues	Tomakesurethatany					
average	inallselectionsfromthecyclewill	sequenceofthissizewill					
	40% (i.e. 75to 170)	coveragoodrangeorvalues.					
Mediumwindow	Theaverageof32sequentialvalues	Tomakesurethewindow					
min/max	inatleast1selectionwillvary	averages are not all the same					
	fromtheaveragebyatleast20%	ortoosimilar.					
	aboveandbyanotherselectionby						
	atleast20%below(i.e.102to						
	153).						
Smallwindow	Theaverageof8sequential values	Tomakesurethatany					
average	inallselectionsfromthecyclewill	sequenceofthissizewill					
	varyfromtheaveragebyatmost	coveragoodrangeofvalues.					
	75%(i.e.31to223).						

 Table 1. AttributesforgradingaRNGsequence.

Attribute	Parameters	Rational
Smallwindow	Theaverageof8sequentialvalues	Tomakesurethewindow
min/max	inatleast1selec tionwillvary	averagesarenotallthesame
	fromtheaveragebyatleast35%	ortoosimilar.
	aboveandbyanotherselectionby	
	atleast35%below(i.e.79to175).	
Repeatedvalues	Aspecificvaluemayormaynotbe	Iwanttodiscussthis, butdo
	repeatedinthe cycle.	notplantospecifya
		requirementortestforthis.
Repeatedsequence	Thereisnotanysequenceof4or	3ormorerepeatednumbers
	morenumbersthatisrepeatedin	maybeuseful/interesting,but
	thecycle.	Idon'twa ntmorethan3.
Value-to-value	Asequencemadefromthe	Tomakesurethereareno
delta	differences in the sequence values	patternsinthedeltas.
	willmeetthesamewindowaverage	
	andrepeatedcriteria.	
Operationsallowed	Theoperationsusedtoimplemen t	Thesearethecommonly
	therandomnumbergeneratorare	offeredmicrocontroller
	8-bitadd,subtract,shift,and,or,	operations, and are found in
	exor,and8 -bitto16 -bitmultiply.	the68HC11.
Sizeofcode	Thesizeofthecodetoimplement	Smallerisbetterwiththe
	thealgorithm istobeassmallas	limitedprogramspace
	possible-atmost100byteslong.	availableinmicrocontrollers.
RequiredRAM	ThenumberofbytesofRAM	Smallerisbetterwiththe
	requiredtobededicatedtothe	limitedRAMspaceavailable
	RNGshouldbeasfewaspossible	inmicrocontrollers.
	andatmost8b ytes.	

## Approaches

Investigatedtwodifferentapproachestoimplementingasimplerandomnumber generatorforthe68HC11;atable -basedmethodandanalgorithm -basedmethod.

#### Table-basedmethod

Foratableof256values, there are 256!, or 8.578e+506, possible ways to order the table. Mytable - based approach uses a table of 128 values, with an algorithm to traverse the table in a number of ways to generate a random number cycle 16 , 384 numbers in length. I used a table of 128 values and generated the other 128 values in the set of {0 -255} by EXOR ing the 128 table values with \$FF. An acceptable set of 128 values for this table include the criteria that EXOR ing these 128 values will generate the remaining 128 values in the set. This gave mean effective table of 256 values using only 128 by tes, plus as mall number of by tes for the implementation logic. It then traverse this set of 256 values in 64 different ways to generate a sequence of 16, 384 values. Once the driving logic of traversing the table was developed, agood set of table values as seen in Table 2 wasdeterminedusingaprogramtotrydifferentsetsofvalues \*andtestingtheresulting tableagainstthe specifiedgradingcriteria.IwroteaJavaprogramtodothis.

**Table 2.** 128valuesusedinmytable-basedRNG.

247, 86,108,103,252, 35,115, 75,202, 70,107, 89, 37, 40,246, 69 111,234, 56, 12,249,146, 19, 80,240,230, 45, 38,197,223, 65,123 26,208, 74,167,130, 79,253,121,173, 93,136,198, 64,204, 90, 20 30,233,222, 16, 14,242,182,174,195,105,159, 0, 84,129,250,155 10, 28,251, 92, 62,201,192,142,104,168,114, 49,172,124, 39, 67 143,254, 7,228,116, 31,154,156, 72, 36,194,161, 71,226, 59, 18 209,177,118,189,221, 98,135,122, 48,153,117,231,238,164, 52,187 44,145,170,213, 77, 97,128,212, 41,160, 11,149,200, 23, 50,179

AsthiswasthefirstalgorithmIgotworking,Iwasveryattachedtoit.Ithendecidedto investigateequation -basedmethodsforcompletenessinmyresearch,butIreallyliked mytable -basedsolutionandthoughtIwoulddecidetogowithitasthebestsolution.

#### Equation-basedmethod

Anequation -basedmethodusesaseedvalue.Foreachrandomnu mbergeneratedthis seedvalueisusedinanalgorithmtogenerateanewseedvalueandalsotherandom number.Inthecaseofour8 -bitrandomnumbergeneratorwewanttheseedvaluetobe largerthanthe8 -bitnumberreturnedsincean8 -bitseedvaluewi llallowatmostacycle of256values.Aclassicalgorithmforgeneratingrandomnumbers [5]is:

unsigned long seed ; ... seed = 1664525L \* seed + 1013904223L ;

Implementcodeusinga16 -bitseedvalue,andlookedatdiffere nt8 -bitselectionsfrom thenewly -calculatedseedasthe8 -bitrandomnumbertoreturn.

## ComparisonofApproaches

Thetable -basedalgorithmgeneratedagoodsetofrandomnumbers.Thecyclewas 16,384numberslong.Theresultingmemory -sizewas4bytesof RAM.Forprogram memoryitrequired62bytes,plus128bytesfortheinitialtable,foratotalof190bytes.

Theequation -basedalgorithmalsogeneratedagoodsetofrandomnumbers, and the cyclewasafull65,536numbers. The resulting memory -sizewas also4bytesof RAM. For programmemory it required only 25 bytes. The resulting equation was:

seed = 181 \* seed + 359 ;

whereseedisa16 -bitnumber.Thereturned8 -bitvalueisthetophalfoftheseed.

<sup>&</sup>lt;sup>\*</sup>IusedtherandomnumbergeneratorinJavatopopulatethe table,checkingfortheEXORcriteria,then testedtheresultingtableagainstallofthegradingcriteria.Runningthiscodefoundmanyacceptable solutionsforthetabledata;Ichooseoneformyimplementation.

UsingaJavaprogramonthePC,Isearchedfora goodsetofconstants.Thenumbers181 and359wereusedintheequation(bothprimenumbers),whichgeneratedasequenceof 65,536valuesthatsatisfiedthegradingcriteria.Basedonthesizeofthecodeandthe complexityofthealgorithm,Ihavetoch oosetheequation -basedalgorithmasthebest randomnumbergeneratorcodeformyapplicationswiththe68HC11microcontroller.

## Testing

The65,536sequenceofnumbersgeneratedbytheequationweretestedagainstthe criteriain Table 1.Ilistthefirstpartofthissequencein Table 3below.Noticethatthe number172isrepeated,whichinformsusthatanumbercanberepeatedinthesequence.

					•					U		•	-		
1	255	117	4	73	222	125	232	15	167	21	110	230	252	49	27
35	65	133	50	218	156	132	185	223	239	99	114	197	223	22	226
226	208	81	76	71	229	135	182	203	4	237	226	1	207	119	85
60	170	216	82	145	187	134	223	211	228	179	190	152	202	84	116
50	209	28	68	182	28	128	52	248	145	181	6	139	210	172	63
196	68	28	34	183	10	119	181	56	9	242	186	219	229	129	182
243	2	216	237	149	132	106	98	148	78	108	218	134	5	210	217
189	13	80	163	78	137	89	59	13	94	34	102	141	48	159	168
35	99	131	69	227	27	68	64	161	59	20	94	241	104	232	35
38	6	115	211	85	56	43	113	81	228	66	194	176	172	172	74
196	244	31	77	162	226	13	206	30	88	171	146	203	251	237	29
254	47	135	179	203	23	236	87	б	153	81	206	66	87	170	156
213	181	171	5	209	217	199	12	10	165	51	118	22	190	227	199
71	136	138	67	178	39	158	237	43	126	81	138	69	50	152	158
85	167	38	109	111	0	113	250	103	34	171	11	208	177	201	33

**Table 3.** Thefirstpartofthe65,536numbersgeneratedbytheequation.

Asafoll ow-onstudyofthesequenceofrandomnumbers, Iplotted the maspairs on 256x256 grid. In Figure 11eft, three sets of 1024 pairs are plotted as red, blue, and black points. In Figure 11eft, half of the 32,768 pairs of numbers are plotted. Both of these plots visually look random, while the right figure is starting to look abitgrid -like.



Figure 1. Threesets of 1024 pairs (left), and alf of the pairs (right) plotted.

In Figure 2belowall32,768pairsofnumbersinthe65,536setofrandomnumbersare plotted.Thisdoesproduceadefinitepattern,butitwastheleastpattern -likepatternI foundinsearchingmany pairsofcandidateconstants.Anotherkeypointofthissetisthat outof32,768pairsofnumbers,therewere32,768uniquepointsplotted ;therewereno duplicatepointsfromthepairsofnumbers.



Figure 2.All32,768pairsplotted,resultinginanoticeablepattern.

Mostotherconstantsusedfortheequationcoveredfarfewerpositionswhenplotted, and generatedamuchmoredefiniteandsimplepattern.Forexample, whenusing 33 and 171 for the constants, the plot as shown in Figure 3 only covers 9728 out of 32,768 positions.



 $Figure \ 3. All 32, 768 pairs plotted for a different set of constants, resulting in a very noticeable pattern$ 

Afterextensivetestingandlookingatbothalgorithmsanddifferentconstantvalues, I settledontheequation -basedwiththeconstants181and359. This is the algorithm implemented in the 68 HC11 example code.

#### TheCode

Belowis68HC11assemblycodeim plementtheequationtogeneraterandomnumbers. Alllinesstartingwith'\*'arecomments,andtextfollowinga';'signarealsocomments. ThiswasassembledwiththeAS11assembler \*andloadedina68HC11E1ona BOTBoardusingPCBUG11fortesting.

```
* Author: Tom Dickens
* Date: 12/10/2002
* Purpose:
* 8-bit random number generator for the 68HC11/68HC12 microcontrollers.
* Method: A 16-bit version of the classic S = S * M + A is used.
* The random number returned is the high byte of the new seed value.
^{\ast} The constants M and A were chosen to be M=181 and A=359 for
* the following reasons:
  - They are both prime numbers.
*
  - The equation cycles through all 65,536 possible seed values.
*
  - The 8-bit random values have the following properties:
     - Each number from 0 to 255 occurs 256 times in the cycle.
     - No sequence of 4 or more numbers is repeated.
     - Every 128-byte window has an average between 104 and 150 (+-20%)
     - Every 32-byte window has an average between 76 and 178 (+-40)
     - Every 8-byte window has an average between 32 and 218 (+-75%)
+
  - Small RAM and Program space required:
*
     - Only 4 bytes of RAM are used
*
     - Only 25 bytes of Program memory is used
*
  - Fast. Each call to Random takes 64 clock cycles, which
*
        is 32 micro-seconds with an 8MHz crystal.
* Usage:
  JSR Random
             ; Register A contains the next random number.
 * RAM Usage: 4 bytes in zero-page RAM, bytes 0 trough 3.
ORG $0000
RandomSeed:
   RMB 2
             ; reserve two bytes in RAM for the SEED value
SEED HIGH EQU RandomSeed ; high byte of the seed
SEED_LOW
        EQU RandomSeed+1 ; low byte of the seed
RandomScratch:
    RMB 2
              ; reserve 2 bytes in RAM for scratch space
* Constants:
MULTIPLIER EQU
              181
ADDER
        EQU
              359
* Program start: Example program to display random numbers on Port C.
 Change ORG $B600 to $F800 for `E2 devices.
ORG $8600
    LDAA #$FF
    STAA $1007 ; Set port C to output
```

\*Version2.1,dated10 -Aug-91,byRandy Sargent.

```
TIDAA
          #$00
     STAA
          $1003
                   ; Clears port C to $00
     BSR RandomInit
                    ; Seed = 0
TopLoop:
     BSR Random
                    ; A = random number
     STAA $1003
                    ; A -> Port C
     LDX #$FFFF
     BSR DelayX
                   ; wait a bit to see the number
     BRA TopLoop
                    ; do it again...
* Initialize the random number generator to the start of the sequence
* by loading the seed value with 0. You can initialize the seed to any
* value you choose, which will start the random number sequence somewhere
* else in the cycle of 65,536 seed values.
RandomInit:
     CLR
          SEED_HIGH
     CLR
          SEED_LOW
     RTS
* Return the next 8-bit pseudo random number in the sequence
* of 65,536 numbers, generated from the following equation:
* SEED = SEED * 181 + 359
* The random value returned is the high byte of the new SEED.
* Return value: The A register.
                       (bytes, cycles) = (25, 64)
*****
Random:
                          ; (1,3) Remember the current value of B
    PSHB
* scratch = seed * multiplier
    LDAA #MULTIPLIER
                        ; (2,2) A = #181
    LDAB SEED_LOW
                         ; (2,3) B = the low byte of the seed
                         ; (1,10) D = A \times B
     MUT.
     STD
         RandomScratch
                         ; (1,4) scratch = D
     LDAA #MULTIPLIER
                         ; (2,2) A = \#181
     LDAB SEED_HIGH
                          ; (2,3) B = the high byte of the seed
     MUL
                          ; (1,10) D = A \times B
* low byte of MUL result is added to the high byte of scratch
     ADDB RandomScratch ; (2,3) B = B + scratch_high
     STAB RandomScratch
                          ; (2,3) scratch = seed * 181
                         ; (2,4) D = scratch
        RandomScratch
    LDD
         n אשעעה
RandomSeed
    ADDD #ADDER
                         ; (3,4) D = D + 359
                      ; (3,4) D - D - 322; (2,4) remember new seed value
     STD
* (A = SEED_HIGH from ADDD instruction)
     PULB
                         ; (1,4) Restore the value of B
     RTS
                          ; (1,5) A holds the new 8-bit random number
DelavX: Delay based on the value in.
DelayX:
     PSHX
LoopX:
     DEX
     BNE
          LoopX
    PULX
     RTS
```

## Conclusion

 $\label{eq:stability} Iwasabletodevelopagoodrandomnumbergeneratorforthe68HC11thatwasquite small(25bytes)andwhichmetallofmygoodnesscriteria. This wasanequation -based algorithmusing the quation S=S*181+359, with the returned random number the top 8-bits of the 16 -bit Svalue. Atable -based algorithm was also developed, which also generated agood sequence of random numbers, but the algorithm and table required 190 bytes of programs pace, much larger than the 25 bytes of the equation -based algorithm. The resulting 68 HC11 code was successfully test edon an HC11E1 system.$ 

# AboutTheAuthor

TomDickensisanengineerandAssociateTechnicalFellowatTheBoeingCompany, specializinginth edesignandimplementationofhardwareandsoftwaresystems.Healso teacheseveningsintheSeattleareaatBoeing,theUniversityofWashington,the UniversityofPhoenix(Washingtoncampus),andHenryCogswellCollege.Tomhas beenanactivememberof theSeattleRoboticsSocietyfor15years,hasbeentheeditorof theSRSnewsletter(theEncoder),hasservedastheSRSsecretary,andhasservedfora numbersofyearsontheSRSboardofdirectorsandtheSRSRobothoncommittee.Tom maintainsaweb -sitededicatedtothe68HC11microcontrollerat:

http://tomdickens.com/6811/intro.html

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