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| 22 | Abstract | <p>Stroke is the fourth leading cause of death and the leading cause of long-term disability in USA. Brain imaging data from experimental stroke models and stroke patients have shown that there is often a gradual progression of potentially reversible ischemic injury toward infarction. Reestablishing tissue perfusion and/or treating with neuroprotective drugs in a timely fashion are expected to salvage some ischemic tissues. Diffusion-weighted imaging based on magnetic resonance imaging (MRI) in which contrast is based on water motion can detect ischemic injury within minutes after onsets, whereas computed tomography and other imaging modalities fail to detect stroke injury for at least a few hours. Along with quantitative perfusion imaging, the perfusion–diffusion mismatch which approximates the ischemic penumbra could be imaged noninvasively. This review describes recent progresses in the development and application of</p> | |

multimodal MRI and image analysis techniques to study ischemic tissue at risk in experimental stroke in rats.

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- 23 Keywords separated by ' - ' MRI - Perfusion–diffusion mismatch - ADC - CBF - DWI - PWI - Experimental stroke model - Rodents - Oxygen challenge - Predictive mode - Magnetic resonance imaging - Rats - Hyperperfusion - fMRI
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REVIEW ARTICLE

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Multimodal MRI of Experimental Stroke

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Abstract Stroke is the fourth leading cause of death and the leading cause of long-term disability in USA. Brain imaging data from experimental stroke models and stroke patients have shown that there is often a gradual progression of potentially reversible ischemic injury toward infarction. Reestablishing tissue perfusion and/or treating with neuroprotective drugs in a timely fashion are expected to salvage some ischemic tissues. Diffusion-weighted imaging based on magnetic resonance imaging (MRI) in which contrast is based on water motion can detect ischemic injury within minutes after onsets, whereas computed tomography and other imaging modalities fail to detect stroke injury for at least a few hours. Along with quantitative perfusion imaging, the perfusion–diffusion mismatch which approximates the ischemic penumbra could be imaged noninvasively. This review describes recent progresses in the development and application of multimodal MRI and image analysis techniques to study ischemic tissue at risk in experimental stroke in rats.

Keywords MRI · Perfusion–diffusion mismatch · ADC · CBF · DWI · PWI · Experimental stroke model · Rodents · Oxygen challenge · Predictive mode · Magnetic resonance imaging · Rats · Hyperperfusion · fMRI

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Introduction 32

Stroke—the fourth leading cause of death and the leading cause of long-term disability [1]—is a medical emergency caused by a disturbance in the blood supply to the brain, resulting in loss of brain functions. There are about 800,000 new or recurrent strokes each year. More than six million Americans have permanent neurological deficits from stroke, and 71% of these stroke survivors cannot return to work. Over \$70 billion is projected to be expended on stroke patient care in 2011 [1]. The cost of stroke care is steadily rising because the conditions that put people at risk for stroke (such as heart disease, diabetes, and obesity) are also steadily on the rise. Despite the tremendous effort invested in stroke research, our ability to identify salvageable tissue and to minimize neurological deficit in stroke patients remains extremely limited. Recanalization by recombinant tissue plasminogen activator (rtPA) therapy is the only proven method that could salvage some brain tissue. rtPA treatment unfortunately is limited to only a small subset of patients because it has serious risk of often fatal hemorrhagic transformation and can only be administered within 4.5 h of stroke onset [2]. As such, the ability to reliably distinguish salvageable versus non-salvageable tissue remains a high priority for clinical decision making in the treatment of acute stroke.

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In humans, the “perfusion–diffusion mismatch” obtained using magnetic resonance imaging (MRI) is presumed to approximate the “ischemic penumbra” and is increasingly

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60 used in clinical decision making in the management of acute
 61 stroke. Although the strict definition of ischemic penumbra
 62 requires correlation with energy metabolism [3–5] and such
 63 a correlation is not feasible in humans, the ischemic penum-
 64 bra and viability thresholds have been operationally defined
 65 based on diffusion-weighted imaging (DWI), perfusion-
 66 weighted imaging, and equivalent modalities. Although
 67 “perfusion–diffusion” mismatch is widely observed in acute
 68 human stroke [6–10], the tissue fate characterized by the
 69 perfusion–diffusion mismatch remains poorly understood
 70 and controversial [11]. Consequently, clinical decision mak-
 71 ing based on perfusion and diffusion imaging has not yet
 72 reached its fullest potential. Animal models in which the
 73 perfusion–diffusion mismatch can be reproducibly studied
 74 under controlled conditions are important to fully character-
 75 ize the tissue fate of ischemic injury (salvageable versus
 76 non-salvageable tissues) and to evaluate the efficacy of
 77 therapeutic intervention.

78 This paper reviews recent progresses, mostly from our
 79 group, in the development and application of multimodal
 80 MRI techniques and image analysis techniques to study isch-
 81 emic tissue at risk in experimental stroke in rats. First, we review
 82 the use of perfusion or diffusion data to characterize acute
 83 ischemic stroke and the use of automated clustering of com-
 84 bined perfusion and diffusion data to improve ischemic tissue
 85 characterization. Second, the compromise between high-
 86 temporal and high spatial resolution for acute stroke imaging
 87 is described. Third, the blood oxygen level-dependent (BOLD)
 88 fMRI of evoked responses to probe the perfusion–diffusion
 89 mismatch is described. Fourth, we describe quantitative predic-
 90 tive models to predict ischemic tissue fate based on acute
 91 perfusion and/or diffusion data. Fifth, we describe a multimodal
 92 MRI approach to investigate the hyperperfusion phenomenon
 93 associated with ischemic stroke. Finally, we review the use of
 94 BOLD fMRI of oxygen challenge to probe tissue fate and
 95 compare it to the perfusion–diffusion mismatch.

96 **Perfusion and Diffusion MRI**

97 MRI provides flexible and clinically relevant information to
 98 image stroke. In particular, DWI [12] in which contrast is
 99 based on water apparent diffusion coefficient (ADC) is
 100 widely recognized as a useful imaging modality because of
 101 its ability to detect stroke within minutes after onsets,
 102 whereas computed tomography and other imaging modal-
 103 ities fail to detect stroke injury for at least a few hours.
 104 Hyperintense regions on DWI correspond to tissues with a
 105 reduced ADC of water. Although the biophysical mecha-
 106 nism(s) underlying ADC reduction remains poorly under-
 107 stood and controversial [12, 13], the ADC decline has been
 108 correlated with energy failure and breakdown of membrane
 109 potential in animal models [3–5].

Cerebral blood flow (CBF) can be measured by using an
 exogenous intravascular contrast agent or by magnetically
 labeling the endogenous water in blood [14, 15]. The most
 widely used perfusion MRI technique is based on dynamic
 susceptibility contrast imaging (see review [15]) in which an
 intravenous bolus of a blood-pool MR contrast reagent such as
 gadolinium is injected while T₂* or T₂ imaging is performed.
 This technique is generally performed only once due to recir-
 culation of the contrast reagent and potential negative side
 effects. An alternative technique is based on the arterial spin
 labeling (ASL) technique that involves noninvasive magnetic
 labeling of blood water protons as they flow into the imaging
 slices, without the need for exogenous contrast reagents [15].
 ASL is becoming increasingly popular for measuring CBF.
 The magnetically labeled water has a short half-life (~blood
 T₁) and thus repeated ASL measurements can be made for
 signal averaging at relatively high spatial and temporal reso-
 lution. Continuous arterial spin labeling technique with the
 two-coil setup offers generally higher sensitivity than over
 single-coil approach. A potential issue with ASL CBF is
 sensitive delayed transit time, which could underestimate
 CBF in and around the occluded territory. A few transit
 time-insensitive ASL techniques are under development [16].

ASL has been applied to evaluate the spatiotemporal pro-
 gression of stroke rats during the acute phase [17, 18]. The
 ADC and CBF maps delineate regions of hypointense abnor-
 mality. Areas with ADC reduction grow from 30 to 180 min
 after ischemia, eventually reaching the CBF-defined lesion
 volume. In the permanent occlusion, ADC-defined lesion
 volume grows until it reaches CBF-defined lesion volume at
 about 180 min [17, 18], which correlated with the TTC infarct
 volume determined at 24 h. The ADC and CBF viability
 threshold in this rat stroke model is $0.53 \pm 0.02 \times 10^{-3} \text{ mm}^2/\text{s}$
 (30%±2% reduction) and $0.30 \pm 0.09 \text{ mL/gram/min}$ (57%±
 11% reduction), respectively [17, 18]. Reperfusion performed
 at 60 min post occlusion demonstrates the perfusion–diffusion
 mismatch was salvaged, with ADC lesion volume at 180 min
 reaching ~50% of the permanent occlusion group [18, 19].
 The degree of salvaged mismatch tissue is dependent on
 occlusion durations as expected [20]. Similarly, a few treat-
 ment drugs have also demonstrated to be effective in reducing
 infarct volume by salvaging the perfusion–diffusion mismatch
 defined by the viability thresholds in rat stroke models [21,
 22]. Quantitative diffusion and perfusion multi-slice imaging
 of the entire rat brain can now be acquired within a few
 minutes (~5 min) and can be longitudinally performed to
 evaluate ischemic evolution.

157 **Clustering Approaches for Delineating Tissue Fates**

Analysis of the CBF–ADC scatterplot offers additional in-
 sight that is not readily evident by inspecting the ADC and

160 CBF per se [17]. In the normal hemisphere, there is a single
 161 cluster with high ADC and CBF. In the ischemic hemisphere
 162 at 30 min, there are three clusters, namely, (1) the “normal”
 163 cluster with normal CBF and ADC; (2) the “core” cluster
 164 with markedly reduced CBF and ADC; and (3) the “mis-
 165 match” cluster with reduced CBF but slightly reduced ADC.
 166 At 180 min, essentially all the mismatch pixels migrated to
 167 the core in the permanent occlusion model. Upon reperfu-
 168 sion [19], the majority of the mismatch pixels and some core
 169 pixels returned to normal. Tissue volumes, ADC and CBF
 170 values of each tissue cluster on the CBF–ADC scatterplots
 171 can be objectively determined using cluster analysis and
 172 each cluster can be mapped back onto the image spaces.

173 Automated cluster analysis of the CBF–ADC data has
 174 been used to objectively cluster pixels of different tissue
 175 fate. Shen et al. developed and applied an improved algo-
 176 rithm based on the automated ISODATA (self-organizing
 177 data analysis algorithm) technique [23] to characterize the
 178 spatiotemporal dynamic evolution of ischemic brain injury
 179 based on high-resolution, quantitative perfusion and diffu-
 180 sion measurements. In contrast to the normal left hemi-
 181 sphere, multiple clusters were resolved in the ischemic
 182 right hemisphere, corresponding to the “normal”, “at risk”
 183 (perfusion–diffusion mismatch), and “ischemic core” tis-
 184 sues. The unique advantage of the ISODATA is that the
 185 number clusters in the dataset can be statistically determined,
 186 where other methods such as *K*-mean requires a priori assign-
 187 ment of the number of clusters. Tissue volumes, ADC, and
 188 CBF of each ISODATA cluster were quantified. Pixels of
 189 different ISODATA clusters were color-coded and mapped
 190 onto the image and ADC–CBF spaces. In some animals,
 191 essentially all the perfusion–diffusion mismatch pixels disap-
 192 peared, while in other animals some mismatch pixels persisted
 193 at 180 min after occlusion. CBF of the “persistent mismatch” at
 194 180 min was statistically higher than the CBF from the anal-
 195 ogous region where the mismatch disappeared at 180 min. The
 196 ADC of the persistent mismatch did not decrease as ischemia
 197 progressed. In marked contrast, the ADC of analogous brain
 198 regions where the mismatch disappeared at 180 min decreased
 199 precipitously as ischemia progressed. Upon reperfusion, the
 200 majority of the mismatch pixels and some core pixels migrated
 201 to the normal clusters. Automated cluster analysis allowed
 202 objective classification of different tissue types and these tissue
 203 types can be mapped back onto the image spaces, providing a
 204 powerful and objective means for pixel-by-pixel visualization
 205 of different tissue fate. MRI is noninvasive and thus is ideally
 206 suited for longitudinal imaging in the same animals [24, 25].

207 **Spatial Resolution Versus Temporal Resolution**

208 It is important to have fast imaging techniques with high spatial
 209 resolution to distinguish different tissue types in ischemic

stroke. Partial volume effect (PVE) could hamper proper de- 210
 211 lineation of normal, ischemic, and at-risk tissues by blurring
 212 the boundaries among different tissue types and tissue viability.
 213 Visual delineation of ischemic lesions by manually drawing
 214 regions of interest on the diffusion- and perfusion-weighted
 215 images is a common clinical practice and the presence of PVE
 216 could lead to significant errors in identifying ischemic tissue
 217 fates. In addition, it is conceivable that a substantial number of
 218 pixels with mild ADC and CBF reduction could arise simply
 219 from the physical effect of partial voluming, thereby confound-
 220 ing the interpretation of the operationally defined ischemic
 221 penumbra. High-resolution imaging could minimize tissue
 222 classification errors. Other advantages of high-resolution im-
 223 aging include finer delineation of anatomic structures and
 224 increase in pixel density, which increases the statistical power
 225 of pixel-by-pixel cluster analysis, and reducing signal loss due
 226 to intravoxel dephasing. The drawbacks of higher spatial res-
 227 olution are longer acquisition time and/or reduced SNR, which
 228 could also hamper efficacy of the imaging method. With
 229 improvement in parallel imaging and RF coils, faster and
 230 higher spatial resolution MRI protocols are expected.

231 Ren et al. evaluated that ADC and CBF standard devia-
 232 tions in the normal left hemisphere were comparable be-
 233 tween high and low resolution [26, 27], despite increased
 234 noise and tissue heterogeneity at high-resolution, substantial
 235 PVE was observed along the normal–abnormal boundaries
 236 on the ADC and CBF maps, PVE resulted in overestimation
 237 of the abnormal tissue volumes at the expense of at risk and/
 238 or normal tissues, and misclassified pixels were quantita-
 239 tively evaluated on a pixel-by-pixel basis; PVE appeared to
 240 be more severe at the early time points postischemia and
 241 further reduction in spatial resolution and zero-filling
 242 resulted in more severe PVE. This study showed that there
 243 are some advantages to acquire stroke data at higher spatial
 244 resolution for the same scan time. Future study needs to
 245 evaluate whether the improved spatial resolution improves
 246 separation of different tissue types.

247 **BOLD fMRI of Perfusion–Diffusion Mismatch**

248 In addition to anatomical MRI techniques based on tissue
 249 perfusion and diffusion, functional MRI of stroke animals
 250 can also be performed to evaluate the functional status of the
 251 perfusion–diffusion mismatch. fMRI is a noninvasive imaging
 252 modality and has been widely exploited for mapping brain
 253 processes, ranging from perceptions to cognitive functions
 254 [28]. The most widely used fMRI technique is based on the
 255 BOLD signal or CBF signal. The BOLD contrast originates
 256 from the intravoxel magnetic field inhomogeneity induced by
 257 paramagnetic deoxyhemoglobin in red blood cells. Changes
 258 in regional deoxyhemoglobin content can be visualized in
 259 susceptibility-sensitized (i.e., T_2^* weighted) BOLD images.

260 The BOLD fMRI technique is based on a principle discovered
261 over 100 years ago [29] that neuronal activity is intricately
262 coupled to cerebral blood flow. When a task is performed,
263 regional blood flow increases disproportionately (which can be
264 measured using the ASL technique), overcompensating the
265 stimulus-evoked increase in oxygen consumption needed to
266 fuel the elevated neural activity and, thus, resulting in a
267 regional reduction in deoxyhemoglobin concentration. Thus,
268 the BOLD signal increases following elevated activity relative
269 to basal conditions, making it possible to dynamically and
270 noninvasively map changes in neural activities.

271 fMRI applications to neurological diseases in animal mod-
272 els are emerging. We and others have previously demonstrated
273 that bilateral forepaw somatosensory stimulation activated the
274 somatosensory cortices of both hemispheres in a normal rat
275 using isoflurane as the anesthetics [30, 31], instead of the more
276 common α -chloralose [32, 33]. Recent development [30] also
277 allows the addition of oxygen consumption imaging to map
278 oxidative metabolism and neural-vascular coupling in stroke
279 rats. In the stroke rat 30 min after occlusion, we demonstrated
280 that activations in the somatosensory cortices were not
281 detected in the ischemic hemisphere [20]. Functional MRI in
282 stroke should be useful in determining whether risky therapeutic
283 intervention should be performed if the perfusion-diffusion
284 mismatch is already nonfunctional. Perfusion, diffusion, and
285 functional (including oxygen consumption) imaging can be
286 carried out within 30 min at reasonably high spatial resolution.

287 Quantitative Prediction of Ischemic Tissue Fate

288 The ultimate goal of acute stroke imaging is to predict tissue
289 fate based on acute MRI data. Sophisticated algorithms have
290 been developed to predict ischemic tissue fate on a pixel-by-
291 pixel basis. They included predictive models based on gen-
292 eralized linear model [34, 35], probability-of-infarct [36,
293 37], and artificial neural network (ANN) [38] and support
294 vector machines (SVM) [39]. These predictive models pro-
295 vide statistical or probabilistic maps of infarct likelihood on
296 a pixel-by-pixel basis utilizing only the acute MRI data.
297 Performance analysis showed accurate prediction when
298 compared with endpoint T2 MRI and/or histology. In addi-
299 tion, the effects of neighboring pixels and infarct incidence
300 on prediction accuracy were also evaluated. Other ~~potential~~
301 a priori information can be incorporated in these predictive
302 models. Prediction accuracy was quantified using receiver-
303 operating characteristic analysis.

304 Wu et al. predicted infarction in normal and hypertensive
305 stroke rats subjected to embolic clot occlusion with and
306 without rtPA treatment at 1 h after stroke using voxel-
307 based generalized linear model algorithm [34]. They found
308 that pretreatment predicted outcome compared with post-
309 treatment histology was highly accurate in saline-treated rats

(92±5%). Accuracy was significantly reduced in rtPA- 310
treated animals (86±8%). Animals that reperfused had sig- 311
nificantly lower predicted infarction risk than nonreperfused 312
animals, suggesting that tissue was more amenable to ther- 313
apy. Shen et al. [36, 37] documented that the probability-of- 314
infarct profiles of stroke rats underwent different middle- 315
cerebral artery occlusion (MCAO) durations. Using only 316
acute ADC and CBF data, pixel-by-pixel prediction was 317
made and compared to endpoint T2 imaging and histology. 318
The AUCs were 87±3%, 90±4%, and 93±3% using ADC+ 319
CBF for the 30-min, 60-min, and permanent MCAO, re- 320
spectively. Huang et al. [38] used ANN prediction algo- 321
rithms and found that the AUCs were 86±3%, 89±2%, 322
and 93±1% using ADC+CBF for the 30-min, 60-min, and 323
permanent MCAO, respectively. Adding neighboring pixel 324
information and spatial information improved performance 325
measures over ADC and CBF alone for the 60-min and 30- 326
min MCAO groups (88±3% and 94±1%, respectively) but 327
only slightly for the permanent MCAO group (94±2%). 328
These differences were expected because permanent MCAO 329
was less variable and ADC and CBF alone sufficiently 330
accounted for prediction accuracy. ANN method performed 331
slightly better than the probability-of-infarct method [37] 332
operated on the same data sets although there were some 333
minor methodological differences in how training groups 334
were assigned. 335

336 Huang et al. used SVM prediction algorithms for predicting 336
tissue fate and found that the AUCs were 86±2.7%, 89±1.4%, 337
and 93±0.8% using ADC+CBF for the 30-min, 60-min, and 338
permanent MCAO, respectively [39]. They found that CBF+ 339
ADC improved prediction accuracy. This is likely because CBF 340
and ADC individually provided unique and relevant informa- 341
tion. For example, in the presence of the perfusion and diffusion 342
mismatch which would likely infarct at later time points, neither 343
ADC nor CBF data alone can capture such information. As 344
such ADC would underestimate infarct volume while CBF 345
overestimate infarct volume in this case. Moreover, perfusion 346
deficit could overestimate final infarct volume if benign olige- 347
mia exists or reperfusion salvaged some tissue with initial 348
perfusion deficit. Moreover, adding neighboring pixel informa- 349
tion and spatial information markedly improved performance 350
measures over ADC and CBF alone for the 60-min and 30-min 351
MCAO groups (94±0.8% and 88±2.8%, respectively) but 352
again only slightly for permanent MCAO group (97±0.9%). 353
The improvement in SVM results was apparent when compar- 354
ed with ANN operated on the same datasets. Differences in 355
animal stroke models (embolic vs suture), anesthetics (halo- 356
thane vs isoflurane), and inclusion of slightly different types of 357
MRI data (dynamic susceptibility contrast vs arterial spin la- 358
beling CBF) preclude quantitative comparison with results 359
reported by other research groups. Nonetheless, these quantita- 360
tive prediction models (general linear model [34], probability- 361
of-infarct method [37], ANN model [38], and SVM [39]) based 362

363 on acute MRI data were overall accurate and yielded compa-
 364 rable AUCs on animal stroke models.

365 **Hyperperfusion**

366 Postischemic hyperperfusion (HP)—also known as “luxury
 367 perfusion” or “hyperemia” in which blood flow exceeds
 368 metabolic needs in the brain—has long been documented
 369 [40] to be a frequent, yet poorly understood, phenomenon.
 370 HP has been studied using positron emission tomography
 371 and MRI techniques in animal stroke models [41, 42] and
 372 stroke patients [43, 44]. Early postischemic HP sometimes
 373 observed immediately after recanalization is a hallmark of
 374 efficient recanalization after stroke [45, 46] and it has been
 375 reported to be both beneficial (i.e., salvage tissue in and
 376 around the ischemic zone or prevent infarct growth) [47]
 377 and harmful (i.e., aggravate edema and hemorrhage, and
 378 neuronal damage from reperfusion injury) [48, 49]. By contrast,
 379 late postischemic HP (48 h after onset) is often associated with
 380 tissue necrosis [50–53]. Many studies have investigated the
 381 mechanisms underlying HP. Accumulated by-products (such
 382 as free radicals) could result in delayed neuronal death as well
 383 as production of vasoactive metabolites (such as lactic acid and
 384 adenosine) that could induce vasodilation through relaxation of
 385 vascular smooth muscle [54, 55]. Some of these metabolites are
 386 implicated in modulating blood–brain barrier (BBB) perme-
 387 ability [56], potentially enhancing cerebral edema. Others have
 388 suggested neurogenic vasodilation [57] and passive physiolog-
 389 ic coupling [47]. Histopathological investigation of stroke cat
 390 showed that late hyperperfusion in the necrotic core could in
 391 part reflects neovascularization with increased capillary density
 392 and endothelial hypertrophy [58]. However, the underlying
 393 spatiotemporal characteristics of postischemic HP and its pro-
 394 gression with respect to other imaging markers (such as T1, T2,
 395 diffusion and contrast-enhanced MRI) remain incompletely
 396 understood. Most published noninvasive longitudinal studies
 397 to characterize HP had limited time points and terminal studies
 398 at different time points were confounded by intersubject varia-
 399 tions. Improved understanding of the HP spatiotemporal char-
 400 acteristics with respect to other imaging markers could improve
 401 understanding of stroke pathophysiology, which could ulti-
 402 mately lead to improved clinical stroke management.

403 Tanaka et al. examined how changes in tissue spin–lattice
 404 relaxation time constant, BBB permeability, and arterial
 405 transit time affect CBF quantification by ASL and dynamic
 406 susceptibility contrast (DSC) in postischemic hyperperfu-
 407 sion in same rats (Tanaka, 2011 #3608). Embolic stroke rats
 408 imaged 48 h after reperfusion showed reliable regional
 409 hyperperfusion. ASL- and DSC–CBF of normal pixels lin-
 410 early correlated whereas in ASL–CBF of hyperperfusion
 411 pixels were higher than DSC–CBF. T1 of hyperperfusion
 412 pixels were higher, transit time was shortened, and ΔR_2^*

time courses showed gadolinium diethylenetriaminepentaace- 413
 tate (Gd-DTPA) leakages in hyperperfusion regions. Hyper- 414
 capnic inhalation, which does not change BBB permeability, 415
 showed overall CBF increase but ASL- and DSC–CBF remain 416
 linearly correlated. Mannitol injection, which increases BBB 417
 permeability, showed ASL–CBF to be higher than DSC–CBF. 418
 Tanaka et al. concluded that: (1) under normal conditions the 419
 commonly used ASL and DSC provide comparable quantita- 420
 tive CBF values, and (2) in ischemic hyperperfusion, T1 and 421
 BBB disruption were responsible for discrepancy in CBF 422
 measured by ASL and DSC. 423

424 Shen et al. also longitudinally evaluated the spatiotem-
 poral dynamics of late HP in same animals subjected to 30- 425
 min, 60-min, and 90-min intraluminal MCAO in rats [59]. 426
 Multi-parametric MRI data including diffusion, perfusion, 427
 T₂, T₁, dynamic susceptibility contrast MRI, and MR angi- 428
 ography were acquired longitudinally at multiple time 429
 points up to 7 days after stroke. The spatiotemporal progres- 430
 sion of HP was compared with T1, T2, diffusion, angio- 431
 graphic, and BBB changes. The main findings were as 432
 followed. The early HP within 3 h of recanalization was 433
 not detected in all three MCAO groups. The late (≥ 12 h) HP 434
 was present consistently in the 30-min MCAO group, pres- 435
 ent in half of the animals in the 60-min MCAO group, and 436
 absent in the 90-min MCAO group. DSC, CBF, MRI, and 437
 MRA independently corroborated HP detected by cASL. 438
 HP preceded T2 increase in some animals, and HP and T2 439
 changes coincided in others. T2 peaked first at 24 h whereas 440
 HP peaked at 48 h post occlusion, and HP resolved by day 7 441
 in most animals at which point the arteries also became 442
 tortuous. HP was exclusively associated with poor outcome 443
 whereas tissue that was not infarcted did not show HP. 444

445 **Oxygen-Challenge MRI**

446 A novel approach was recently introduced to further probe
 tissue at risk by using T₂*-weighted MRI of transient oxygen 447
 challenge in ischemic stroke [60, 61]. T₂*-weighted signal 448
 intensity is sensitive to relative concentration of deoxyhemo- 449
 globin [62, 63]. The infarct core showed little or no change in 450
 T₂*-weighted signal intensity during oxygen challenge (OC). 451
 The at-risk regions surrounding the infarct core showed an 452
 exaggerated increase in OC T₂*-weighted signal intensity 453
 compared to the homologous region in the contralateral hemi- 454
 sphere. OC brings in oxygenated blood displacing the high 455
 deoxyhemoglobin concentration in the at-risk region where 456
 CBF is partially compromised but its metabolic activity 457
 remains significant. It was thus hypothesized that tissue with 458
 exaggerated increase in T₂*-weighted signal intensity during 459
 OC is potentially salvageable [60, 61]. 460

461 Shen et al. characterized the effects of transient oxygen
 challenge on T₂*-weighted signal intensities in permanent 462

463 stroke rats [64]. The major findings were as follows. The
 464 ischemic core cluster, derived automatically from perfusion
 465 and diffusion data, showed no significant response, whereas
 466 the mismatch cluster showed markedly higher percent
 467 changes relative to normal tissue in the acute phase. The
 468 exaggerated OC responses are more apparent in the primary
 469 somatosensory cortex than other brain structures at risk in
 470 this stroke model. Many of the mismatch pixels showed
 471 some exaggerated OC responses which became hyperin-
 472 tense on T₂-weighted MRI at 24 h. Basal T₂*-weighted
 473 signal intensities on the perfusion–diffusion contourplot
 474 were high in the normal cluster and low in the core cluster,
 475 with a sharp transition in the mismatch cluster. OC-induced
 476 changes on the perfusion–diffusion contourplot dropped as
 477 perfusion and diffusion values fell below their respective
 478 viability thresholds. (1) Basal T₁ increased slightly in the
 479 ischemic core. OC decreased T₁ significantly in the normal
 480 hemisphere, indicative of hyperoxia-induced vasoconstric-
 481 tion or increased dissolved oxygen in the plasma, but OC
 482 had no significant T₁ effect in the ischemic core and mis-
 483 match pixels. (2) OC decreased CBF significantly in the
 484 normal hemisphere but had no significant CBF effect on
 485 the mismatch and the ischemic core pixels.

486 **Conclusions and Perspectives**

487 This review summarizes our recent works on multimodal MRI
 488 of the perfusion–diffusion mismatch. The combined use of
 489 perfusion, diffusion, physiological, and functional MRI and
 490 image analysis methodologies provides powerful tools to im-
 491 prove characterization of cerebral ischemia, to longitudinally
 492 monitor ischemic progression, evaluation of drug efficacy, and
 493 statistically predict ischemic tissue fate. Animal stroke models
 494 in which the perfusion–diffusion mismatch and its functional
 495 status can be reproducibly studied under controlled conditions
 496 are highly valuable for establishing exploring novel MRI mo-
 497 dalities and to characterize ischemic tissue fate. A major con-
 498 found of animal stroke studies is the need for anesthetics which
 499 could have undesirable (i.e., neuroprotective) effects. However,
 500 with the use of proper controls and careful analysis, this con-
 501 found can be mitigated or at least minimized. Translations of
 502 these methodologies require vigorous testing and improvement
 503 in temporal efficiency of these MRI protocols. Noninvasive
 504 MRI methodologies can be readily be applied to study stroke in
 505 high-order animals, such as non-human primates [65, 66], as
 506 well as in humans.

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