***Homework 2 - Inverse Kinematics***

*for*

***A Robot Arm with Hand to reach a target***

One of the tasks my team wants for our Robot Arm with Hand is to be able to shake hands with someone standing in front of it. Our plan is to place a proximity sensor on the hand and use a search method to discover objects in front of it. When the search is complete, we have arrived at our target (a human hand extended to shake). Homework 1 used a Law of Cosines approach to calculate the angles needed for a 2 DOF robot arm. This homework builds on these results and compares a random search approach to finding the angles for placing the end of the robot arm at the target.

**Law of Cosines Inverse Kinematics**

As show in Homework 1, by simplifying the model to 2 DOF (shoulder, elbow) we are able to bypass some of the trickier math involving transforms and translations and use the Law of Cosines with a simple triangle to find the angles needed to place the hand at the target. Here is an example of the Arm:



And here is the geometric representation with a desired target:



From the Law of Cosines we have:

$$c^{2}=a^{2}+ b^{2}- 2ab cos⁡(c)$$

Inserting our values and solving for the desired angles gives us:

$$θ\_{2}=arccos\left[\frac{x^{2}+ y^{2}- L\_{1}^{2}-L\_{2}^{2}}{2L\_{1}L\_{2}}\right]$$

And

$$θ\_{1}= arcsin\left[\frac{L\_{2}sin⁡(θ\_{2})}{\sqrt{x^{2}+ y^{2}}}\right]+ arctan2\left(\frac{x}{y}\right)$$

Using the above angle formulas and some input values, a simple c++ program was written to calculate the output angles needed to place the hand at the target. Below is an example of the program running. Note that the angle calculations are converted from radians to degrees by multiplying by ($^{180}/\_{π})$. When the function is integrated into our control program the degrees will be converted to the pulse width necessary to move the servos to the desired angles.

**Random Angle Method**

The Law of Cosines Inverse Kinematic method will be compared with a Random Angle method. This idea is to use a random angle change to reduce the error between where the robot is and the target. If the new angles do not move the hand closer to the target they are discarded and new random angles are generated.

Algorithm for Random Angle method:

* While (hand > (some close value to target) –OR– (count is too high))
	+ Increment count
	+ Move servo 1 small random distance
	+ Move servo 2 small random distance
	+ If hand is farther from goal
		- Discard new angle values
	+ Else
		- Copy new angle values

The problem with the above algorithm is there are possible positions where one servo would need to move away from the target to allow another servo to move. This is an example of local minima prohibiting the realization of a goal. I had to simplify the model and create a mathematical method (such as a computer program) to input the values and output the solution.

This is the code for the above program:

#include <iostream>

#include <string>

#include <math.h>

#include "StdAfx.h"

using namespace std;

//#define RAD\_CONVERT = 57.296 //(180 / pi)

// Three points in space: base, elbow, hand

int \_points\_x [3] = {0, 0, 0};

int \_points\_y [3] = {0, 500, 1000};

// Random points in space: base, elbow, hand for randSearch algorithm

int \_points\_x\_rand [3] = {0, 0, 0};

int \_points\_y\_rand [3] = {0, 500, 1000};

//private var \_distance:Number = 100;

int \_distance = 1000; //millimeters, assume 3ft arm reach and (upper arm)=(lower arm)=500 mm

int \_distance\_rand = 1000;

//private var \_lastPoint:int = 2;

int \_lastPoint = 2;

int randIter = 0;

// Give a starting target in millimeters

float targetX = 250;

float targetY = 250;

//arm angles start straight, ie hanging down

float \_theta1 = 0;

float \_theta2 = 0;

//Random Angle variables

//arm angles start straight, ie hanging down

float \_theta1\_rand = 0;

float \_theta2\_rand = 0;

float RAD\_CONVERT = (180 / 3.14159);

void randSearch()

{

 // Compute difference between start and end points

 float dx = (targetX - \_points\_x\_rand[0]);

 float dy = (targetY - \_points\_y\_rand[0]);

 cout << "Difference of (x, y): " << "(" << dx << ", " << dy << ")" << endl;

 // Compute distance between start and end points

 float dist = sqrt(dx\*dx + dy\*dy);

 cout << "Distance from base to target: " << dist << endl;

 // Compute angle between start and end points

 //float theta = atan2(dy,dx);

 float theta = float(rand() % 180 + 1);

 float temp = (int(theta \* RAD\_CONVERT)) % 180;

 cout << "Angle between base and target: " << temp << endl;

 // Clamp the distance

 float totalLength = \_distance\_rand \* 2;

 //if( dist < totalLength ) {

 // Calculate first angle: http://en.wikipedia.org/wiki/Dot\_product#Geometric\_interpretation

 \_theta1\_rand = (int(acos( dist / totalLength ) + theta)) % 180;

 dx = dx - \_distance\_rand \* cos( \_theta1\_rand );

 dy = dy - \_distance\_rand \* sin( \_theta1\_rand );

 // Calculate second angle from first angle and segment

 \_theta2\_rand = atan2(dy, dx);

 //} else {

 // If the distance is greater than arm length, arm is straight

 // \_theta1\_rand = \_theta2\_rand = int(theta);

 //}

}

void solve()

{

 // Compute difference between start and end points

 float dx = (targetX - \_points\_x[0]);

 float dy = (targetY - \_points\_y[0]);

 cout << "Difference of (x, y): " << "(" << dx << ", " << dy << ")" << endl;

 // Compute distance between start and end points

 float dist = sqrt(dx\*dx + dy\*dy);

 cout << "Distance from base to target: " << dist << endl;

 // Compute angle between start and end points

 float theta = atan2(dy,dx);

 float temp = (theta \* RAD\_CONVERT);

 cout << "Angle between base and target: " << temp << endl;

 // Clamp the distance

 float totalLength = \_distance \* 2;

 if( dist < totalLength ) {

 // Calculate first angle: http://en.wikipedia.org/wiki/Dot\_product#Geometric\_interpretation

 \_theta1 = acos( dist / totalLength ) + theta;

 dx = dx - \_distance \* cos( \_theta1 );

 dy = dy - \_distance \* sin( \_theta1 );

 // Calculate second angle from first angle and segment

 \_theta2 = atan2(dy, dx);

 } else {

 // If the distance is greater than arm length, arm is straight

 \_theta1 = \_theta2 = int(theta);

 }

 //cout << "Angle between base and target: " << temp << endl;

/\* // Compute positions from angles

 \_points[1].x = \_points[0].x + Math.cos( \_theta1 ) \* \_distance;

 \_points[1].y = \_points[0].y + Math.sin( \_theta1 ) \* \_distance;

 \_points[2].x = \_points[1].x + Math.cos( \_theta2 ) \* \_distance;

 \_points[2].y = \_points[1].y + Math.sin( \_theta2 ) \* \_distance;

}

\*/

}

int main()

{

 int xVal, yVal;

 cout << "Your initial target coords are (" << targetX << ", " << targetY << ") in mm" << endl;

 cout << "Enter the X-coordinate (mm) for the target:" << endl;

 cin >> xVal;

 cout << "Enter the Y-coordinate (mm) for the target:" << endl;

 cin >> yVal;

 targetX = float(xVal);

 targetY = float(yVal);

 cout << "Your target coordinates are (" << xVal << ", " << yVal << ") in mm" << endl;

 cout << "Your float target coords are (" << targetX << ", " << targetY << ") in mm" << endl;

 solve();

 float temp1 = (\_theta1 \* RAD\_CONVERT);

 float temp2 = (\_theta2 \* RAD\_CONVERT);

 cout << "Theta 1 is " << temp1 << endl << "Theta 2 is " << temp2 << endl;

 // Try random angle measurement, see how many iterations until at previously calculated best angles

 cout << "Random angle search: " << endl;

 randSearch();

 float temp1\_rand = (\_theta1\_rand \* RAD\_CONVERT);

 float temp2\_rand = (\_theta2\_rand \* RAD\_CONVERT);

 cout << "Theta 1 rand is " << temp1\_rand << endl << "Theta 2 rand is " << temp2\_rand << endl;

 cout << "It took " << randIter << " iterations to match calculated IK values" << endl;

}

**Conclusions:**

In this homework I have compared the concepts of inverse kinematics and created a function to solve a 2 DOF robot arm problem. It is definitely more efficient to calculate the values of the angles directly. The calculations from Homework 1 worked much better, however they are limited to 2 DOF in the current implementation. Other methods such as Jacobian, or translation and transformation, are scalable and preferred for multiple DOFs.

**References:**

**Braunl, Thomas, "EMBEDDED ROBOTICS. Mobile Robot Design and Applications with Embedded Systems" Springer, 2008**

**Luger, George, “Artificial Intelligence” Addison-Wesley, 2008**

[www.lorenzonuvoletta.com/tag/inverse-kinematics](http://www.lorenzonuvoletta.com/tag/inverse-kinematics)

<http://generalrobotics.org/ppp/Kinematics_final.ppt>

<http://freespace.virgin.net/hugo.elias/models/m_ik2.htm>