

Evolutionary Inverse Kinematics for ROS and MoveIt

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Master Thesis

ROS, MoveIt



Robot Operating System

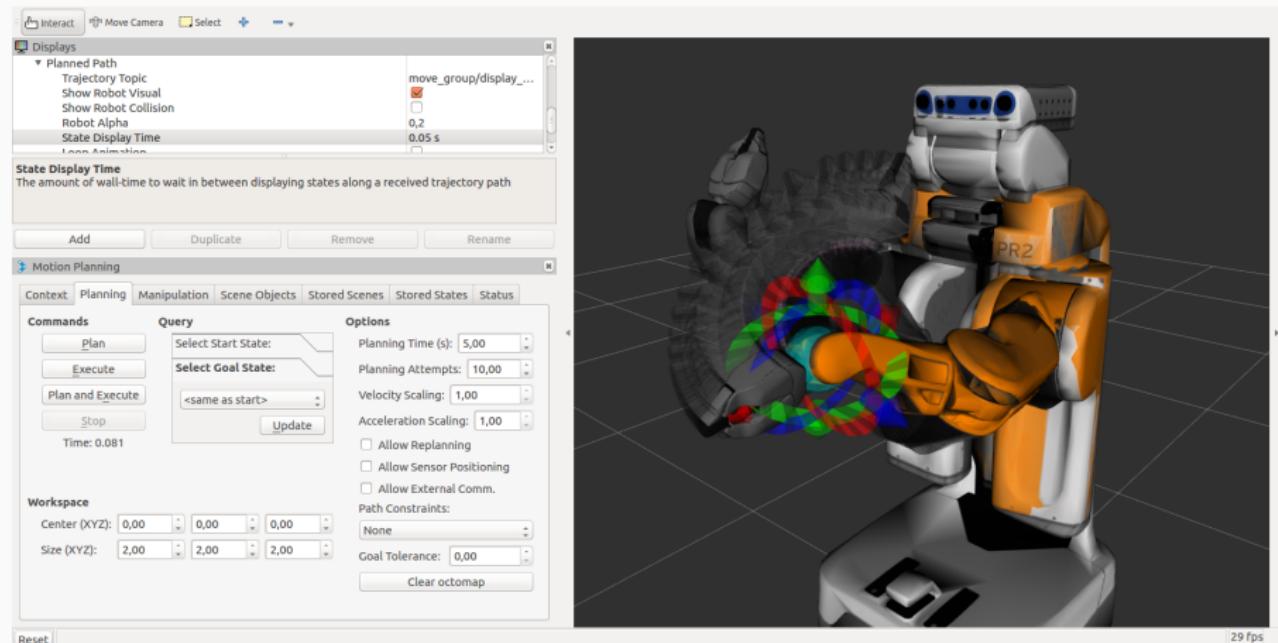


- Forward & inverse kinematics
- Perception
- Motion planning

MoveIt

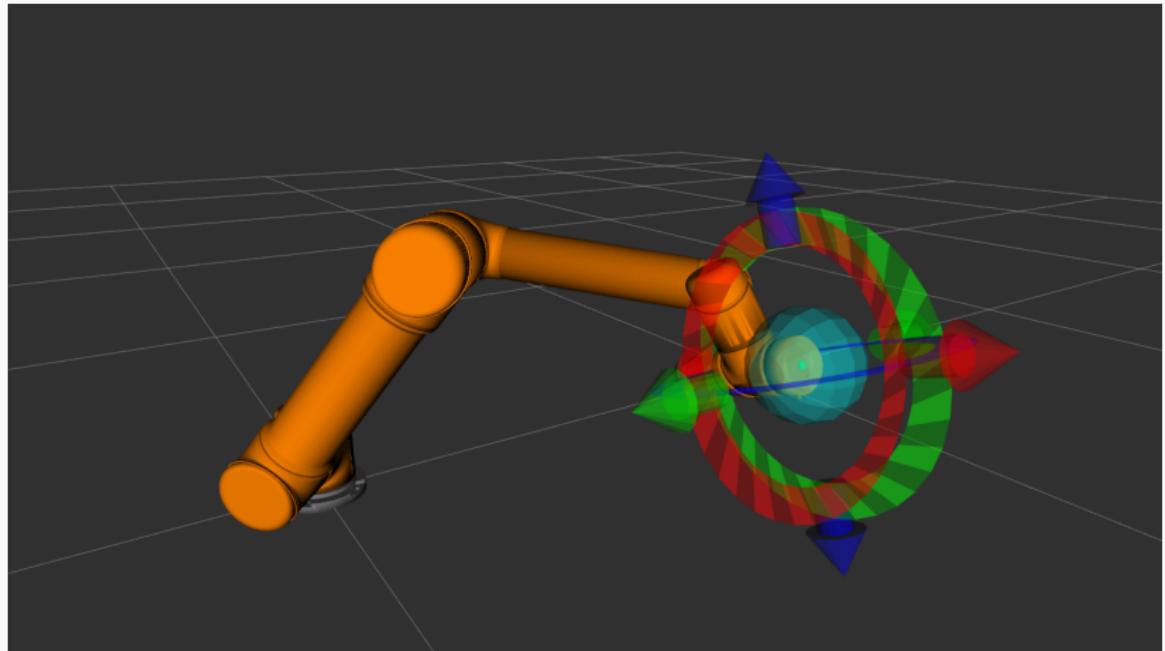
video

Movelt



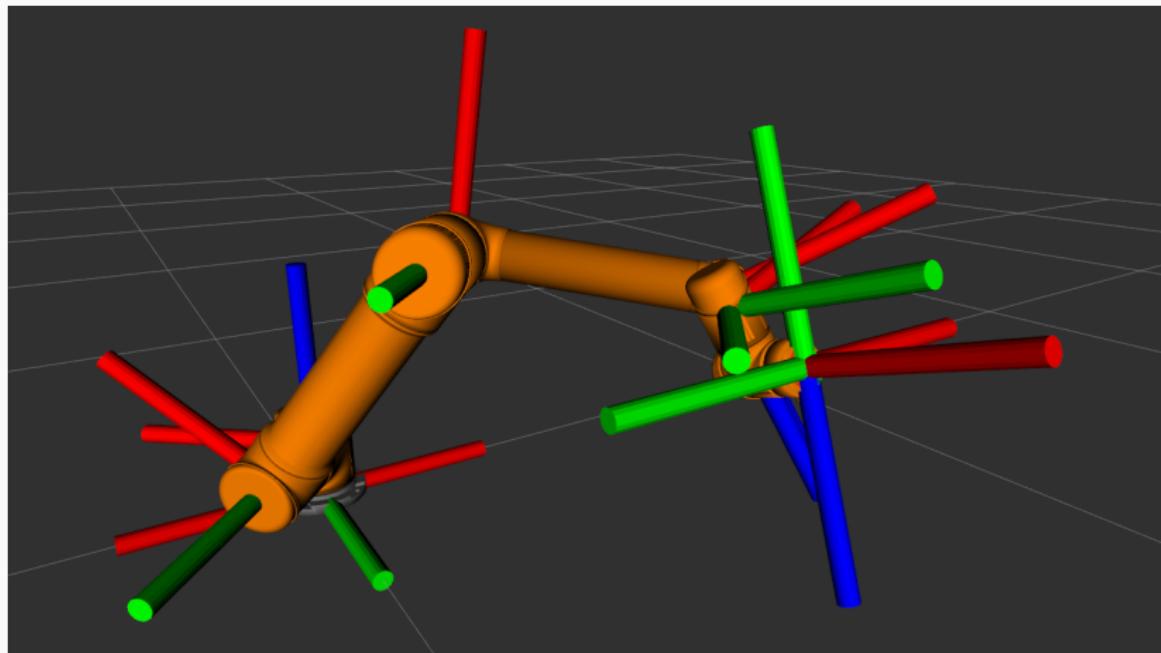
Inverse Kinematics

End effect pose → joint angles



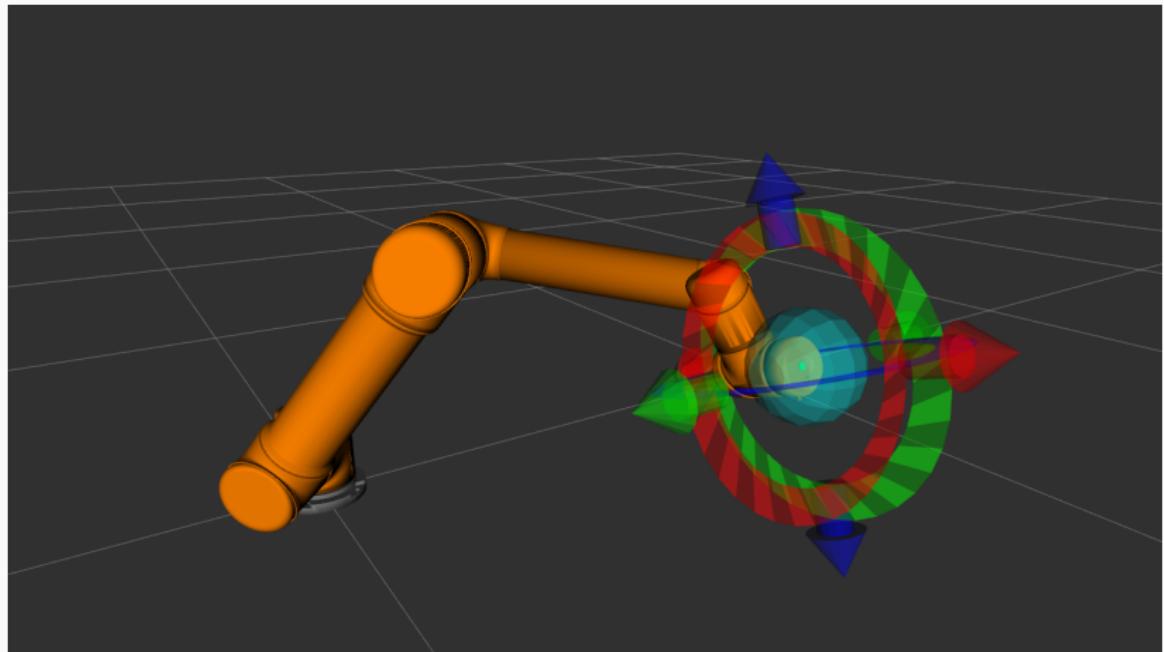
Forward Kinematics

Joint angles → end effect pose



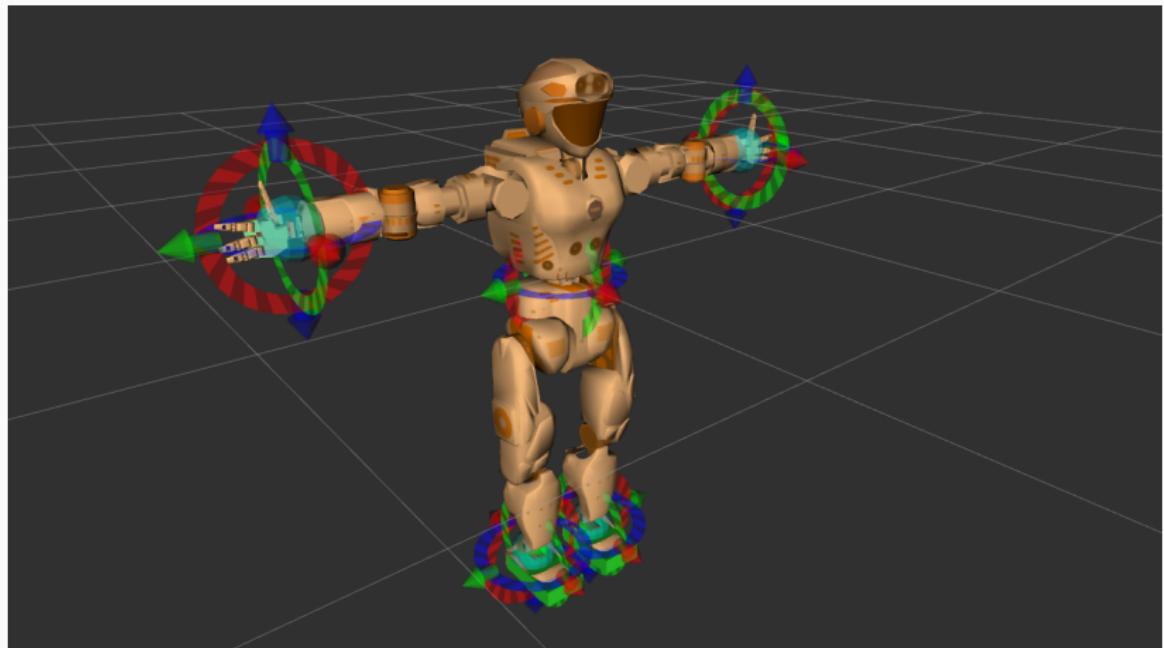
Inverse Kinematics

End effect pose → joint angles

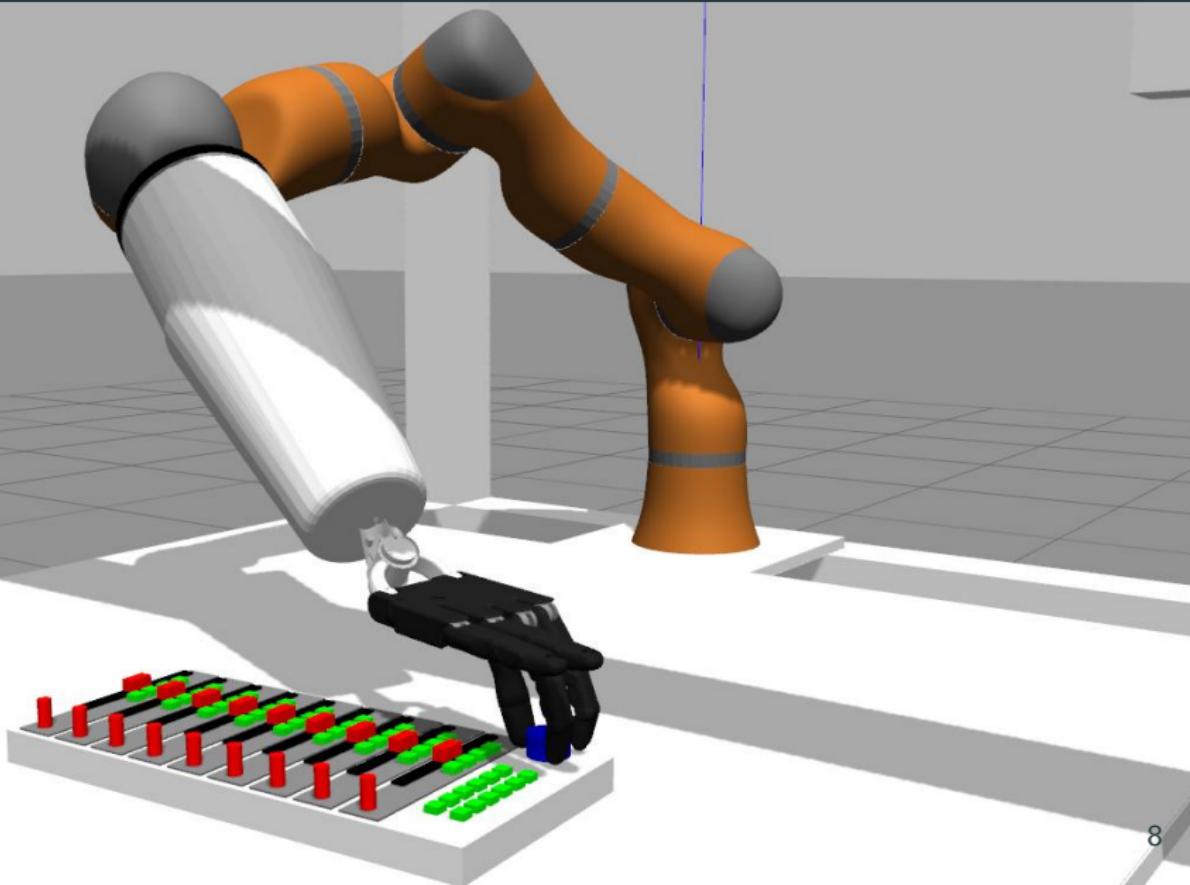


Inverse Kinematics

Multi-Goal IK (multiple eef, or generalized: arbitrary constraints on robot pose)



Inverse Kinematics



IK Solvers in MoveIt

- IK Plugin interface
(kinematics::KinematicsBase)
- Analytic
 - Specialized solvers
 - ikfast
- Numeric
 - KDL
 - TRAC-IK
- Currently single-goal only!

Analytic Solution

- Manually solve equations
- Very fast at runtime
- Only practical for simple cases
- Solver is specific to a certain robot and goal configuration
- New robot or different goal types?
 - write new solver

ikfast

- Generates code automatically
- (Generated solver is specific to a certain robot and goal configuration)
- Only kinematic chains, only some goal types
- Long compilation time
- Doesn't always work

"for 6D ik, there are still mechanisms it cannot solve"
(<http://openrave.org/docs/0.8.2/openravepy/ikfast/>)

$$6D = \text{Translation (3D)} + \text{Orientation (3D)}$$

KDL / Pseudo-Inverse Jacobian

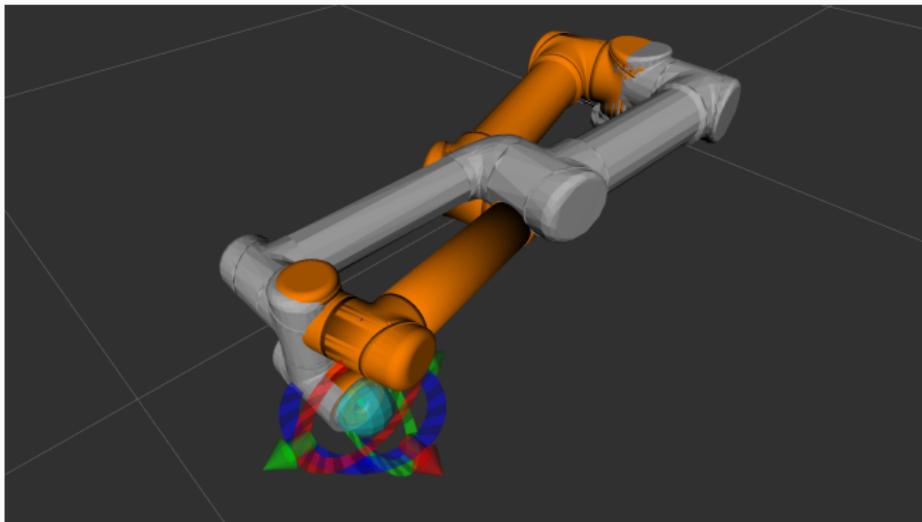
- Pseudo-inverse jacobian
- Minimize tip frame offset

$$\begin{matrix} & X & Y & Z \\ \begin{matrix} j1 \\ j2 \\ j3 \\ \dots \end{matrix} & \left(\begin{matrix} \delta x / \delta j_1 & \delta y / \delta j_1 & \delta z / \delta j_1 & \dots \\ \delta x / \delta j_2 & \delta y / \delta j_2 & \delta z / \delta j_2 & \dots \\ \delta x / \delta j_3 & \delta y / \delta j_3 & \delta z / \delta j_3 & \dots \\ \dots & \dots & \dots & \ddots \end{matrix} \right) \end{matrix}$$

$$\Delta J = -\Delta P * J^{-1}$$

Pseudo-Inverse Jacobian - Issues

- Joint limits
- Possibly degenerate pseudo-inverse
- Local minima



TRAC-IK

- Random restarts if stuck to mitigate local minima
- In parallel: sequential quadratic programming
- Faster than KDL

Inverse Kinematics, MoveIt

Limitations of built-in IK solvers:

- Only single-goal
- Only pose goal or position goal
- Analytic or gradient-based
- Hard to extend

Extensibility & Flexibility:

- Arbitrary cost function
- Minimize cost function

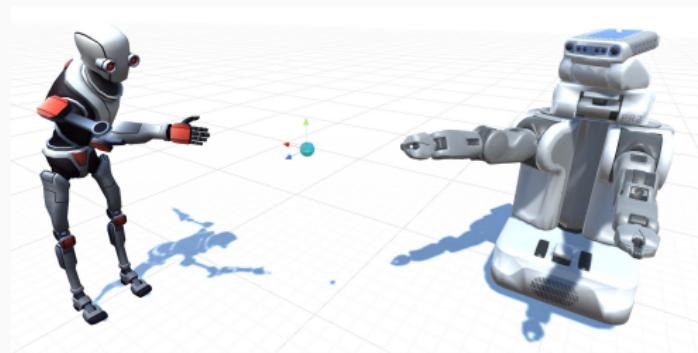
Cost Function

Cost Function

- Sum of errors (eg. square distances)
- Goals: Position, Orientation, Pose, LookAt, MaxDistance, MinDistance, Touch, Line, AvoidJointLimits, MinimalDisplacement, JointVariable, CenterJoints, JointFunction, Balance,

BioIK

BioIK, Sebastian



- Multi-goal, arbitrary cost function
- Biologically Inspired
- Hybrid: Genetic Algorithm
 - + Particle Swarm Optimization
- C#, Unity, Game Development

Genetic algorithm

- Population, individuals, genes
- Genome = joint values
- Mutation, recombination, selection

Particle swarm optimization

- Momentum

BioIK for ROS - Requirements

- C++
- MoveIt
- High Accuracy (10^{-5} , $10\mu m$, $0.0018deg$)
- Fail if unreachable (optional, default)
- Performance (default timeout: 5ms)
- Compatibility: Joint Types, Mimic Joints, etc.

Modified Evolution

Genome and Mutation

Genome

- Genes = Joint positions (eg. joint angles)
- Only active variables
- **double[] genes;**

Mutation

- $r * 2^{-s}$
 - r: random number gene
 - s: random scale per individual, 2^{-s} fast bit-shift
 - Small mutations most likely \Rightarrow high accuracy
- Clipping at joint limits

Genome and Mutation

NOT used:

- **int []** genes;
- Mutation: mutate random bits
- Hamming walls: single mutated bit = 2 to 3 (20 to 21), but not 7 to 8 (0111 to 1000)
- Gray code: conversion overhead
- Different mutation scales per gene
- Floating point numbers & exponential term in mutation function = similar distribution

Fitness & Selection

- IK goals = fitness function
- Compute fitness for each individual
- Select fittest individuals
- Genotype: active joint values (eg. angles)
- Phenotype: end effector link frames (and active joint values)
- Forward kinematics
 - = genotype-phenotype mapping
 - = most computation time
 - focus performance optimization on genotype phenotype mapping / forward kinematics

Secondary Objectives

- Primary and secondary objectives
- eg. primary: position goal, secondary: minimal displacement
- Pre-selection by secondary objectives
Random number of survivors

Wipeouts

- Small mutations most likely
- Might temporarily be stuck at local optimum
- Wipeout if no improvement for a few generations
- Random wipeouts
- Wipeout = random reset of all genes
- Same for all individuals, to prevent premature convergence

Islands / Parallelization

- Multiple islands
- Independent evolution
- One thread per island

Multiple Species

- Problem: wipeout might accidentally destroy best solution
- Two competing species per island
- Only less fit species wiped out

Initialization

- Initial guess from Movelt

Termination

- Timeout
- Good fitness
- Specialized pose comparison for compatibility

Particle Swarm Optimization

- Per-gene momentum
- **struct Individual**

```
{    aligned_vector<double> genes ;    aligned_vector<double> momentum ;}
```
- genes += momentum
- momentum randomly doubled or erased

Memetics

- Evolution + local search
- Tried solvers from CppNumericalSolvers - slower
- Quadratic optimization
- Approximate local fitness landscape by parabola along gradient
- Find extremum & update if better
- Evolution for n generations
- Quadratic optimization, up to m steps

Extrapolated Forward Kinematics

Forward Kinematics

Most computation time spent on computing forward kinematics

- Compute joint transforms
 - Trigonometry, sin/cos
- Concatenate frames
 - Quaternion or matrix operations

Forward Kinematics

Movelt / KDL

- Rotation matrices (3x3 matrix, 3 basis vectors)
- Concatenation:
27 multiplications, 18 additions

BioIK

- Quaternions (4 elements)
- Concatenation:
16 multiplications, 12 additions

Forward Kinematics

Extrapolated Forward Kinematics

- Compute exact forward kinematics only every n steps
- Extrapolate forward kinematics for most mutations
- Rotation: linear, $x = a + b * dj$
- Position: quadratic, $x = a + b * dj + c * dj^2$

Extrapolated Forward Kinematics

Extrapolation

- No trigonometry
- Only active joints, no static or inactive joints, no link frames
- No expensive quaternion-quaternion or matrix-matrix multiplications
 - only addition and multiplication by scalar
- Simple control flow (single loop)
 - Low risk of branch misprediction
 - Pipelining

Extrapolated Forward Kinematics

- FK: analytic derivatives, similar to pseudo-inverse jacobian methods
- Also 2nd order / quadratic
- IK goals: evolution
- Arbitrary fitness function
- Less problems at joint limits & with local minima
- No explicit matrix inversions

SIMD

SIMD

- SIMD (Single Instruction Multiple Data)
- load/store/add/mul/sub/... multiple values at the same time
- Vectors
- SIMD registers, SIMD instructions
- SSE, AVX, FMA
- Aligned memory access
- Programming:
 - SIMD intrinsics (eg. `_mm_add_pd(a, b)`)
 - auto-vectorization (eg. `#pragma omp simd`)
- Function multiversioning

SIMD Mutation

Mutation

- SIMD auto-vectorization

```
#pragma omp simd
    aligned(genes_min:32),
    aligned(genes_max:32),
    ...
for(size_t gene_index = 0; ... )
{
    ...
}
```

SIMD Forward kinematics

- Forward kinematics
- Quaternion operations: bad for SIMD - shuffle
- Can't be efficiently vectorized
- Instruction latencies
- **double r_x =**
$$(p_w * q_x + p_x * q_w) +$$
$$(p_y * q_z - p_z * q_y);$$
- **double r_y =**
$$(p_w * q_y - p_x * q_z) +$$
$$(p_y * q_w + p_z * q_x);$$
- **double r_z =**
$$(p_w * q_z + p_x * q_y) -$$

SIMD Forward kinematics

- Forward kinematics
- Extrapolation = ideal for SIMD
 - px += delta_pos.x() * delta_joint;
py += delta_pos.y() * delta_joint;
pz += delta_pos.z() * delta_joint;
 - __m256d p = _mm256_load_pd(
 tip_frame_ptr);
 [...]
p = _mm256_fmadd_pd(ff ,
 _mm256_load_pd(joint_delta_ptr) , p);
 [...]
_mm256_store_pd(tip_mutation_ptr , p);

Neural Networks

Neural Networks

- Library: FANN
("Fast Artificial Neural Network", C++)
- Input: goal
- Output: joint values
- ... inaccurate!

Neural Networks

- Learn relative movements instead!
- Input:
 - normalized relative goal pose (x,y,z, rx, ry, rz)
 - current joint values
- Output: joint offsets
- Multiple iterations
- Fully connected $[n, 50, m]$ network

PR2 Arm	Neural Network	5ms	10^{-5}	12%
PR2 Arm	Neural Network	100ms	10^{-5}	29%
PR2 Arm	KDL Network	5ms	10^{-5}	55%

Experiments

PR2 KDL

video

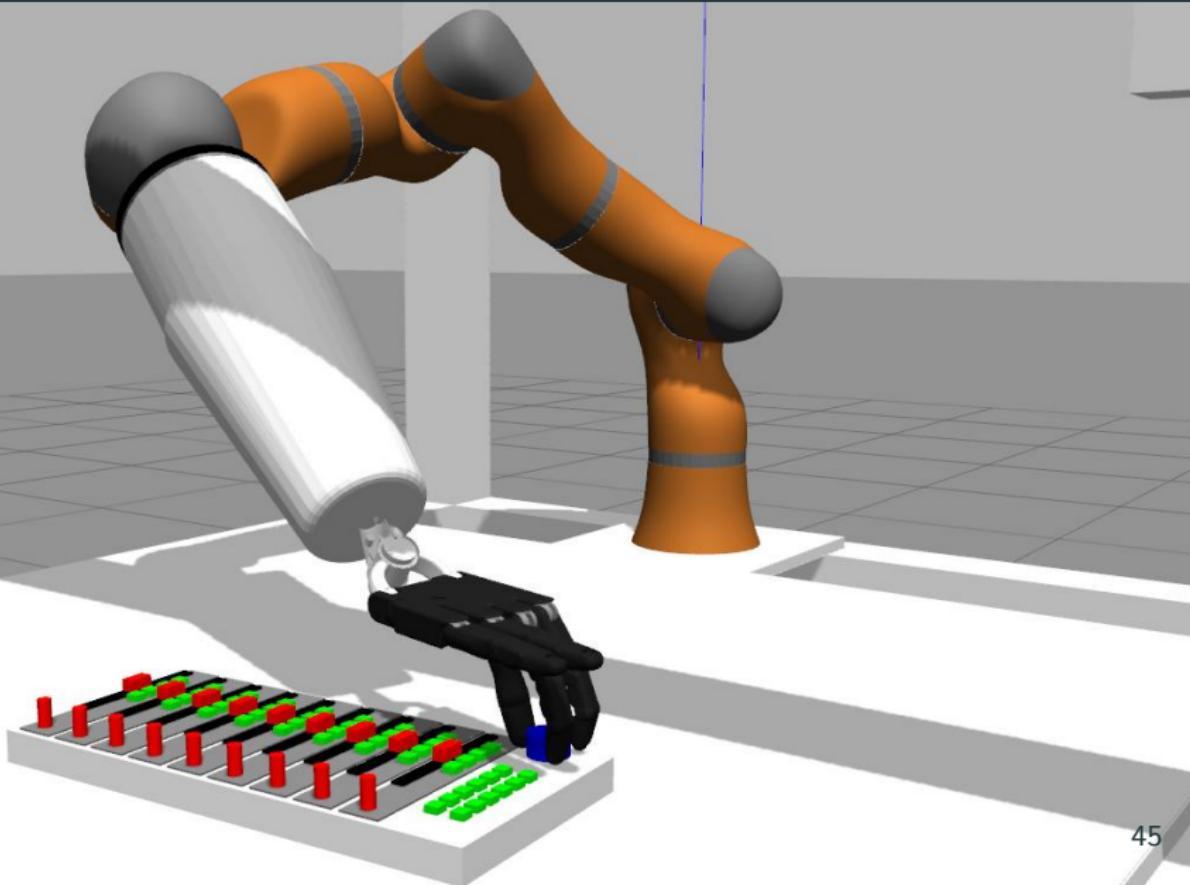
PR2 BiolK Minimal Displacement

video

PR2 BiolK Multi-Goal

video

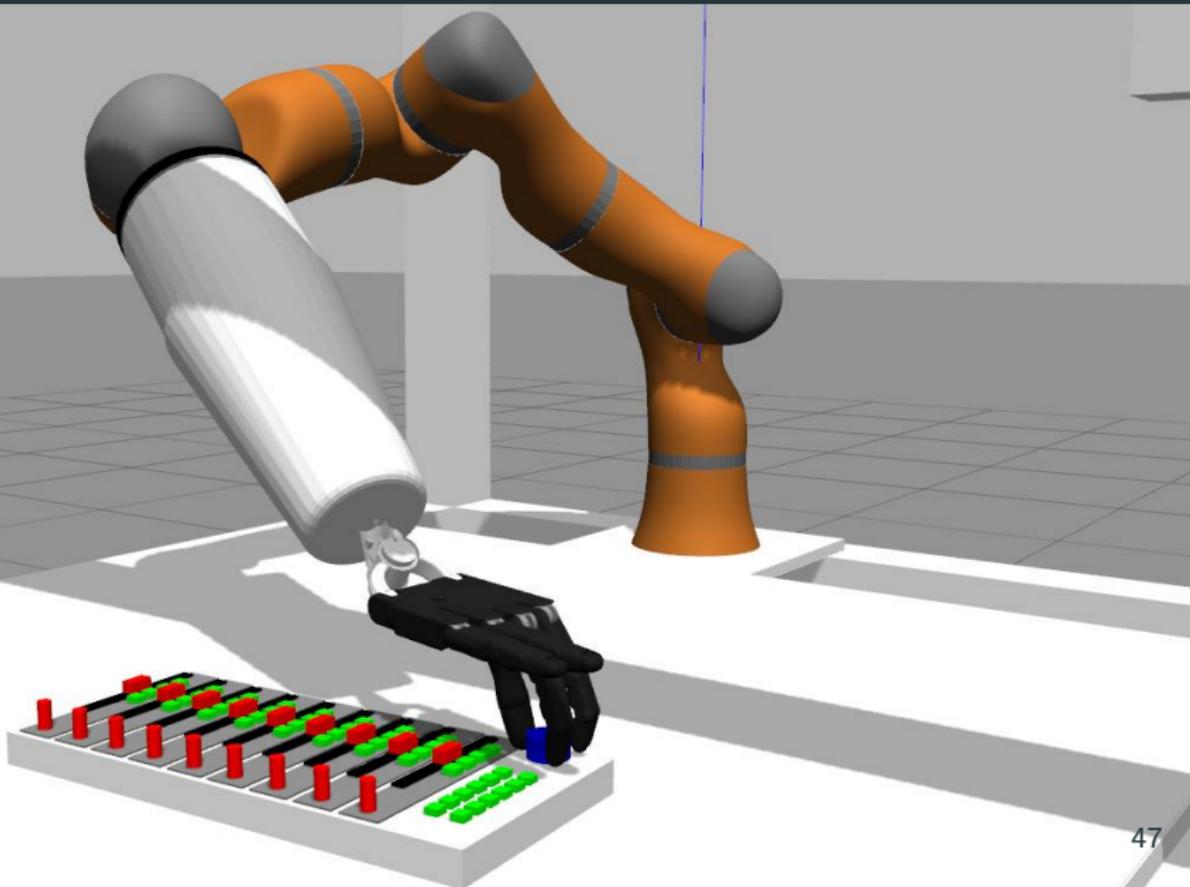
Shadow Hand



Shadow Hand BiolK

video

Shadow Hand Mixer Test



Shadow Hand Mixer Test

IK Goals

- Finger tips (Line, Touch, LookAt)
- Palm horizontal (LookAt)
- Coupled joint pair (J1, J2)
- MinimalDisplacement (secondary),
CenterJoints (secondary)

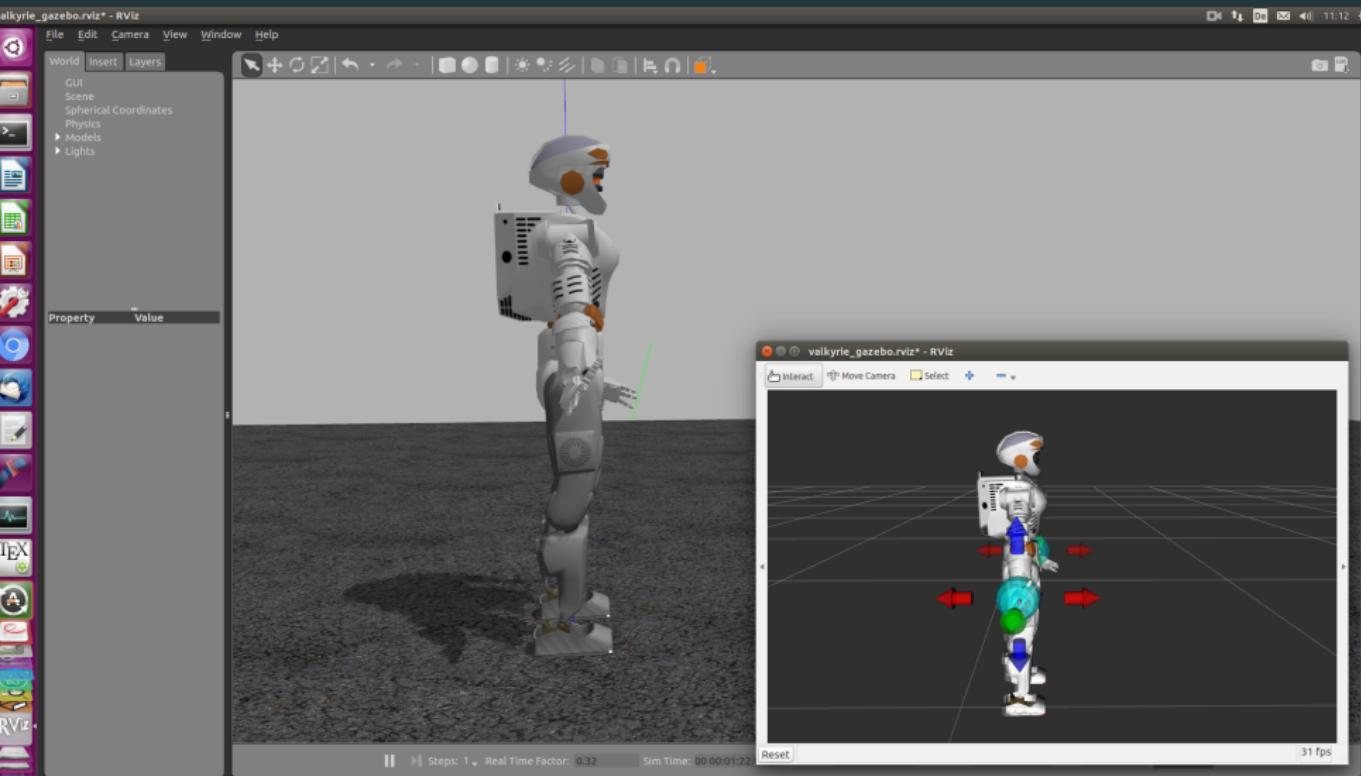
Shadow Hand Mixer Test

video

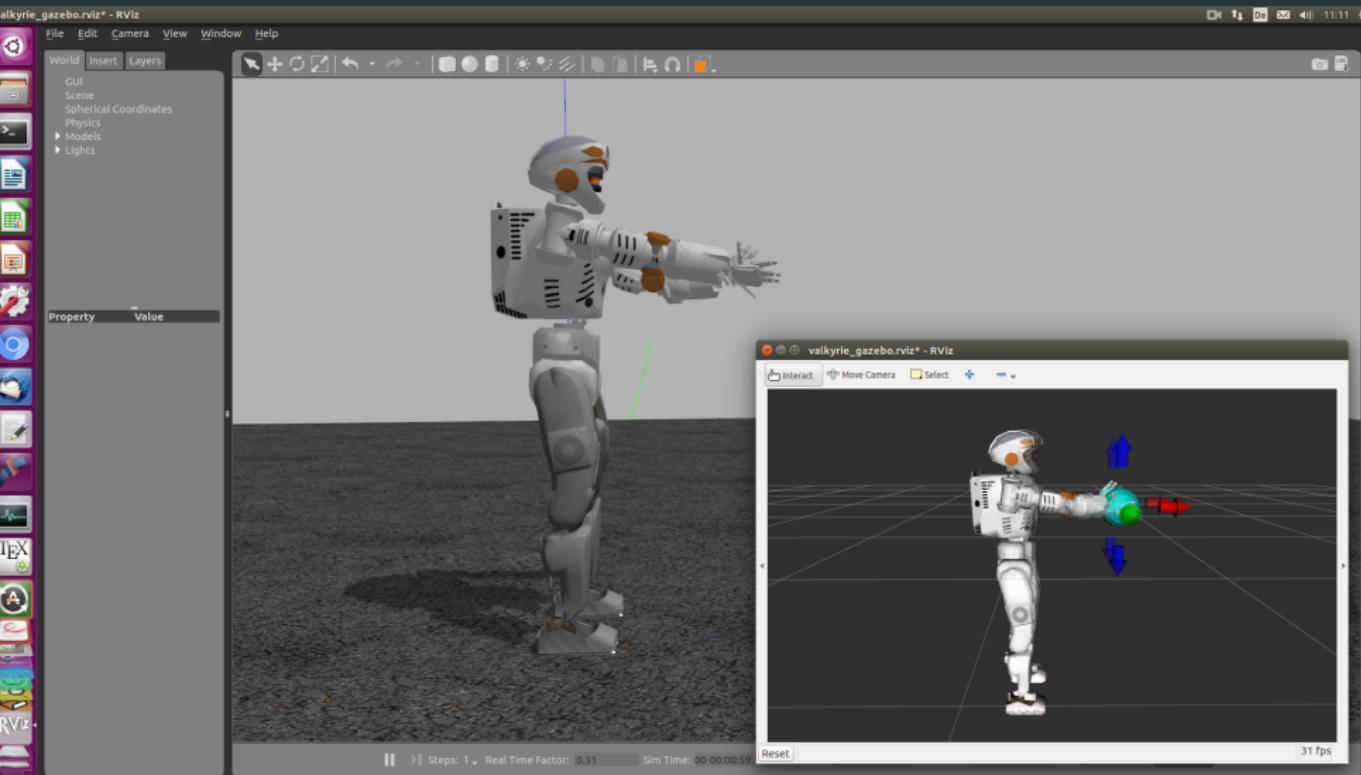
Valkyrie Balancing Test

- Hands (PositionGoal, interactive)
- Feet (PoseGoal)
- Foot reflexes
- Body upright (PositionGoal, OrientationGoal, OrientationGoal)
- BalanceGoal

Valkyrie Balancing Test



Valkyrie Balancing Test



Valkyrie Balancing Test

video

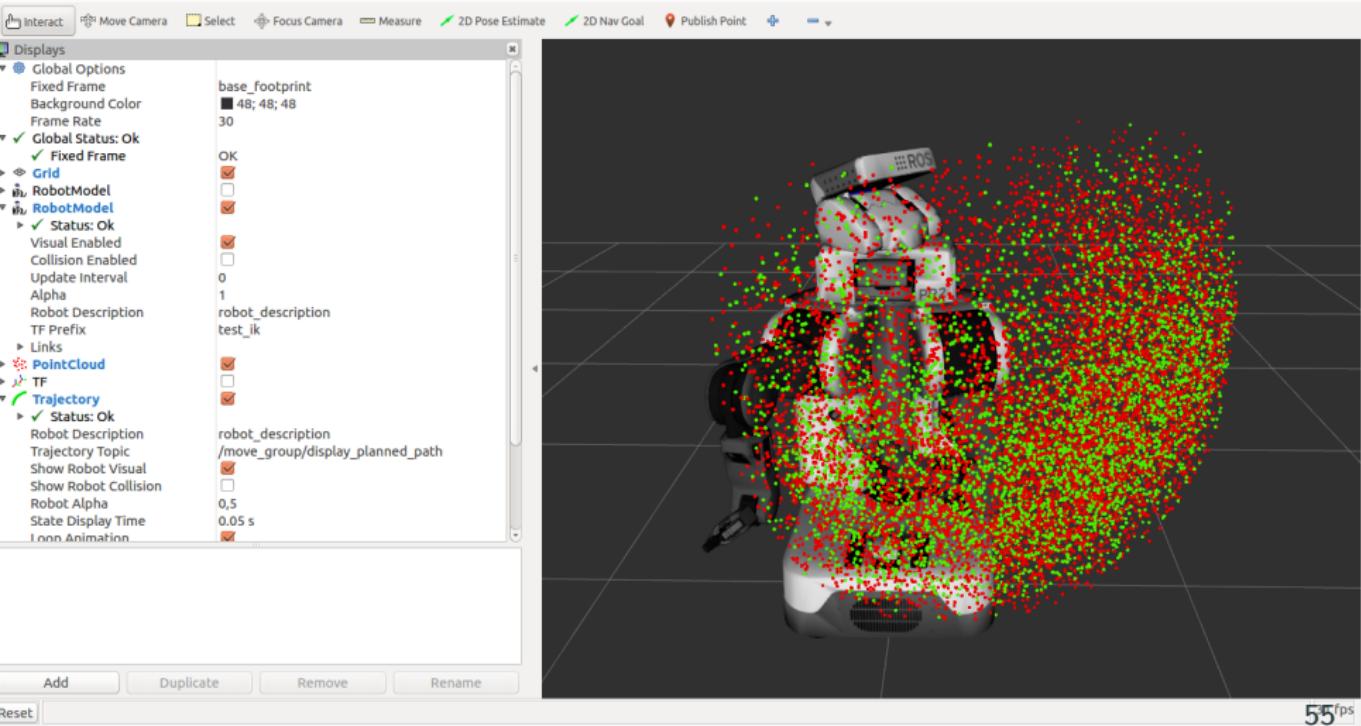
Benchmark

Benchmark

- Generate random joint values
- Compute forward kinematics
- Compute inverse kinematics
- Compare results
- Timeout (5ms)
- Max error (10^{-5})
- 10000 samples
- Success rate
- Average solve time

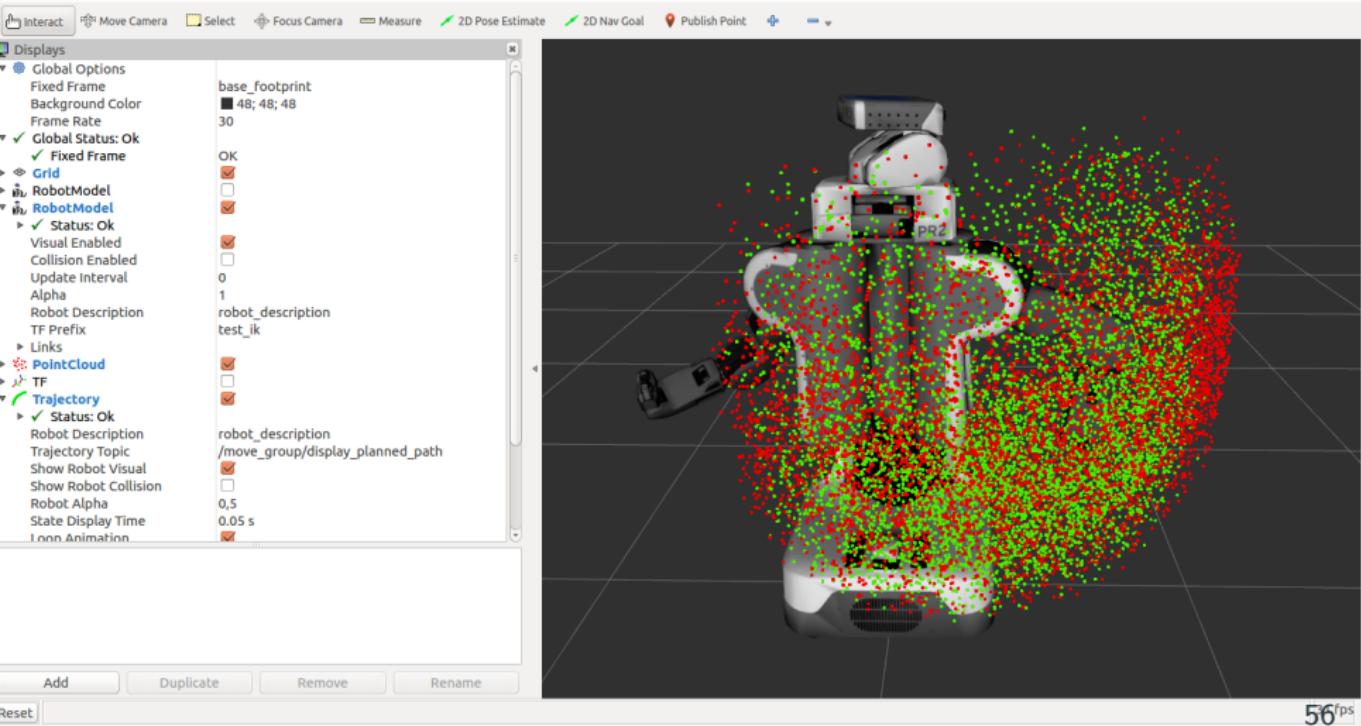
IK Test - PR2 - KDL

- 53.44% success rate



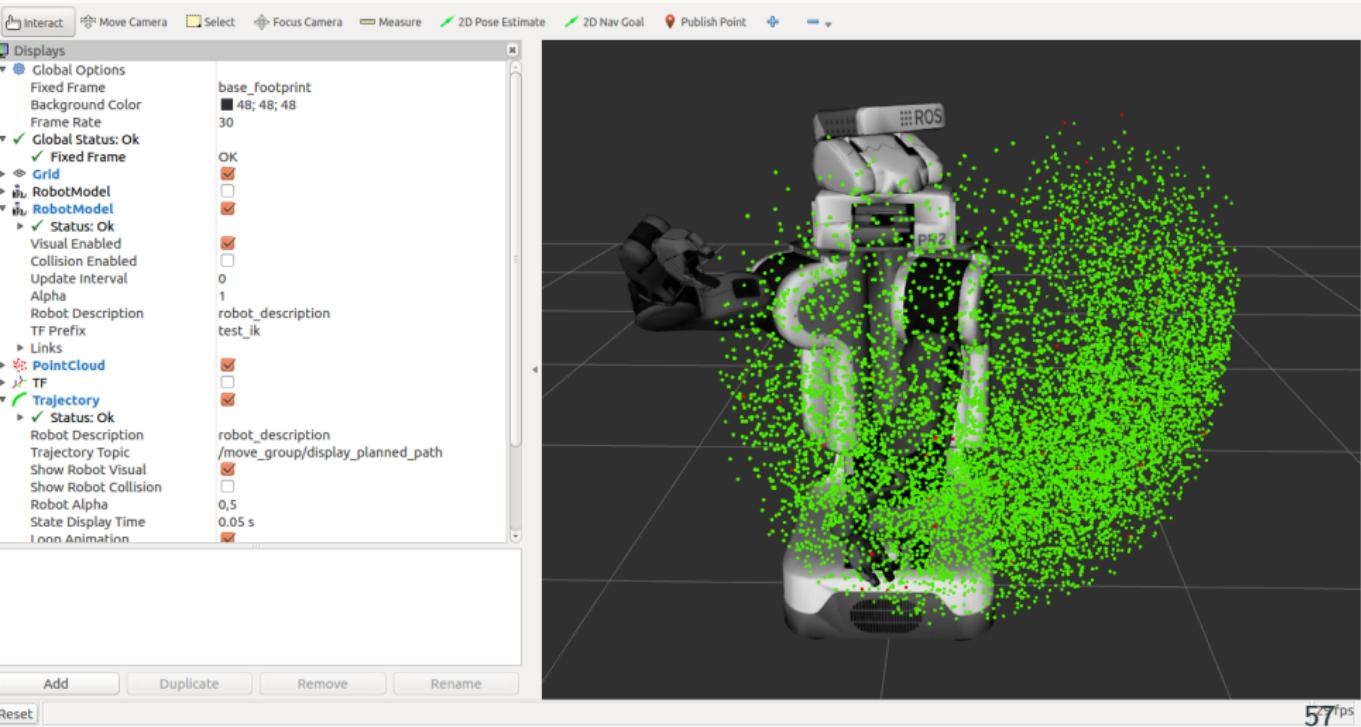
IK Test - PR2 - BiolK 1

- 70.34% success rate



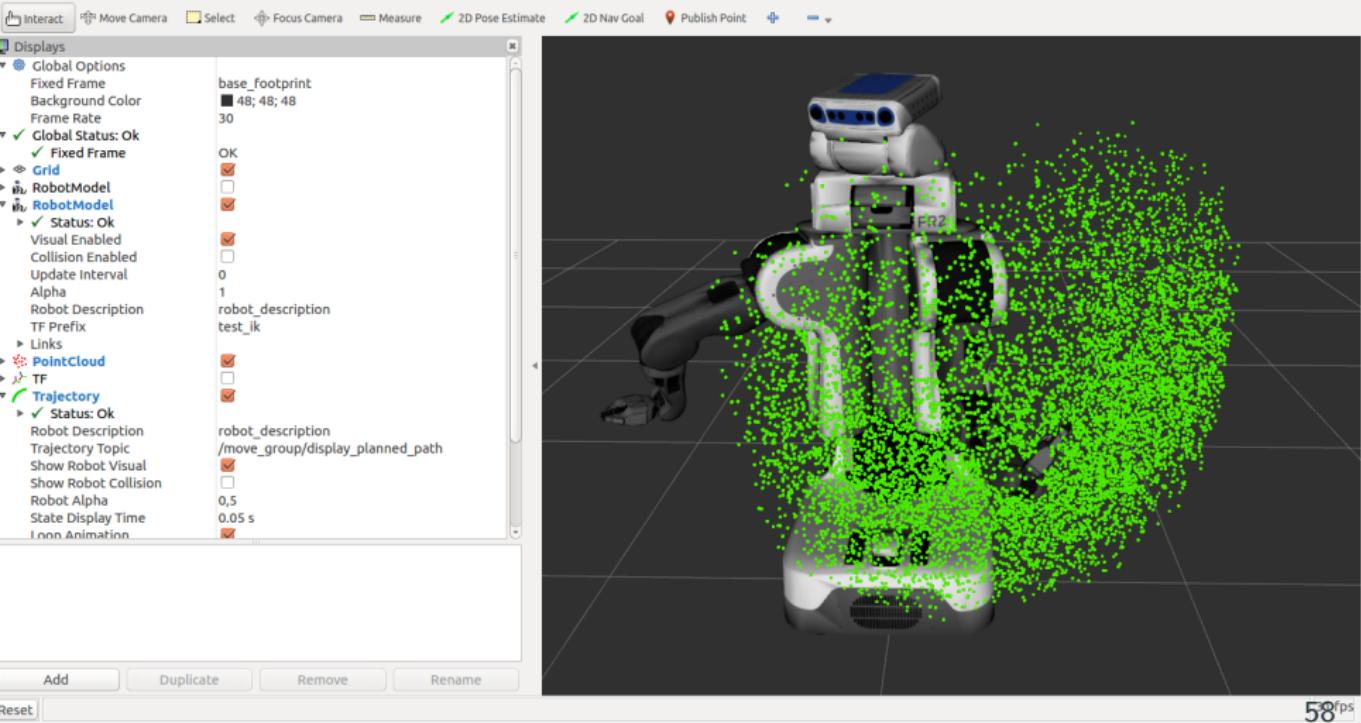
IK Test - PR2 - TRAC IK

- 99.69% success rate



IK Test - PR2 - BiolK 2

- 99.99% success rate



Benchmark

Success Rate

-	PR2	UR5	Valkyrie arm	Valkyrie foot	iiwa	avg	err
bio2_memetic	100.00%	99.90%	99.93%	100.00%	99.88%	99.94%	0.06%
bio2_memetic_I	100.00%	99.79%	99.97%	100.00%	99.83%	99.92%	0.08%
bio2	99.99%	99.78%	99.74%	99.99%	99.90%	99.88%	0.12%
trac_ik	99.79%	99.34%	99.55%	99.98%	99.87%	99.71%	0.29%
bio2_memetic_lbfsgs	99.98%	99.55%	98.68%	100.00%	99.56%	99.55%	0.45%
jac_4	83.31%	94.95%	71.76%	92.69%	78.17%	84.18%	15.82%
gd_c_4	82.56%	87.55%	63.41%	94.42%	84.35%	82.46%	17.54%
bio1	76.04%	50.79%	29.23%	70.00%	67.51%	58.71%	41.29%

BioIK 99.94%, TRAC_IK 99.71%

Benchmark

Average Solve Time

-	PR2	UR5	Valkyrie arm	Valkyrie foot	iiwa	avg
bio2_memetic	0.47ms	0.51ms	0.51ms	0.30ms	0.48ms	0.45ms
trac_ik	0.77ms	0.52ms	0.73ms	0.19ms	0.42ms	0.53ms
bio2_memetic_l	0.52ms	0.64ms	0.66ms	0.36ms	0.60ms	0.56ms
bio2	0.97ms	0.89ms	1.38ms	0.79ms	0.96ms	1.00ms
bio2_memetic_lbfsgs	1.28ms	1.31ms	1.88ms	0.75ms	1.35ms	1.31ms
jac_4	2.02ms	0.68ms	1.95ms	0.52ms	1.52ms	1.34ms
gd_c_4	2.36ms	1.84ms	2.94ms	0.99ms	1.82ms	1.99ms
bio1	3.93ms	4.13ms	4.66ms	3.14ms	3.50ms	3.87ms

BioIK: 0.45ms, TRAC_IK: 0.53ms

Benchmark

BioIK

- Higher success rate
(BioIK 99.94%, TRAC_IK 99.71%)
- Lower average solve time
(BioIK: 0.45ms, TRAC_IK: 0.53ms)

