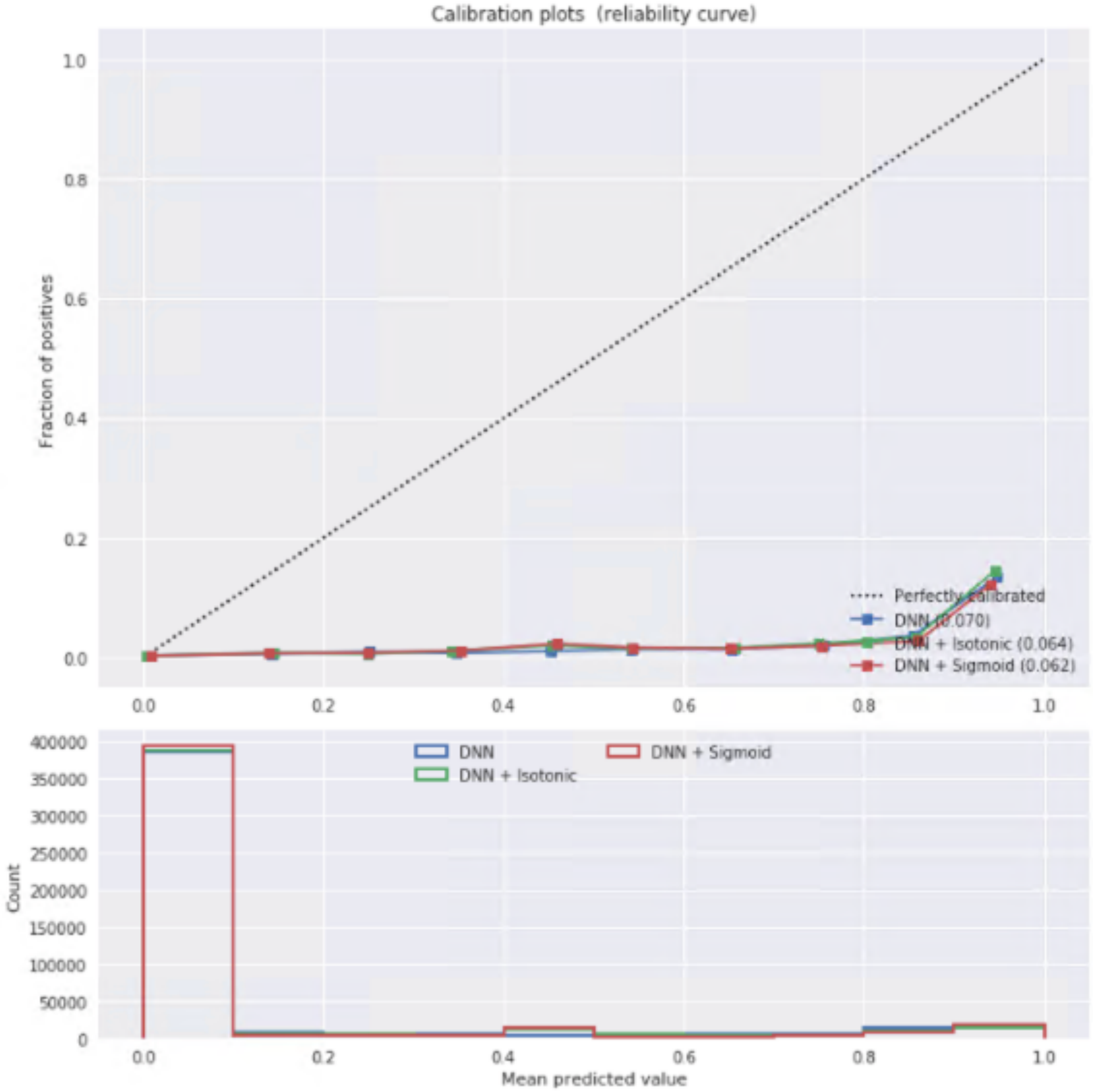
Supplemental Figure 1. Calibration curve for optimal model with results from re-scaling methods



Legend: Calibration plot for the DNN with alternative curves demonstrating lack of improvement in calibration with use of rescaling methods (both Platt’s rescaling and isotonic regression).

Python Code:

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    "from keras.models import Sequential\n",
    "from keras.layers import Dense, Activation, Dropout\n",
    "from keras.layers.advanced_activations import LeakyReLU, PReLU\n",
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    "    This function prints and plots the confusion matrix.\n",
    "    Normalization can be applied by setting `normalize=True`.\n",
    "    \"\"\"\n",
    "    plt.imshow(cm, interpolation='nearest', cmap=cmap)\n",
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    "    tick_marks = np.arange(len(classes))\n",
    "    plt.xticks(tick_marks, classes, rotation=0)\n",
    "    plt.yticks(tick_marks, classes)\n",
    "\n",
    "    if normalize:\n",
    "        cm = cm.astype('float') / cm.sum(axis=1)[:, np.newaxis]\n",
    "        #print(\"Normalized confusion matrix\")\n",
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    "\n",
   "    #print(cm)\n",
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   "                 color=\"white\" if cm[i, j] > thresh else \"black\")\n",
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