

# Human Body Pose Estimation Missing Joints Reconstruction through Interpolations Comparison

In this work, we delve into the various n dimensional interpolation used to estimate the missing joints from the pose estimation of OpenPose.

1. Dataset Creation: to accomplish the goal of assign the best interpolation method for reconstructing missing frames of human body joints, a dataset has to be created.

In order to obtain a ground truth dataset, all the sequences that have at least 5 sequential fully known values frames, have been retrieved from the original dataset, obtaining a dataset with 10,028 rows and 74 columns, for a total of 742,072 values.

From this dataset, random sequences of missing points have been placed, dividing each columns in sequences based on sequential frame and video name.

| NoseX    | NoseY    | NoseC    | LEyeX    | LEyeY    | LEyeC    | REyeX    | REyeY    | REyeC    | LEarX    | LEarY    | LEarC    | REarX    | REarY    | REarC    | LShoulderX | LShoulderY | LShoulderC |
|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|------------|------------|------------|
| 0.401606 | 0.5202   | 0.831084 | 0.401527 | 0.510641 | 0.730255 | 0.398759 | 0.520298 | 0.697503 | 0.404232 | 0.491339 | 0.593905 | 0.398907 | 0.530033 | 0.409009 | 0.423458   | 0.486276   | 0.81084    |
| 0.401591 | 0.520243 | 0.849303 | 0.401529 | 0.510664 | 0.750677 | 0.396182 | 0.520339 | 0.719071 | 0.404252 | 0.491309 | 0.60373  | 0.39889  | 0.530109 | 0.392076 | 0.423454   | 0.486305   | 0.806153   |
| 0        | 0        | 0        | 0        | 0        | 0        | 0        | 0        | 0        | 0        | 0        | 0        | 0        | 0        | 0        | 0.423469   | 0.486342   | 0.801978   |
| 0        | 0        | 0        | 0        | 0        | 0        | 0        | 0        | 0        | 0        | 0        | 0        | 0        | 0        | 0        | 0.423468   | 0.486339   | 0.80523    |
| 0        | 0        | 0        | 0        | 0        | 0        | 0.398763 | 0.520419 | 0.66846  | 0.404274 | 0.491372 | 0.602178 | 0        | 0        | 0        | 0          | 0          | 0          |
| 0.401573 | 0.520413 | 0.828501 | 0.401545 | 0.51545  | 0.764389 | 0.398763 | 0.520447 | 0.648469 | 0.40432  | 0.495998 | 0.594207 | 0.401423 | 0.530097 | 0.292935 | 0.423516   | 0.486391   | 0.800324   |
| 0.398829 | 0.520434 | 0.814817 | 0.398816 | 0.515389 | 0.77608  | 0.39074  | 0.520568 | 0.627781 | 0.401661 | 0.49136  | 0.656558 | 0.398746 | 0.529962 | 0.195311 | 0.423477   | 0.486178   | 0.82624    |
| 0.398836 | 0.520334 | 0.826516 | 0.398814 | 0.510719 | 0.770468 | 0.393355 | 0.520476 | 0.633394 | 0.40168  | 0.491242 | 0.672432 | 0.398774 | 0.525473 | 0.20905  | 0.423513   | 0.481695   | 0.832449   |
| 0.398888 | 0.520368 | 0.839831 | 0.398854 | 0.510719 | 0.780024 | 0.393398 | 0.52051  | 0.62711  | 0        | 0        | 0        | 0.39877  | 0.525413 | 0.192991 | 0          | 0          | 0          |
| 0.401448 | 0.520275 | 0.830452 | 0.398882 | 0.510619 | 0.763428 | 0        | 0        | 0        | 0        | 0        | 0        | 0.398829 | 0.525395 | 0.236703 | 0.426074   | 0.481602   | 0.838588   |
| 0.401444 | 0.520163 | 0.817774 | 0.398859 | 0.510519 | 0.745219 | 0        | 0        | 0        | 0        | 0        | 0        | 0.398853 | 0.525364 | 0.292568 | 0.423515   | 0.481555   | 0.849206   |
| 0.398904 | 0.520192 | 0.820488 | 0.398843 | 0.510544 | 0.748448 | 0.393445 | 0.520335 | 0.624543 | 0.401641 | 0.491217 | 0.67187  | 0.398814 | 0.525403 | 0.27113  | 0.423503   | 0.481573   | 0.847431   |
| 0.401536 | 0.515476 | 0.804385 | 0.401516 | 0.50569  | 0.742377 | 0.396181 | 0.520108 | 0.694673 | 0.404196 | 0.491298 | 0.550921 | 0.398838 | 0.525258 | 0.441969 | 0.423471   | 0.481638   | 0.835873   |
| 0        | 0        | 0        | 0.401498 | 0.505694 | 0.741592 | 0.396169 | 0.520101 | 0.687774 | 0.401701 | 0.491307 | 0.55983  | 0.398841 | 0.525274 | 0.418002 | 0.423467   | 0.481631   | 0.838082   |
| 0.401515 | 0.520286 | 0.833196 | 0.401498 | 0.510667 | 0.780196 | 0.396168 | 0.520364 | 0.660748 | 0        | 0        | 0        | 0.398802 | 0.525357 | 0.263767 | 0          | 0          | 0          |
| 0        | 0        | 0        | 0.401503 | 0.515405 | 0.819898 | 0.396133 | 0.520547 | 0.665273 | 0        | 0        | 0        | 0.398775 | 0.525409 | 0.189452 | 0          | 0          | 0          |
| 0.401494 | 0.520517 | 0.897993 | 0.401501 | 0.515426 | 0.867931 | 0.396091 | 0.524968 | 0.713403 | 0.404291 | 0.491295 | 0.713471 | 0.396177 | 0.525453 | 0.160551 | 0.426085   | 0.481639   | 0.825437   |
| 0.401664 | 0.520432 | 0.838309 | 0.401623 | 0.51535  | 0.775447 | 0.398868 | 0.52051  | 0.523883 | 0.404421 | 0.491332 | 0.648427 | 0.401512 | 0.525299 | 0.170042 | 0.42889    | 0.48621    | 0.83261    |
| 0.401712 | 0.520461 | 0.832494 | 0        | 0        | 0        | 0        | 0        | 0        | 0.404432 | 0.49132  | 0.658833 | 0.401546 | 0.525253 | 0.189156 | 0.428892   | 0.48616    | 0.834981   |
| 0.404215 | 0.525056 | 0.868245 | 0.401715 | 0.515504 | 0.828685 | 0        | 0        | 0        | 0.407023 | 0.491308 | 0.692642 | 0        | 0        | 0        | 0          | 0          | 0          |
| 0        | 0        | 0        | 0        | 0        | 0        | 0.398915 | 0.525032 | 0.529516 | 0.404486 | 0.491258 | 0.698668 | 0        | 0        | 0        | 0.428948   | 0.481716   | 0.834024   |
| 0        | 0        | 0        | 0.404208 | 0.515374 | 0.813818 | 0.401497 | 0.524967 | 0.574043 | 0.407056 | 0.491252 | 0.68526  | 0        | 0        | 0        | 0.428958   | 0.48616    | 0.822059   |
| 0.404264 | 0.520539 | 0.833052 | 0.404221 | 0.515429 | 0.79056  | 0.401513 | 0.520583 | 0.53415  | 0.40704  | 0.491319 | 0.670467 | 0.401652 | 0.525271 | 0.159643 | 0.428973   | 0.486174   | 0.818843   |
| 0.600854 | 0.360113 | 0.837288 | 0.614338 | 0.364729 | 0.760266 | 0.603472 | 0.340848 | 0.760931 | 0.60624  | 0.374622 | 0.467758 | 0.597877 | 0.335918 | 0.559898 | 0.579085   | 0.428127   | 0.794178   |
| 0.600736 | 0.360054 | 0.867444 | 0.608996 | 0.364745 | 0.797349 | 0.600834 | 0.345523 | 0.769306 | 0        | 0        | 0        | 0        | 0        | 0        | 0.581759   | 0.432984   | 0.79035    |
| 0.600738 | 0.364925 | 0.894064 | 0.608996 | 0.365143 | 0.805143 | 0.600865 | 0.355193 | 0.789909 | 0.603583 | 0.3844   | 0.510346 | 0        | 0        | 0        | 0          | 0          | 0          |
| 0.600715 | 0.364934 | 0.884783 | 0.611641 | 0.365193 | 0.78202  | 0.600856 | 0.35519  | 0.769349 | 0        | 0        | 0        | 0        | 0        | 0        | 0.578968   | 0.437838   | 0.822133   |
| 0.600677 | 0.365057 | 0.849987 | 0.608924 | 0.369715 | 0.736491 | 0.600803 | 0.355323 | 0.748469 | 0        | 0        | 0        | 0        | 0        | 0        | 0          | 0          | 0          |
| 0        | 0        | 0        | 0.603585 | 0.370061 | 0.78368  | 0        | 0        | 0        | 0        | 0        | 0        | 0        | 0        | 0        | 0.57885    | 0.442548   | 0.81938    |
| 0        | 0        | 0        | 0.606228 | 0.369711 | 0.728597 | 0.600778 | 0.355227 | 0.656174 | 0.600868 | 0.398841 | 0.625283 | 0        | 0        | 0        | 0.57351    | 0.437742   | 0.764452   |
| 0        | 0        | 0        | 0.606193 | 0.369931 | 0.674828 | 0        | 0        | 0        | 0.600673 | 0.403626 | 0.574186 | 0        | 0        | 0        | 0.567991   | 0.432902   | 0.683871   |
| 0        | 0        | 0        | 0.603534 | 0.374795 | 0.487962 | 0        | 0        | 0        | 0        | 0        | 0.597933 | 0.360199 | 0.33768  | 0.568039 | 0.41844    | 0.491893   | 0          |
| 0        | 0        | 0        | 0.603449 | 0.370031 | 0.37056  | 0.6007   | 0.364988 | 0.386532 | 0.598111 | 0.398954 | 0.28167  | 0        | 0        | 0        | 0          | 0          | 0          |
| 0.598084 | 0.379568 | 0.249924 | 0.600832 | 0.374771 | 0.180003 | 0.600702 | 0.369955 | 0.23093  | 0.597976 | 0.379604 | 0.139244 | 0.597972 | 0.369794 | 0.21779  | 0.55983    | 0.394056   | 0.280756   |
| 0.407116 | 0.442465 | 0.718413 | 0.407047 | 0.428097 | 0.677203 | 0.404453 | 0.432924 | 0.30462  | 0.417896 | 0.408622 | 0.859012 | 0.404252 | 0.423293 | 0.118526 | 0.445325   | 0.398982   | 0.758768   |

Small portion of how the missin values dataset looks like.

2. Analysis on this dataset: Some analyses that can be made on it, regards the total number of zeros generated.

3. Interpolation comparison:

- Linear
- Inverse Distance Weight
- Nearest Neighbor
- CubicSpline
- Akima
- Pchip
- Whittaker Shannon

All these interpolations are then applied on the zero filled dataset.

4. Evaluation Metrics:

The metrics involved to compare the interpolations are the following:

- RMSE
- Pearson Coefficient
- Cosine Similarity
- Euclidean Distance

The metrics are applied only on the reconstructed sequences, the known values are not taken into account.

The results are then listed:

Akima :

Euclidean Distance: 15.286

Cosine Similarity: 0.99917

Pearson Correlation Coefficient: 0.9929780441232688

Root Mean Squared Error: 0.025471

IDW :

Euclidean Distance: 13.99

Cosine Similarity: 0.9993

Pearson Correlation Coefficient: 0.9940978100464436

Root Mean Squared Error: 0.023312

Linear :

Euclidean Distance: 12.613

Cosine Similarity: 0.99943

Pearson Correlation Coefficient: 0.9952066445134503

Root Mean Squared Error: 0.021018

Spline :

Euclidean Distance: 16.715

Cosine Similarity: 0.999

Pearson Correlation Coefficient: 0.9916231349463192

Root Mean Squared Error: 0.027852

Pchip :

Euclidean Distance: 12.967

Cosine Similarity: 0.9994

Pearson Correlation Coefficient: 0.9949367262151254

Root Mean Squared Error: 0.021606

Nearest :

Euclidean Distance: 15.286

Cosine Similarity: 0.99917

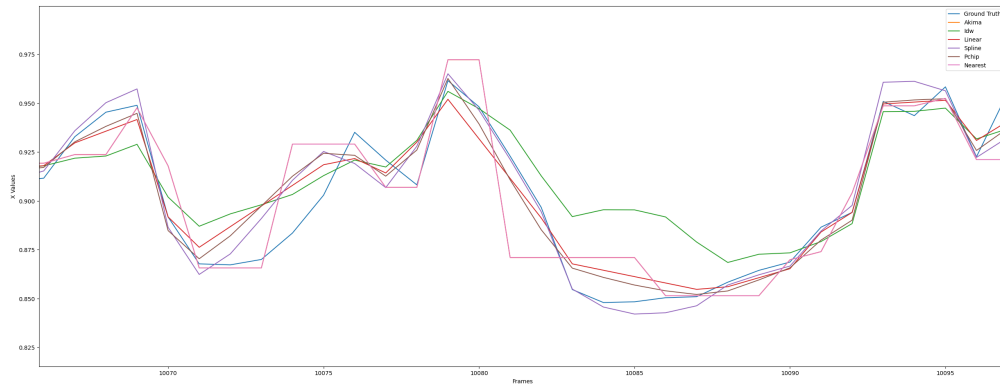
Pearson Correlation Coefficient: 0.9929780441232688

Root Mean Squared Error: 0.025471

From the results, the Linear Interpolation seems to interpolate values with a higher accuracy than others.

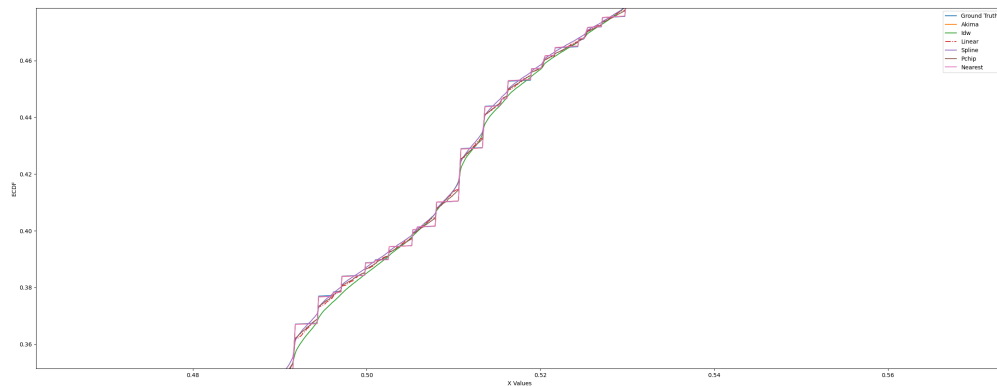
## 5. Visual Inspections

The results are then illustrated:



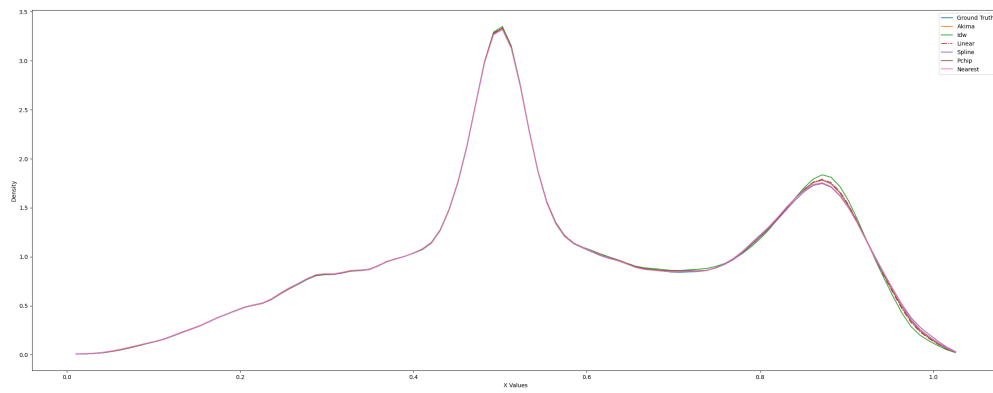
Small portion of the interpolation comparison.

Considering the high quantity of unique continuous values, a ECDF representation has been employed:



Small portion of the whole ECDF function.

Finally, a density estimation plot has been created, to observe the sequences behaviour with less noise.



And a closer looks is then illustrated:

