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# Tables

|  |  |  |
| --- | --- | --- |
| ***α*** | **GMM (8 mix.) Static+∆** | **MixAR (4-mix.)****Static** |
| 0.00 | 1.50 (288) | 1.50 (240) |
| 0.25 | 3.25 (576) | 3.50 (240) |
| 0.50 | 10.25 (576) | 6.25 (240) |
| 0.75 | 24.75 (576) | 9.75 (240) |
| 1.00 | 26.75 (576) | 13.75 (240) |

Table . A comparison of classification error rates is shown for a GMM system using static+delta features and a MixAR system operating only on the static features is shown on synthetic data. The number of parameters for each system is shown in parentheses. The GMM system, which uses static and delta features, performs significantly worse than the MixAR system as the nonlinearity in the data increases.

|  |  |  |
| --- | --- | --- |
| **No.Mixtures** | **GMM****Static+∆+∆∆** | **MixAR****Static Only** |
| 2 | 23.1 (216) | 24.1(120) |
| 4 | 21.7 (432) | 19.2(240) |
| 8 | 20.5 (864) | 19.1(480) |
| 16 | 20.5 (1728) | 19.2(960) |

Table . EERs are shown as a function of the number of mixtures. MixAR performs slightly better with almost half the number of parameters.

|  |  |  |
| --- | --- | --- |
| **Features** | **GMM16-mix.** | **MixAR****8-mix.** |
| Static(12) | 22.1 | 19.1 |
| Static+E(13) | 33.1 | 41.1 |
| Static+Δ(24) | 20.6 | 20.4 |
| Static+Δ+ΔΔ(36) | 20.5 | 20.5 |

Table . Speaker verification EERs are shown for MixAR and GMM for a variety of feature vector combinations. MixAR does not need delta features since the model itself encodes temporal dynamics.

|  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- |
| **GMM(1168)** | **SNR (dB)** | **Clean** | **Car Noise** | **WhiteNoise** | **Babble Noise** |
|  | **2.4** |  |
| 10 dB |  | 19.7 | 48.7 | 40.6 |
| 5 dB |  | 31.2 | 50.0 | 44.7 |
| 0 dB |  | 39.3 | 49.8 | 48.2 |
| **MixAR (480)** |  | **1.8** |  |
| 10 dB |  | 13.7 | 47.0 | 36.9 |
| 5 dB |  | 23.2 | 47.6 | 42.8 |
| 0dB |  | 33.9 | 48.5 | 47.6 |

Table . EERs are shown for a variety of noise conditions.

|  |  |  |
| --- | --- | --- |
|  | **Training Utterance Duration** | **EER** |
| **GMM****(864)** | 120 | 20.5 |
| 90 | 20.4 |
| 60 | 20.4 |
| 30 | 24.4 |
| 15 | 29.5 |
| **MixAR****(480)** | 120 | 19.2 |
| 90 | 21.5 |
| 60 | 21.8 |
| 30 | 21.8 |
| 15 | 24.3 |

Table . Performance is analyzed as a function of the duration of the training data utterances. The evaluation utterance durations were held constant and varied between 20 and 40 seconds.

|  |  |  |
| --- | --- | --- |
| **Database** | **GMM Static+∆+∆∆(1728)** | **MixAR Static Only(480)** |
| TIMIT | 2.4 | 1.8 |
| NTIMIT | 21.0 | 20.9 |

Table . EERs are shown for TIMIT (clean data) and NTIMIT (noisy data).

|  |  |  |
| --- | --- | --- |
|  | **EvaluationUtterance Duration** | **EER** |
| **GMM****(864)** | 30 | 20.5 |
| 15 | 21.8 |
| 10 | 21.5 |
| 5 | 24.4 |
| 3 | 26.9 |
| **MixAR****(480)** | 30 | 19.2 |
| 15 | 23.4 |
| 10 | 23.1 |
| 5 | 25.6 |
| 3 | 25.6 |

Table . Performance is analyzed as a function of the duration of the evaluation data is shown. The training utterance duration was fixed and averaged around 120s.