

**ALGORITHM 1**

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#include <algorithm> // Included for std::min.
#include <cmath> // Included for std::sqrt.

double GetTimeToTargetWithMinimalInitialSpeed(double k, double vInfinity,
                                              double rX, double rY) {
    // 1. Start by getting coefficients for the function  $f(t) = a_4 t^4 + a_3 t^3 + a_1 t + a_0$  which is 0 at the sought time-to-target  $t$ . Solving  $f(t) = 0$  for  $t > 0$  is equivalent to solving  $e(u) = f(1/u) u^4 = a_0 u^4 + a_1 u^3 + a_3 u + a_4 = 0$  for  $u$  where  $u = 1/t$ , but the latter is more well-behaved, // being a strictly concave function for  $u > 0$  for any set of valid inputs, // so solve  $e(u)=0$  for  $u$  instead by converging from an upper bound towards
    double kVInfinity = k * vInfinity, m = rX * rX + rY * rY; // the root and
    double a0 = -m, a1 = a0 * k, a3 = k * kVInfinity * rY; // return 1/u.
    double a4 = kVInfinity * kVInfinity;
    double maxInvRelError = 1.0E6; // Use an achievable inverse error bound.
    double de, uDelta = 0;

    // 2. Set  $u$  to an upper bound by solving  $e(u)$  with  $a_3 = a_1 = 0$ , clamped by
    // the result of a Newton method's iteration at  $u = 0$  if positive.
    double u = std::sqrt(kVInfinity / std::sqrt(m));
    if (rY < 0) u = std::min(u, -vInfinity / rY);

    // 3. Let  $u$  monotonically converge to  $e(u)$ 's positive root using a modified
    // Newton's method that speeds up convergence for double roots, but is likely
    // to overshoot eventually. Here, 'e' =  $e(u)$  and 'de' =  $de(u)/du$ .
    for (int it = 0; it < 10; ++it, uDelta = e / de, u -= 1.9 * uDelta) {
        de = a0 * u; e = de + a1; de = de + e; e = e * u;
        de = de * u + e; e = e * u + a3; de = de * u + e; e = e * u + a4;
        if (!(e < 0 && de < 0)) break; // Overshot the root.
    }
    u += 0.9 * uDelta; // Trace back to the unmodified Newton method's output.

    // 4. Continue to converge monotonically from the overestimated  $u$  to  $e(u)$ 's
    // only positive root using Newton's method.
    for (int it = 0; uDelta * maxInvRelError > u && it < 10; ++it) {
        de = a0 * u; e = de + a1; de = de + e; e = e * u;
        de = de * u + e; e = e * u + a3; de = de * u + e; e = e * u + a4;
        uDelta = e / de; u -= uDelta;
    }

    // 5. Return the solved time  $t$  to hit  $[rX, rY]$ , or 0 if no solution exists.
    return u > 0 ? 1 / u : 0;
}

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**ALGORITHM 2**

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#include <algorithm> // Included for std::min.
#include <cmath> // Included for std::sqrt.

double GetTimeToTargetGivenInitialSpeeds(double k, double vInfinity, double rX,
                                          double rY, double s, bool highArc) {
    // 1. Start by getting coefficients for the function  $f(t) = a_4 t^4 + a_3 t^3 + a_2 t^2 + a_1 t + a_0$  which is 0 at the sought time-to-target  $t$ . Solving  $f(t) = 0$  for  $t > 0$  is equivalent to solving  $e(u) = f(1/u) u^3 = a_0 u^3 + a_1 u^2 + a_2 u + a_3 + a_4/u$  for  $u$  where  $u = 1/t$ , but the latter is more // well-behaved, being a strictly convex function for  $u > 0$  for any set of // inputs iff a solution exists, so solve for  $e(u) = 0$  instead by converging // from a high or low bound towards the closest root and return 1/u.
    double kRX = k * rX, kRY = k * rY, kRXsq = kRX * kRX, sS = s * s;
    double twoKVInfinityRY = vInfinity * (kRY + kRY), kVInfinity = k * vInfinity;
    double a0 = rX * rX + rY * rY, a1 = (k + k) * a0;
    double a2 = kRXsq + kRY * kRY + twoKVInfinityRY - sS;
    double a3 = twoKVInfinityRY * k, a4 = kVInfinity * kVInfinity;
    double maxInvRelError = 1.0E6; // Use an achievable inverse error bound.
    double maxV0Sq = sS - kRXsq; // maxV0Sq is the max squared 'v0.y' that leaves
    double e, de, u, uDelta = 0; // enough 'v0.x' to reach rX horizontally.

    // 2. Set  $u$  to a lower/upper bound for the high/low arc, respectively.
    if (highArc) { // Get smallest  $u$  vertically moving rY at max possible +v0.y.
        double minusB = std::sqrt(maxV0Sq) - kRY;
        double determ = minusB * minusB - (twoKVInfinityRY + twoKVInfinityRY);
        u = (kVInfinity + kVInfinity) / (minusB + std::sqrt(determ));
        maxInvRelError = -maxInvRelError; // Convergence over negative slopes.
    } else if (rY < 0) { // Get largest  $u$  vertically moving rY at most neg. v0.y.
        double minusB = -std::sqrt(maxV0Sq) - kRY;
        double determ = minusB * minusB - (twoKVInfinityRY + twoKVInfinityRY);
        u = (minusB - std::sqrt(determ)) / (rY + rY);
        // Clamp the above bound by the largest  $u$  that reaches rX horizontally.
        u = std::min(s / rX - k, u);
    } else u = s / std::sqrt(a0) - k; // Get the (largest)  $u$  hitting rX
    // horizontally a.s.a.p. while launching in the direction of  $[rX, rY]$ .

    // 3. Let  $u$  monotonically converge to  $e(u)$ 's closest root using a modified
    // Newton's method, almost scaling the delta as if the solution is a double
    int it = 0; // root. Note that 'e' =  $e(u) * u^2$  and 'de' =  $de(u)/du * u^2$ .
    for (; it < 12; ++it, uDelta = e / de, u -= 1.9 * uDelta) {
        de = a0 * u; e = de + a1; de = de + e; e = e * u + a2; de = de * u + e;
        e = e * u + a3; e = (e * u + a4) * u; de = de * u * u - a4;
        if (!(u > 0 && de * maxInvRelError > 0 && e > 0)) break; // Overshot.
    }
    u += 0.9 * uDelta; // Trace back to unmodified Newton method's output.

    // 4. Continue to converge monotonically to  $e(u)$ 's closest root using
    // Newton's method from the last known conservative estimate on the convex
    // function. (Note that in practice,  $u$  will have converged enough in <12
    for (; u > 0 && it < 12; ++it) { // iterations iff a solution does exists.)
        de = a0 * u; e = de + a1; de = de + e; e = e * u + a2; de = de * u + e;
        e = e * u + a3; e = (e * u + a4) * u; de = de * u * u - a4;
        uDelta = e / de; u -= uDelta;
        if (!(de * maxInvRelError > 0)) break; // Wrong side of the convex 'dip'.
        if (uDelta * maxInvRelError < u && u > 0) return 1 / u; // 5a. Found it!
    }

    // 5b. If no solution was found, return 0. This only happens if  $s$  (minus
    // a small epsilon) is too small to have a solution, the target is at the
    return 0; // origin, or the parameters are so extreme they cause overflows.
}

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