

Angle Recoding 2. Wu

3. MVR

20180925 Tue

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② MVR (Modified Vector Rotational)

two modifications

① repetition of elementary angles

each micro-rotation of elementary angle
can be performed repeatedly

- more possible combinations
- smaller ξ_m

② Confinement of total micro-rotation number

confine the iteration number
in the micro-rotation phase
to R_m ($R_m \ll W$)

The role of R_m is quite similar
to the number of non-zero digit
 N_D in CSD recoding scheme

MVR

* repetition of elementary angles

$$\theta = \alpha(2) * \lambda = \tan^{-1}(2^2) * 2$$

AR

$$+ \alpha(1) + \alpha(5) - \alpha(8) - \alpha(10) - \alpha(14)$$



```
k= 0 theta= 0.4899573 ik= 1 uik=+1 a[ 1]= 0.4636476, new theta= 0.0263097  
k= 1 theta= 0.0263097 ik= 5 uik=+1 a[ 5]= -0.0312398, new theta= -0.0049301  
k= 2 theta=-0.0049301 ik= 8 uik=-1 a[ 8]= 0.0039062, new theta= -0.0010239  
k= 3 theta=-0.0010239 ik=10 uik=-1 a[10]= 0.0009766, new theta= -0.0000473  
k= 4 theta=-0.0000473 ik=14 uik=-1 a[14]= 0.0000610, new theta= 0.0000137
```

Conventional
COORDIC

$$+ \alpha(0) - \alpha(1) + \alpha(2) - \alpha(3) + \alpha(4) - \alpha(5) + \alpha(6) + \alpha(7)$$

$$- \alpha(8) - \alpha(9) - \alpha(10) - \alpha(11) + \alpha(12) + \alpha(13) - \alpha(14) - \alpha(15)$$

```
k= 0 theta= 0.4899573 u[ 0]=+1 a[ 0]= 0.7853982, new theta= -0.2954408  
k= 1 theta= -0.2954408 u[ 1]=-1 a[ 1]= 0.4636476, new theta= 0.1682068  
k= 2 theta= 0.1682068 u[ 2]=+1 a[ 2]= 0.2449787, new theta= -0.0767719  
k= 3 theta= -0.0767719 u[ 3]=-1 a[ 3]= 0.1243550, new theta= 0.0475831  
k= 4 theta= 0.0475831 u[ 4]=+1 a[ 4]= 0.0624188, new theta= -0.0148357  
k= 5 theta= -0.0148357 u[ 5]=-1 a[ 5]= 0.0312398, new theta= 0.0164041  
k= 6 theta= 0.0164041 u[ 6]=+1 a[ 6]= 0.0156237, new theta= 0.0007804  
k= 7 theta= 0.0007804 u[ 7]=+1 a[ 7]= 0.0078123, new theta= -0.0070319  
k= 8 theta= -0.0070319 u[ 8]=-1 a[ 8]= 0.0039062, new theta= -0.0031257  
k= 9 theta= -0.0031257 u[ 9]=-1 a[ 9]= 0.0019531, new theta= -0.0011726  
k=10 theta= -0.0011726 u[10]=-1 a[10]= 0.0009766, new theta= -0.0001960  
k=11 theta= -0.0001960 u[11]=-1 a[11]= 0.0004883, new theta= 0.0002923  
k=12 theta= 0.0002923 u[12]=+1 a[12]= 0.0002441, new theta= 0.0000481  
k=13 theta= 0.0000481 u[13]=+1 a[13]= 0.0001221, new theta= -0.0000740  
k=14 theta= -0.0000740 u[14]=-1 a[14]= 0.0000610, new theta= -0.0000129  
k=15 theta= -0.0000129 u[15]=-1 a[15]= 0.0000305, new theta= 0.0000176
```

data : w -bit word length

the iteration number : N $N \leq w$

the restricted iteration number : R_m $R_m \ll w$

$$i \in [0, R_m - 1] \quad R_m < W$$

the angle quantization error

$$\xi_{m,MVR} \triangleq \theta - \sum_{i=0}^{R_m-1} \alpha(i) a(s(i))$$

the rotational sequence

$$s(i) \in \{0, 1, \dots, W-1\} \quad * \text{repetition allowed}$$

the micro-rotation angle
in the i -th iteration

the directional sequence

$$\alpha(i) \in \{-1, 0, +1\}$$

the direction of the i -th
micro-rotation of $a(s(i))$

$$\alpha(i) a(s(i)) = \tilde{\theta}(i)$$

the rotational sequence

$$s(i) \in \{0, 1, \dots, w-1\} \quad [0, 3, 6, 7]$$

the directional sequence

$$\alpha(i) \in \{-1, 0, +1\} \quad [1, -1, -1, 1]$$

$$\text{atan}(2^0) - \text{atan}(2^{-3}) - \text{atan}(2^{-6}) + \text{atan}(2^{-7})$$

$$\alpha(i) \alpha(s(i)) = \tilde{\theta}(j)$$

MVR-CORDIC Algorithm with $R_n = 4$	Greedy Algorithm	3	$\bar{a} = [1 \ -1 \ -1 \ -1]$ $\bar{s} = [0 \ 3 \ 6 \ 7]$	5.2891×10^{-4}
	Semi-greedy Algorithm ($D = 2$)	4	$\bar{a} = [1 \ -1 \ -1 \ 1]$ $\bar{s} = [0 \ 3 \ 5 \ 7]$	5.2033×10^{-4}
	TBS Algorithm	5	$\bar{a} = [1 \ 1 \ -1 \ -1]$ $\bar{s} = [1 \ 2 \ 4 \ 7]$	2.5911×10^{-4}

```
>> s = [0, 3, 6, 7]
>> alpha = [1, -1, -1, -1]
>> sum(atan(2.^(-s) .* alpha))
ans = 0.63761
>>
```

```
>> s = [0, 3, 5, 7]
>> alpha = [1, -1, -1, 1]
>> sum(atan(2.^(-s) .* alpha))
ans = 0.63762
```

```
>> alpha = [1, 1, -1, -1]
>> s = [1, 2, 4, 7]
>> sum(atan(2.^(-s) .* alpha))
ans = 0.63840
```

0	7.85398163397448e-01
1	4.63647609000806e-01
2	2.44978663126864e-01
3	1.24354994546761e-01
4	6.24188099959574e-02
5	3.12398334302683e-02
6	1.56237286204768e-02
7	7.81234106010111e-03
8	3.90623013196697e-03
9	1.95312251647882e-03
10	9.76562189559319e-04
11	4.88281211194898e-04
12	2.44140620149362e-04
13	1.22070311893670e-04
14	6.10351561742088e-05
15	3.05175781155261e-05

0	7.85398163397448e-01	s(0)
1	4.63647609000806e-01	.
2	2.44978663126864e-01	.
3	1.24354994546761e-01	s(1)
4	6.24188099959574e-02	.
5	3.12398334302683e-02	.
6	1.56237286204768e-02	s(2)
7	7.81234106010111e-03	s(3)
8	3.90623013196697e-03	.
9	1.95312251647882e-03	.
10	9.76562189559319e-04	.
11	4.88281211194898e-04	.
12	2.44140620149362e-04	.
13	1.22070311893670e-04	.
14	6.10351561742088e-05	.
15	3.05175781155261e-05	.

$W=16$

0	7.85398163397448e-01	s(0)
3	1.24354994546761e-01	s(1)
6	1.56237286204768e-02	s(2)
7	7.81234106010111e-03	s(3)

$R_m = 4$

$s(0)=0$
 $s(1)=3$
 $s(2)=6$
 $s(3)=7$

0	7.85398163397448e-01
1	4.63647609000806e-01
2	2.44978663126864e-01
3	1.24354994546761e-01
4	6.24188099959574e-02
5	3.12398334302683e-02
6	1.56237286204768e-02
7	7.81234106010111e-03
8	3.90623013196697e-03
9	1.95312251647882e-03
10	9.76562189559319e-04
11	4.88281211194898e-04
12	2.44140620149362e-04
13	1.22070311893670e-04
14	6.10351561742088e-05
15	3.05175781155261e-05

AQ & MVR CORDIC

$$\xi_{m, \text{MVR}} \triangleq \theta - \left[\sum_{j=0}^{R_m-1} \alpha(j) a(s(j)) \right]$$

the rotational sequence $s(j)$

$$j = 0, 1, 2, \dots, R_m-1$$

\downarrow

$$s(j) \in \{0, 1, \dots, W-1\} \quad \text{rotational sequence}$$

determines the micro-rotation angle $a(s(j))$
in the j -th iteration

the directional sequence $\alpha(j)$
 $\alpha(j) \in \{-1, 0, +1\}$

controls the direction of the j -th
micro-rotation of $a(s(j))$

$$\alpha(j) a(s(j)) = \tilde{\theta}(j)$$

$$i = 0, 1, 2, 3, \dots, W-1$$

$$s(j) = 0, 1, 2, 3, \dots, W-1 \quad \text{rotational sequence}$$

$$\alpha(j) = -1, 0, +1, \dots, -1 \quad \text{directional sequence}$$

$$j = 0, 1, 2, \dots, R_m-1 \quad \text{effective iteration number}$$

$$R_m \ll W$$

i	j	$S(j)$	
0	0	$S(0) = 0$	
1			
2			
3			
4	1, 4	$S(1) = 4, S(4) = 4$	repetition allowed
5	2	$S(2) = 5$	
6			
7	3	$S(3) = 8$	
8			
9			
10			
11			
12			
13			
14			
$W-1 = 15$			

effective iteration number

rotational sequence

conventional

i C O R D I C

$$S(0) = 0$$

$$S(1) = 1$$

$$S(2) = 2$$

$$S(3) = 3$$

$$S(4) = 4$$

$$S(5) = 5$$

$$S(6) = 6$$

$$S(7) = 7$$

$$S(8) = 8$$

$$S(9) = 9$$

$$S(10) = 10$$

$$S(11) = 11$$

$$S(12) = 12$$

$$S(13) = 13$$

$$S(14) = 14$$

$$S(15) = 15$$

j S(j)

$$S(0) = 0$$

$$S(1) = 1$$

$$S(2) = 2$$

$$S(3) = 3$$

$$S(4) = 4$$

$$S(5) = 5$$

$$S(6) = 6$$

$$S(7) = 7$$

$$S(8) = 8$$

$$S(9) = 9$$

$$S(10) = 10$$

$$S(11) = 11$$

$$S(12) = 12$$

$$S(13) = 13$$

$$S(14) = 14$$

$$S(15) = 15$$

effective
iteration
number

rotational
sequence

W-1 =

$$\text{sub-angle } (\alpha(j) \alpha(s(j))) \sim \tilde{\theta}(j)$$

$$\begin{aligned}\xi_{m,AR} &= \theta - \left[\sum_{j=0}^{N-1} \tilde{\theta}(j) \right], \quad \tilde{\theta}(j) = \tan^{-1}(\alpha(j) \cdot 2^{-s(j)}) \\ &= \theta - \left[\sum_{j=0}^{N-1} \tan^{-1}(\alpha(j) \cdot 2^{-s(j)}) \right]\end{aligned}$$

$$N' \triangleq \sum_{j=0}^{N-1} |\mu(j)| \quad \text{the effective iteration number}$$

EAS formed by MVR-CORDIC
is the same as AR
also performs AQ

the EAS consists of all possible values of $\tilde{\theta}(j)$

the EAS S_1 in AR

$$S_1 = \left\{ \tan^{-1}(\alpha^* \cdot 2^{-s^*}) : \alpha^* \in \{-1, 0, +1\}, s^* \in \{0, 1, \dots, N-1\} \right\}$$

The major difference

1) the total number of sub-angles N_A

the total iteration number

in the micro-rotation phase

is kept fixed to a pre-defined value of R_m

$$N_A = R_m$$

2) the sub-angle θ_i corresponds to $\alpha^{(j)} \alpha(s(j))$

$$\theta_j = \alpha^{(j)} \alpha(s(j)) = \tilde{\theta}_j$$

MVR (Modified Vector Rotation)

1) Repeat of Elementary Angles

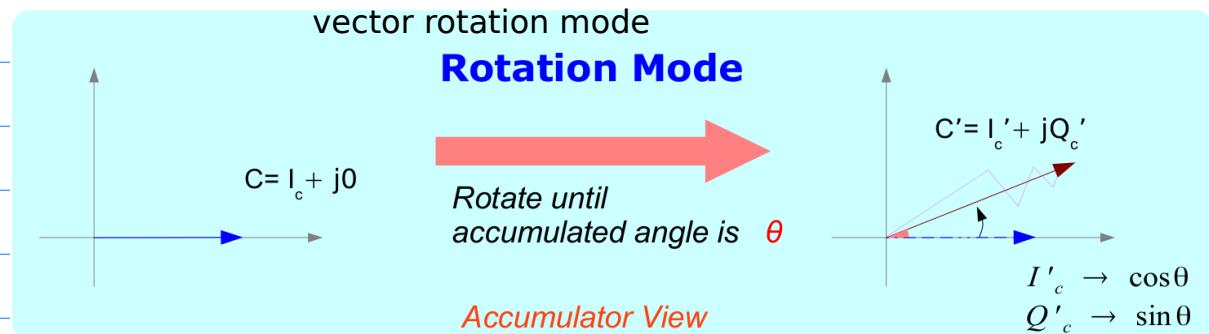
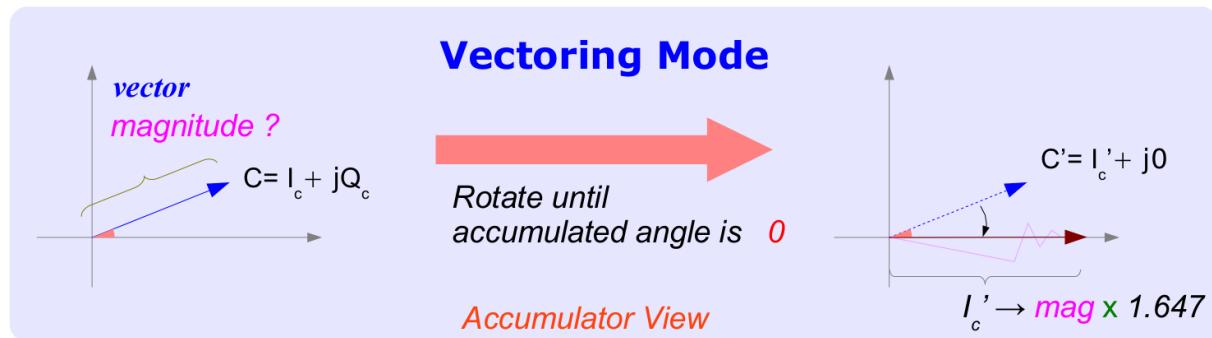
θ_i, θ_i

2) fixed total micro-rotation Number

R_m

* Vector Rotation Mode

* and the rotation angles are known in advance



Modified Vector Rotational MVR CORDIC

- reduce the iteration number
- maintaining the SQNR performance
- modifying the basic microrotation procedure

Three Searching Algorithm

- ① the selective preotation
- ② the selective scaling
- ③ iteration - tradeoff scheme

Optimization Problem

EAS point of view

Given θ , find the combination of R_m elementary angles from EAS S_1 , such that the angle quantization error $|\xi_{m, \text{MUR}}|$ is minimized.

Semi-greedy algorithm
trade offs between computational complexities
and performance

key issue in the MVR-CORDIC
is to find the best sequences of
 s_{ci} and α_{ci} to minimize $|\xi_m|$
subject to the constraint that
the total iteration number is confined to R_m

- 1) Greedy Algorithm
- 2) Exhaustive Algorithm
- 3) Semigreedy Algorithm

data : w -bit word length

the iteration number : N $N \leq w$

the restricted iteration number : R_m $R_m \ll w$

Hu's greedy algorithm

$$\Theta(0) = \Theta, \{ \mu(i) = 0, 0 \leq i \leq N-1 \}, k=0 .$$

repeat until $|\Theta(k)| < \alpha(N-1)$ Do

choose i_k , $0 \leq i_k \leq N-1$

$$||\Theta(k)| - \alpha(i_k)|| = \min_{0 \leq i \leq N-1} ||\Theta(k)| - \alpha(i_k)||$$

$$\Theta(k+1) = \Theta(k) - \mu(i_k) \alpha(i_k)$$

$$\mu(i_k) = \text{Sign}(\Theta(k))$$

$$\mathcal{J}(i) = |\Theta(i) - \alpha(i) \alpha(s(i))| \text{ is minimized}$$

I) Greedy Algorithm

given Θ , W , R_m

try to approach the target rotation angle, Θ , step by step
in each step, decisions are made on $\alpha(i)$ and $s(i)$
by choosing the best combination of $\alpha(i) a(s(i))$
so as to minimize $|\xi_m|$

$\alpha(i)$ and $s(i)$ are determined such that

the error function $J(i) = |\Theta(i) - \alpha(i) a(s(i))|$ is minimized

$\Theta(i)$: the residue angle in the i -th step

$$\Theta(i) = \Theta - \sum_{m=0}^{i-1} \alpha(m) a(s(m))$$

the searching is terminated

{ if no further improvements can be found
 $J(i) \geq J(i-1)$
if the iteration (i) reaches $R_m - 1$

$\alpha(R_m - 1)$ and $s(R_m - 1)$ are determined
at the end of the searching

the greedy algorithm terminates

Only when the residue angle error
cannot be further reduced.

Initialization:

given Θ angle

W wordlength

R_m restricted iteration number

$$\Theta(i) = \Theta - \sum_{m=0}^{i-1} \alpha(m) a(s(m))$$

Select $\alpha(i) \in \{-1, 0, +1\}$

$s(i) \in \{0, 1, 2, \dots, W-1\}$

to minimize $J(i) = \Theta(i) - \alpha(i) a(s(i))$

$s(m)$ repetition allowed

N

$J(i) < J(i-1)$

Y

$i = i + 1$

$$\Theta(i+1) = \Theta(i) - \alpha(i) a(s(i))$$

Store $\alpha(i)$ and $s(i)$

$i \geq R_m - 1$

N

Y

2) Exhaustive Algorithm

Search for the entire solution space

$$\begin{array}{c} \alpha(i) \\ \{-1, 0, +1\} \\ 3 \end{array} \times \begin{array}{c} a(s(i)) \\ \{s(0), s(1), \dots, s(w-1)\} \\ w \end{array} \times \begin{array}{c} i \\ \{0, 1, \dots, R_m-1\} \\ R_m \end{array} \Rightarrow (3w)^{R_m}$$

all possible combinations of

$$\sum_{i=0}^{R_m-1} \alpha(i) a(s(i))$$

in a single step

decisions for $\alpha(i)$ and $s(i)$, $0 \leq i \leq R_m-1$

by minimizing the error function

$$J = \left| \theta - \sum_{i=0}^{R_m-1} \alpha(i) a(s(i)) \right|$$

global optimal solution

Initialization:

Given Θ , W , R_m

let $\Theta(0) = \Theta$,

$i = 0$

$J(-1) = \infty$



Select $\alpha(i) \in \{-1, 0, +1\}$

$s(i) \in \{0, 1, 2, \dots, W-1\}$

for $0 \leq i \leq R_m - 1$

to minimize $J(i) = \Theta - \sum_{l=0}^{R_m-1} \alpha(i) \alpha(s(i))$

$$(3 \cdot W) \cdot (3 \cdot W) \cdots (3 \cdot W) \\ = 3^{R_m} \cdot W^{R_m}$$



Store $\alpha(i)$ and $s(i)$

for $0 \leq i \leq R_m - 1$

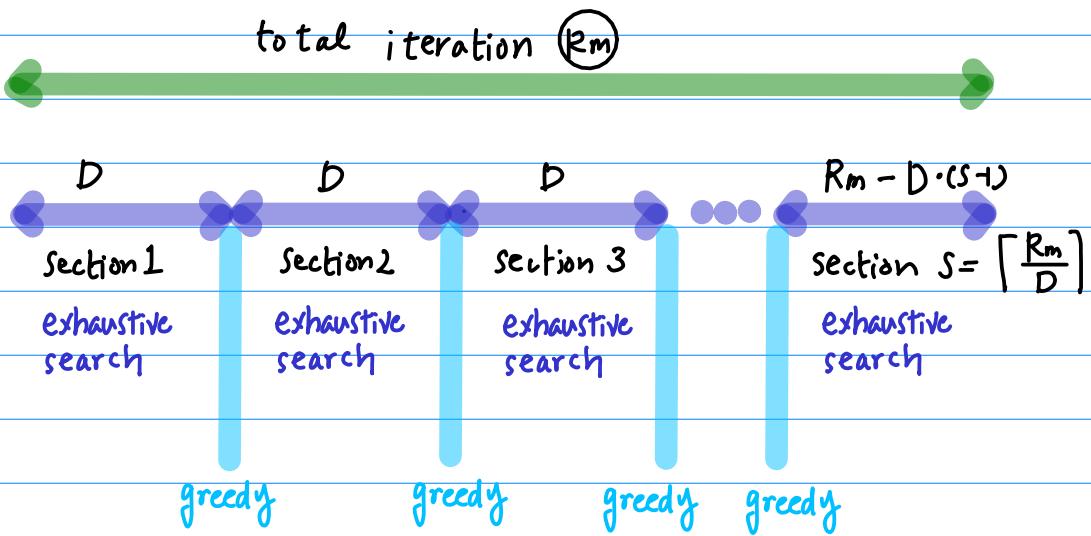
3) Semi-greedy Algorithm

a combination of greedy and exhaustive algorithm

the search space of $\alpha(i)$ and $s(i)$ for $0 \leq i \leq R_m - 1$
are divided into several sections

with D iterations as a segment
 \downarrow block length \downarrow block

the segmentation scheme



in the i -th block

decision of $\alpha(k)$ and $s(k)$ for $iD \leq k \leq (i+1)D-1$

minimizes $J = \left| \theta(i) - \sum_{k=iD}^{(i+1)D-1} \alpha(k) \alpha(s(k)) \right|$

where $\theta(i) = \theta - \sum_{m=0}^{i-1} \left[\sum_{k=mD}^{(m+1)D-1} \alpha(k) \alpha(s(k)) \right]$

the residue angle in the i -th step

$$s = \left\lceil \frac{R_m}{D} \right\rceil$$

$$\theta(i) = \theta - \left[\sum_{k=0D}^{1D-1} \alpha(k) \alpha(s(k)) + \sum_{k=1D}^{2D-1} \alpha(k) \alpha(s(k)) + \dots + \sum_{k=2D}^{S \cdot D-1} \alpha(k) \alpha(s(k)) \right]$$

Initialization:

given Θ , W , R_m

let $\Theta(0) = \Theta$,

$i = 0$

$J(-1) = \infty$

Select $\alpha(k) \in \{-1, 0, +1\}$

$s(k) \in \{0, 1, 2, \dots, W-1\}$

for $iD \leq k \leq (i+1)D - 1$

to minimize $J(i) = \Theta(i) - \sum_{k=iD}^{(i+1)D-1} \alpha(k) \alpha(s(k))$

N

$J(i) < J(i-1)$

$i = i + 1$

$\Theta(i+1) = \Theta(i) - \sum_{k=iD}^{(i+1)D-1} \alpha(k) \alpha(s(k))$

Store $\alpha(k)$, $s(k)$

$i \geq \lceil \frac{R_m}{D} \rceil - 1$

N

Y

```

#include <stdio.h>
#include <math.h>

#define N 16
#define Rm 8

//-----
// conventional cordic
// input initial angle : angle[0]
// ouput residue angles : angle[1] .. angle[N]
//-----
void conventional_cordic(double a[], double angle[]) {
    int k = 0, uk;

    printf("* conventional cordic ... \n");

    for (k=0; k<N; ++k) {
        uk = (angle[k] >= 0) ? +1 : -1;

        printf("k=%2d angle[%2d]=%10.7f ", k, k, angle[k]);
        printf("uk=%+d ", uk);

        angle[k+1] = angle[k] - uk * a[k];

        printf("a[%2d]=%10.7f ", k, a[k]);
        printf("angle[%2d]=%10.7f \n", k, angle[k+1]);
    }
}

int find_multiple_angle(double a[], double angle[], int k) {
    int m, mm;
    double residue, minval;

    mm = -1;
    minval = 1e+100;
    for (m=1; m+k<N; ++m) {
        if (angle[k] > 0) residue = angle[k] - (m+1)*a[k+m];
        else             residue = angle[k] + (m+1)*a[k+m];

        if (fabs(angle[k+m+1]) >= fabs(residue)) {
            if (minval >= fabs(residue)) {
                minval = fabs(residue);
                mm = m;
            }
        }
    }
    return mm;
}

```

```

int main(void) {
    double a[N], angle[N+1];
    double theta; // = 4*atan(pow(2,-5));

    int uk;
    int k, i, s;
    int m, km;

    for (i=0; i<N; ++i) {
        a[i] = atan(1./pow(2, i));
    }

for (s=2; s<3; ++s) {

    angle[0] = theta = s * atan(pow(2, -6));

    printf("theta= %2d * atan(pow(2,-6) = %10g \n", s, theta);

    conventional_cordic(a, angle);

    for (k=0; k<N; ++k) {
        uk = (theta >= 0) ? +1 : -1;

        m = find_multiple_angle(a, angle, k);

        if (m > 0) {
            km = k+m;
            for (i=0; i<=m; i++) {
                printf("k=%2d theta=%10.7f ", k, theta);
                printf("i=%2d uk=%+d m=%2d ", i, uk, m-i);

                theta = theta - uk * a[km];
                k++;

                printf("a[%2d]=%10.7f, new theta=%10.7g \n", km, a[km], theta);
            }
            k--;
        } else {
            printf("k=%2d theta=%10.7f ", k, theta);
            printf("i=%2d uk=%+d m=%2d ", 0, uk, m);

            theta = theta - uk * a[k];

            printf("a[%2d]=%10.7f, new theta=%10.7g \n", k, a[k], theta);
        }
    }
}
}

```

```

theta= 2 * atan(pow(2, -6) = 0.0312475
* conventional cordic ...
k= 0 angle[ 0]= 0.0312475 uk=+1 a[ 0]= 0.7853982 angle[ 0]=-0.7541507
k= 1 angle[ 1]=-0.7541507 uk=-1 a[ 1]= 0.4636476 angle[ 1]=-0.2905031
k= 2 angle[ 2]=-0.2905031 uk=-1 a[ 2]= 0.2449787 angle[ 2]=-0.0455244
k= 3 angle[ 3]=-0.0455244 uk=-1 a[ 3]= 0.1243550 angle[ 3]= 0.0788306
k= 4 angle[ 4]= 0.0788306 uk=+1 a[ 4]= 0.0624188 angle[ 4]= 0.0164118
k= 5 angle[ 5]= 0.0164118 uk=+1 a[ 5]= 0.0312398 angle[ 5]=-0.0148281
k= 6 angle[ 6]=-0.0148281 uk=-1 a[ 6]= 0.0156237 angle[ 6]= 0.0007956
k= 7 angle[ 7]= 0.0007956 uk=+1 a[ 7]= 0.0078123 angle[ 7]=-0.0070167
k= 8 angle[ 8]=-0.0070167 uk=-1 a[ 8]= 0.0039062 angle[ 8]=-0.0031105
k= 9 angle[ 9]=-0.0031105 uk=-1 a[ 9]= 0.0019531 angle[ 9]=-0.0011573
k=10 angle[10]=-0.0011573 uk=-1 a[10]= 0.0009766 angle[10]=-0.0001808
k=11 angle[11]=-0.0001808 uk=-1 a[11]= 0.0004883 angle[11]= 0.0003075
k=12 angle[12]= 0.0003075 uk=+1 a[12]= 0.0002441 angle[12]= 0.0000634
k=13 angle[13]= 0.0000634 uk=+1 a[13]= 0.0001221 angle[13]=-0.0000587
k=14 angle[14]=-0.0000587 uk=-1 a[14]= 0.0000610 angle[14]= 0.0000023
k=15 angle[15]= 0.0000023 uk=+1 a[15]= 0.0000305 angle[15]=-0.0000282
k= 0 theta= 0.0312475 i= 0 uk=+1 m=-1 a[ 0]= 0.7853982, new theta=-0.7541507
k= 1 theta=-0.7541507 i= 0 uk=-1 m=-1 a[ 1]= 0.4636476, new theta=-0.2905031
k= 2 theta=-0.2905031 i= 0 uk=-1 m= 1 a[ 3]= 0.1243550, new theta=-0.1661481
k= 3 theta=-0.1661481 i= 1 uk=-1 m= 0 a[ 3]= 0.1243550, new theta=-0.04179311
k= 4 theta=-0.0417931 i= 0 uk=-1 m=-1 a[ 4]= 0.0624188, new theta= 0.0206257
k= 5 theta= 0.0206257 i= 0 uk=+1 m= 3 a[ 8]= 0.0039062, new theta=0.01671947
k= 6 theta= 0.0167195 i= 1 uk=+1 m= 2 a[ 8]= 0.0039062, new theta=0.01281324
k= 7 theta= 0.0128132 i= 2 uk=+1 m= 1 a[ 8]= 0.0039062, new theta=0.008907012
k= 8 theta= 0.0089070 i= 3 uk=+1 m= 0 a[ 8]= 0.0039062, new theta=0.005000781
k= 9 theta= 0.0050008 i= 0 uk=+1 m=-1 a[ 9]= 0.0019531, new theta=0.003047659
k=10 theta= 0.0030477 i= 0 uk=+1 m= 1 a[11]= 0.0004883, new theta=0.002559378
k=11 theta= 0.0025594 i= 1 uk=+1 m= 0 a[11]= 0.0004883, new theta=0.002071096
k=12 theta= 0.0020711 i= 0 uk=+1 m=-1 a[12]= 0.0002441, new theta=0.001826956
k=13 theta= 0.0018270 i= 0 uk=+1 m=-1 a[13]= 0.0001221, new theta=0.001704886
k=14 theta= 0.0017049 i= 0 uk=+1 m= 1 a[15]= 0.0000305, new theta=0.001674368
k=15 theta= 0.0016744 i= 1 uk=+1 m= 0 a[15]= 0.0000305, new theta=0.00164385

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