WF *6 $W - - A\hat{e} \cdot KW >$

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学位论文题目:				
作者签名:	日期:	年	月	日

摘 要

<!• Ê } > ` . ´ * Ä Simultaneous Localization and Mapping, SLAM Å _ j ~ êN¶ ,´£K^°_{0 È k?±@ \tilde{a} j~ \hat{e} 0+ ØE÷0;],´Ê}>)ß 3 O. L NÈ Ä+a ¾?ö?ù P O ~,´) (© H ï È?ö?ù SLAM @ jF ¤,´.D0¦' &é Ä* ¾(© ±,´-(jD DZ é# _ ...- ?ö ?ùSLAM] 0 h+X,´° _ È-O2î.ž``AÑ *-(j}/¦ØÝ*jŸ,´ ...5 ´Ä6<È $(\textcircled{0} \pm))$ ß 3,´ (1 + 1); $(\textcircled{0} \pm)$)ß 3,´ (1 + 1); $(\textcircled{0} \pm)$; $(\textcircled{0} \pm)$; ?ùSLAM é# °° ^ X-p > n, ´00+¿ WL NÈ È8\$¦) j Ÿ, ´ O. pL€ ¾ ...5 ´ rM' b6< # "# Ä+X ¾Gb F j Ÿ Ò#{ È8 k ,8Z1y ÎLu + Ñ] Ä) ¾MŽ5 ´ W j ,´0P ö \$! Ö9c Èōb ...@ Ö•B\$°° Lî C`@ ã Ä §0 /Ž4ÿ5•5 Ä Convolution Neural Network, CNN ÅX $j \ddot{Y}$, $\dot{Q} r \ddot{Y} \frac{1}{4}F > |\ddot{y}3P4 \times [, ', + \tilde{N} \acute{e}M'] \frac{4\ddot{y}}{4} \ddot{Q} \parallel W$,´@ Ï È ¦ Ø Ý,´\$! Ö _0P ö D ~ p2î.ž,´ È >D DZE÷0;],´5 ´ Ÿ-(Â>• Ä < & CNN, '\$! Ö O. é ? È X ~4é*6 j Ÿ; § 9 \$ -, 'R± W \·Y+X * ¾(© ±, ´-(jD DZ ¼ CNN \$! Ö Ø Ý È 6 ... é# ¼\$! Ö - • M #-(5 8 È þ\$! Ö=½ 8@ Ö.D0¦ ...- ?ö?ù SLAM , OP öGý *L NÈ Ä k?± @ Ì 5 Ö Ä1 ÅAĵAÑ ¦ ´ * ¶ 0 ‡ ¼ ¤, ´M' A0P öGý *, ´ ...- ?ö?ù SLAM 3+5 ÈH F-(jD DZE÷0;], ´(© ± 1†+• ¦ E •Q x, 'B iG}1Ç# È Q }1] ž £6,,, '!".ž W ÄÄ2ÅX\$!Ö` AÑ éM' È * * 3/4 Resnet 5 ´,´ J j Ö\$! Ö `AÑ5•5 ÈÎ)à ¶MŽ5 ´ W j ,´OP ö\$! Ö 9ç ȦΡ¶3+5 X ~4é*6 ¡ Ÿ ¼4ß ûEœF Ø ;,´W7- Ä Ä3 Å=½ 8E÷0;] È5, 8 Y +X-(jD DZ, ´D 7&é ¼ CNN Ø Ý, ´\$! Ö Ÿ È u 663<• j Ö Q2úL NÈ ¼\$! Ö ž, ´ 2 W È ´*¶ * ¾ FO"r@ 1Ç#,´\$! ÖGý * v æ Ä

关键词:视觉 SLAM;深度估计;稠密重建;稀疏地图点;卷积神经网络

Monocular Visual SLAM towards Dense Reconstruction

Abstract

Simultaneous Localization and Mapping (SLAM) is one of the key technologies in the field of robotics, mainly focusing on localization and environment perception. Due to the unique advantages of visual sensors, visual SLAM has become a research hotspot in recent years. Feature-based camera tracking is the most widely used technology in monocular visual SLAM, and is capable of accurately tracking camera pose and inferring structure of the environment. However, features are heavily dependent on the environment, and are not performing well in textureless scenes. So, feature-based monocular visual SLAM always suffers from sparsity problem in reconstructed map, causing its sensing ability limited at the level of scene structure, which cannot be used in practical task, such as field scene detection and autonomous driving. Dense depth map estimation in non-structural area cannot be solved well only from the aspect of pure geometry theory. Convolutional neural network (CNN) has made a great progress on extracting high level feature and regression task of pixel level. In general, CNN-inferred depth is dense and globally accurate, which is complementary with structure data in process of camera tracking. Meanwhile, it has a better robustness because of characteristics of CNN architecture in low-texture scene.

In this thesis, the monocular visual SLAM towards dense reconstruction has been addressed in term of depth fusion, which makes full use of feature-based camera tracking and CNN-inferred depth, and combines the traditional geometry methods with the concept of deep learning. The main research work includes: (1) We design and implement a complete Visual SLAM system towards dense reconstruction. An improved feature extraction strategy and wrong match culling algorithm are introduced to raise the accuracy of data correspondence. (2) In the dense depth estimation, an original CNN architecture with multi-scale module based on Resnet has been proposed, which enhances the performance of whole system under pure rotational camera motion and low-texture scene. (3) The scale ambiguity of monocular SLAM and uncertainty of depth map are well processed in the framework of depth fusion and reconstruction speeding up by a fast solver.

Key Words Övisual SLAM; Depth Estimation; Dense Reconstruction; Sparse Map Points; CNN Architecture

目 录

^	?	'±	I
Ał	ostract		II
1	5 Aê	······	. 1
	1.1	B.NÈ.D0¦6ü Ÿ ¼ ? y	. 1
	1.2	?ö?ùSLAM L NÈ ²F	. 3
		1.2.1 SLAM L NÈ], F Ø >?ò#{ Q »	. 3
		1.2.2 ?ö?ùSLAM ,´ >.D0¦)à(æ	. 5
	1.3	M' A0P öGý *,´ ?ö?ù SLAM .D0¦	. 8
	1.4	\ ·,´ k?± µ é	10
2	-(£,	*.p*6Aê >6ü Ÿ. Aö	12
	2.1	-(j Q » >) ±	12
		2.1.1 J _s "-(j Q »	12
		2.1.2) ± > *.p. Le	13
		2.1.3 \CX. Le 6@	15
	2.2	SLAM ,´ .>~/j	16
		2.2.1 Ø 1CM & ß5•5	16
		2.2.2 €	17
		2.2.3 Pœ D0 [L¿jj	17
	2.3	0 ? ¼ ^ H F	18
		2.3.1 0 ? ¼ ^ H FL NÈ ²F	18
		2.3.2 4ï W 0 ? ¼ ^ H F	19
		2.3.3 MŽ4ï W 0 ? ¼ ^ H F	19
	2.4	\1 ?5	23
3	* 3/4((© ±, ´-(jD DZ	24
	3.1	£ ¾ ?ö?ù SLAM Î)à0P öGý *,′ M63	24
	3.2	* ¾(© ±,´-(jD DZ1Ç# AîAÑ	27
	3.3	* ³ ⁄ ₄ 5• I, ´(© ± ÿF > iG}	28
		3.3.1 * ¾5• l,′(© ±	28
		3.3.2 * 3/45• I, 'B iG} "L"	30
	3.4	`. M û F	32
	3.5	-(jD DZ	33

3.5.1 WL\$D DZ	33
3.5.2 pG `.D DZ	33
3.6 £K^ W.ž Ê	34
3.7 `.1Ñ*6 ¼ pG H F	34
3.8 \1 ?5	35
4 M' A0P öGý *, ´\$! Ö `AÑ	36
4.1 * 3/4 Resnet 5 ´,´ J j Ö\$! Ö `AÑ	36
4.2 * ¾=½ 81†+•,´0P ö\$! ÖGý *	37
4.2.1 RANSAC væ;,′jÖ, 0F	38
4.2.2 \$! Ö=½ 8 >Gý *	39
4.3 L)ß ð#{ > Q!"	42
4.3.1 L)ß ð#{	42
4.3.2 L)ß Q!"	43
4.4 \1 ?5	44
5 ÎP¼ > 6 À	45
5.1 ÎP¼Aî5ž > žLö û4ý	45
5.1.1 ÎP¼Aî5ž >B\$ >	45
5.1.2 ÎP¼ žLö û4ý	45
5.2 -(jD DZ >L)ß ð#{ ÎP¼	47
5.3 OP ö\$! ÖGý * ÎP¼	48
5.3.1 ò F2G} > j Ö €"r@ ÎP¼	
5.3.2 \$! ÖGý *, ´0P ö WAô`	50
5.3.3 \$! ÖGý *,´2î ÖAô`	51
5.3.4 Q ‡ 9 x W, P¼Añ	54
5.4 F > &L\$ 6 À	56
5.5 \1 ?5	56
5 Aê	58
ò 63 ·)^	
kB+ − } OL\$ >~ − _Aê · õ å	64
8\$ BR	65
WF *6 W − − }Aê ·(x s ¯+X ¸ s	66

1 绪论

1.1 课题研究背景和意义

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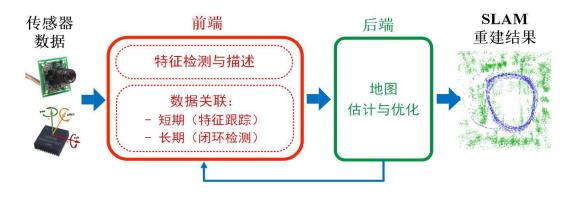


. 1.1 NASA & O Ò#{~ÃBg!| êE-¼ ÞEŸ,′ HoloLens Fig. 1.1 Mars rover, Google autonomous car and Microsoft HoloLens

`Y0ª Ä m1Ñ þ*6Aê ¼ ² % rM': •B\$ È SLAM L NÈ >ÛAÔ j¸-,´@ ã ¶ Ä v _ X ´* 0 ZFJ+X,´ SLAM 3+5 éM' Í'f 9¸ J%Œ X,´L NÈ ÈT ¦ _ > g Y+X ¼9ç ` ü,´ ` . Ÿ Ä

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1.2 视觉 SLAM 问题概述



. 1.2 SLAM E÷0;, " » v æ5 ' /j?.

Fig. 1.2 Illustration to typical framework of SLAM process

1.2.1 SLAM 问题中的运动与观测模型

f j $^{\circ}$ ê X Z. j $^{\circ}$ j $^{\circ}$ O#{ & È+a 3 4 # 9¢)ß $^{\circ}$, $^{\circ}$ x P½ $^{\circ}$ Ep $^{\circ}$ N« $^{\circ}$)à < & Ê } > $^{\circ}$.Gý $^{\circ}$ ÄFÓ x È þ $^{\circ}$ O $^{\circ}$ ·; ÈF T Z Q $^{\circ}$ 6 [) Ä ¶ SLAM L NÈ], $^{\circ}$ T Z Q $^{\circ}$ OF Ø >?ò#{ $^{[10]}$ Ä m1Ñ j $^{\circ}$ ê $^{\circ}$ O#{E÷0; _F 5 , $^{\circ}$ È v $^{\circ}$ P O $^{\circ}$ G÷ gNÁ)·, $^{\circ}$ L ∈ f È A 6F!åF 5 , $^{\circ}$ F Ø E÷0; 6@ @/ë " & k t = 1,2 ... K & +O, $^{\circ}$ F Ø È p A M0?± £# F Ë & k;, $^{\circ}$ }/ > $^{\circ}$.Ä A +X x > $^{\circ}$ /j j $^{\circ}$ ê, $^{\circ}$ } / ÈFÓ x j $^{\circ}$ ê $^{\circ}$ ZF Ø E÷0;], $^{\circ}$ E $^{\circ}$ O;], $^{\circ}$ E $^{\circ}$ B +X $^{\circ}$ X₁, $^{\circ}$ X₂, $^{\circ}$ X₃ ... Y_N Ä

 $\delta x_F \mathcal{O} \hat{\mathbf{U}} F \mathcal{O} \acute{e}0; k+k,'_=< \& kj^ê \}/ x,' F \ddot{A} FJh \~o å; j^ê Jj VF <math>\mathcal{O} \# G \mathcal{O} \sim ^2G \mathcal{O}; A \tilde{N} F 65 W \# G \mathcal{O} ... s \dot{E} \tilde{S} \mathcal{O} \# G \mathcal{O} ... s \dot{E} \tilde{A}$ • [11] $\dot{E} > \tilde{C} / \tilde{I} \otimes \tilde{C}$

$$\mathbf{U} = \{ \boldsymbol{\mu}_1 \ \boldsymbol{\mu}_2 \ \boldsymbol{\mu}_3 \ , ... \boldsymbol{\mu}.$$
 Äl. 1 Å

Ij~ê X T ZF 5 & kL\$ } /-() ¾F Ø#{Gÿ I, ´£3+ +a ;M', ´MŽ4ï WF Ø Q »>~/j Ö

$$x_{t+1} = f(x_t, u_t) + w_t$$
 Ä1. 2 Å

FØPO~3+5,′š FJh>Û*Q@ 0 wlé j →,′Q ß 63 È+a š N© w_t >″/jÖ w_t \square $N(0,\Sigma_t)$ Ä1.3 Å

!" È t+1 & k } / • >Û k+k @Q ß 6 3 Ö

$$x_{t+1} \square N(f(x_t, u_t), \Sigma_t)$$
 Ä1. 4 Å

$$x_{t+1} = f^{CTRV}(x_t, u_t) = x_t + \begin{cases} v_t \cos(\theta_t) \Delta t \\ v_t \sin(\theta_t) \Delta t \\ 0 \end{cases}, \omega_t = 0 \\ \frac{v_t}{\omega_t} (\sin(\theta_t + \omega_t \Delta t) - \sin(\theta_t)) \\ \frac{v_t}{\omega_t} (\cos(\theta_t) - \cos(\theta_t + \omega_t \Delta t)) \\ \frac{v_t}{\omega_t} \Delta t \end{cases}, \omega_t \neq 0$$

ð x _?ò#{ Û ?ò#{ é0; ÿF ,´ _ X t & k È j ~ ê ¾ x_t 4)D 7&é y_i ×+O?ò#{ l $z_{t,i}$,´E÷0; Ä j ~ ê j V,´?ò#{ P O ~ ,´?ò#{5 Ì > ~/j @ Ö

$$Z = (z_{11}, z_{12}, ..., z_{1N}, z_{21}, ..., z_{ti}, ..., z_{KN-1}, z_{KN})$$
 Ä1. 6 Å

lj~êX t&k)D 7&é y_i ,´?ò#{l $z_{t,i} >$ }/ x_t ,´£3+²;Ö

$$Z_{t,i} = h(x_t, y_i) + \lambda_{t,i}$$
Ä1. 7 Å

?ò#{ P O ~3+5 ,´ š FJ h>Û * Q @ 0 w l é j " ,´Q ß 6 3 È+a š N© $\lambda_{t,i}$ >~/j Ö

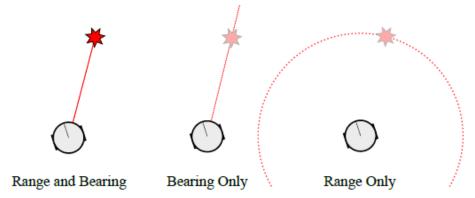
 $\lambda_{ti} \square N(0, \Lambda_{ti})$

Ä1.8 Å

ΙÖ

 $z_{t,i} \square N(h(x_t, y_i), \Lambda_{t,i})$ Ä1. 9 Å

>F Ø - f 0 g È?ò#{ Q »],´ íB' - h ,´ § f '? ã ¾?ò#{ P O ~ ,´/ý2« Ä 08\ õ å ; È J¸) = <,´ ?ò#{ P O ~D 7&é,´?ò#{ Q ? 6 @ 92« Ö93 \$ ¼ é } ÄRange and Bearing Å È4ß é } Ä Bearing Only Å ¼4ß93 \$ ÄRange Only Å [11] È ² . 1. 3 p/j Ä



. 1.3 D 7&é,´9/ý?ò#{ Q ?

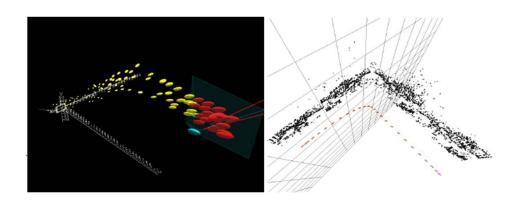
Fig. 1.3 Three different modes of landmark observations

f?ò#{PO~j...--(j&È 7-9ç ?ö?ùD 7&é,´é}È9 L¿-pj~ê,´F ØÈ
BD 7&é>Û ¼!Q?ò#{`È!!"& FJE÷F Ø?ö •.žÊD 7&é,´93 \$£D /ëj~ê,´
D /ë Ä ²Ì?ò#{PO~_ ü--(j F65 _ RGB-D -(j &È I -\$ Õ#{Gÿ *93 \$ > é}È
v ~+X ~*3 ,8Z > O3+5 ÄGNSS Å F65 Wi-Fi & a 7-9ç 93 \$Ä

1.2.2 视觉 SLAM 的发展与研究现状

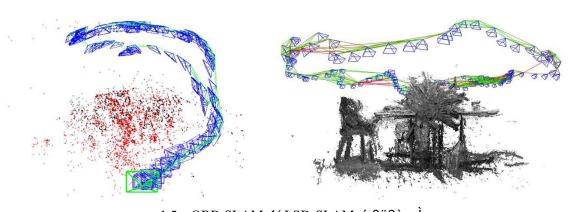
?ö?ùSLAM þ 0 M ²)- f3+ ;, ´(æ 1 `AÑ ... `)à X, ´MŽ4ï W H F"r@ È h ó μ é ¦"Ñ 9 +O Z W, ´ F Äk#q, ´?ö?ù SLAM ° _ W8\$4ÿ ¶ ¶ 9 Z ...Lf!å Ö* ¾% # ~Lf!å Ã * ¾ £K^ W H FLf!å ¼ * ¾-\$ ÕD DZLf!å Ä

MonoSLAM^[12] _ * ¾ P5 CM & ß% # ~ é# ,´ " » > È+a Davison 1y ê ¾ 2007 $\cong 0$ È _OÆ Z õ ÍC† ... - -(j,´?ö?ù SLAM v æ Ä ¯+X Shi-Tomasi @ & $\stackrel{!}{e}$ ³] • k+k)ß ³], ´00+¿(© ± Èà ð#{ `,´(© ± = Ý > ¶ "(© ±F >| iG} È 6<9ç j Ÿ,´\$! Ö Ÿ È ' = *)ß ³,´ 95\$5 ´ È | ¯+X ™ ... ' D ,% # ÄExtend Kalman Filter ÈEKF Å)3+5 ,´(æ 1F >| \$ à Ä +a ¾ X3+5 (æ 1 `AÑ] È j ¶1° FAÑ1Ç È!ÿ 0 & k,´`AÑ 63<• f }

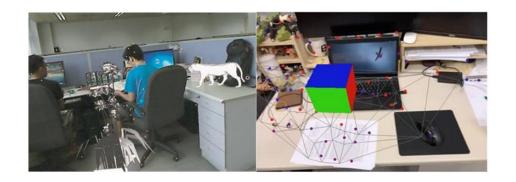


. 1.4 MonoSLAM ¼ PTAM , Gý *5 Ì
Fig. 1.4 The reconstruction results of MonoSLAM and PTAM

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. 1.5 ORB-SLAM ¼ LSD-SLAM ,´ ?ö?ù x Ì Fig. 1.5 The visual effects of ORB-SLAM and LSD-SLAM



. 1.6 RDSLAM ¼ VINS-Mono X AR], ' $\ddot{A}+X$.../j Fig. 1.6 The effects of RDSLAM and VINS-Mono in application to AR

1.3 面向稠密重建的单目视觉 SLAM 研究

X 1.2 82, $\acute{}$ 0 d \grave{E} "4 \dddot{y} jB3 \P P5 SLAM > SFM , $\acute{}$ j [$\ifmmode 14 \ensuremath{\%}\xspace$ y = < \grave{E} F G \ddot{u} M0 ?±B\$ >, ´£ $\frac{3}{4}$ SLAM \$ D# Gý Ê }6< a F ¶ `.Gý * , ´> F 7 X.ž Ê }2î Ö Cã O ¼3+5 F > | x)·, ´ õ å ; • ÿ A) 9 } j Ÿ&é, ´Gý *Ä 08\>~)à j00+¿, ´D 7&é F (© \pm &é ` . Å È = $_{-}$ 7 * . \DÛ, ´Gý? \pm W>Û a F Ä -(ý È * . \pm Ñ _ SLAM 3+5 , ´ TWhó{0 \dot{E}^3 , \dot{k} ?± $(\dot{E}+\dot{X})$ à \dot{X} TéM'Ö \dot{O} \dot{E} \dot{x} .7-O) § f, \dot{f} + \ddot{N} \ddot{E} \ddot{o} ?±, \dot{f} 1 È SLAM, ´Ä+X j Ÿ; M0?± = < '?, ´`. • ¼ @(© Ê, ´ + Ñ x !!Q È`.7- OL€ f $j \tilde{e}(\approx 1 \tilde{A}\tilde{N}E \div 0;] \times +0, \hat{B} \dot{E}^2 \dot{I}\tilde{N} 9$. ÈFÓ x } /, ´`AÑ ¸ é C +0%20+ Ä ² ú È L¿-p SLAM µ é, ´ = Ý & h > Ä+X93 \$, ´ = Ý ™ W È Ê y ' ? • Ç \$ Đ&¥#k Ä SLAM 3+5 \$ D# Gý $| \ddot{A}$ +X $| \ddot{Y} \stackrel{.}{E} \tilde{o}N \ll i \overset{.}{z} = <, MO"r •) SLAM , G &F > | i Ø 1/4$ Î j Ä T ¦ X0P ö * . éM' È P5 , '00+ \dot{z} ` . # >Û Ä+X ` § f, 'ÎD% f] ÈC° •C° J ,´Ä+XG-M0?±0Z~p08\$,´0P ö`.È»²ÈXAèJË» F"A+OÄ+X]j~ê,´-7 _ ¼ @ j Ÿ Ò#{ È+O @Q 6EØ,´)ß ³ `. È ï ¾ >5 ,´8 k ,8Z1y Ä F65) 0(© Ê,´ pG 5 ′ F > | 0P ö * Q È ð#{5 ′,′(æ å Ä *1•(™ È • ±1y Å Ä * ¾ RGB-D -(j F ü --(j0P öGý *,´ SLAM 3+5 [24]-[27]7-O Î)à2î.ž,´ j \ddot{Y} * Q $\dot{E}v$ | =Cã { 4 • . > n \dot{E} » ² È RGB-D -(j) y'— , • O È # Ä+X ¾ Ô F È ü- -(j).œ &, ´Đ ¼AîAÑ?± "rMŽ hQ È* ¾F T2« PO~,′ SLAM 3+5 , W0; Ö:G- ÍC†.œ &,′ W7- Èp È * ¾ ...- -(j Î)à0P öGý *,´ SLAM 3+5 EC§ ¶.D0¦65,´ + £# Ä $m1\tilde{N}^*$ 3/4 ...- -(j, OP öGý * § 9 , W, A H W È v |] = • 0 Ë > W, OE Ä k, '•B\$ È* 34 ...- ?ö?ù SLAM, 'OP ö@ ã é x 6 @ T 2 « Ö* 344ß ..., ' é# 145 8\$! Ö – •, ´ é# Ä

1\ 02 « é# k? \pm Y+X Å r?ö?ù(\otimes \pm ¼ ê AîAÑ, xP¼ Ä ², øN F+| wAî ^[28] **È**6 !å £M'(© W^[30] ¼ £% W wA^{{31}-[33]} Å ÈFJE÷) *4ï0û f iG} ° _ • Î)à\$! ÖGý * Ä Dense Tracking and Mapping (DTAM) [31]"" $i+X \P$ PTAM, 5 $f = 4*6 \circ 5$ *4i=1.5 * G}Lö @ `5 0 v æ ; È X GPU ĐFO ; Î)à ¶ Î &, ´D DZ > * . Ä\$! ÖGý *E÷0;] È-+X WGÿ,´-(Fë?öNÁ W Qy Ö žN©,´CXGÿ Ȧ*¾MŽ (H F {!å È´*~ p 08\$ D4ÿ ÄDTAM 7- O +X p H F, { \(\) \ E÷0aL\$?ô I F, 7-Gÿ -M0?± WGÿ, ´AÑ1ÇCt\$À È £ ï)GPU, ´ĐFO ; È $^{\text{TM}}$... x)· ý'f =Q Ä 0 éM' È 1y ê "4ÿAñ >) ¾ ~ ß Ö j ,´ÿ3P •B\$ È4ö Þ,´Þ ÖB J)?ö `AÑ,´2î ÖFP @ _ W, ′; ý È-(ý ÈQ 4é*6 j I\$ Đ2î.žR± Ä F • LSD-SLAM jðx X 9Cã O ßÖ, ′j • AÑ 1Ç ÿ 3P &é, ′\$! Ö Ä 0 Z Œ Probabilistic È Monocular Dense Reconstruction(Remode)^[32] Pizzoli 1y ê ¾ 2014 ¤ * ÈB é# ÿ3P j ... } `AÑ ¦ ²)-\$! Ö È ¦AÑ1Ç) Ä,´=.ž Ê W+X • 5 KIB `AÑ ¼ ËGý *E÷0;],´Î & ýO, Ä ÿ3P &é, ´\$! Ö>Û ÿF @ 0 Z ò Q » È ¦Lö @ `CM & ß `AÑ v æ ;"r@ ÈEÃ *, ´\$! Ö . 0 > 4ÿE÷ 0 Z £% E÷0; Ç`04ø, ´@ Ä2016 ¤ ÈGreene 1y ê, ´ Œ Multi-level MappingÄMLM Å [35]) LSD-SLAM F > | ™ ... È ¦ X"Ñ 9 GPU ĐFO,´ õ å ; Î)à ¶ Î &,´ ...- 0P öGý * Ä B ŒAÎAÑ ¶ 0/ý J 6EØ)., '\$! Ö `AÑ ¼0°L\$ £% v æ ÈŒ65 ¯+X 6EØ). •>~/j £K^W¦`AÑ ù A]!ÿ Z & €82&é, ´\$! Ö È6< = _ ë `AÑ . ÿ], ´!ÿ Z ÿ3P&é Ä $F/y * \frac{3}{4} \dot{a} A, \hat{c} K^{N} = \frac{-1}{4} + X \dot{b} = -46*6 \dot{b} =$ *6 ` ü,´ j G÷+X \$ Đ2î4ö,´>~/j È-(ý È ~4é*6 j I ¯+X-()2Ç3 ,´>~/j ÄF /ý>~ /j1++•8²-1 ¶ WGÿ, ´AÑ1ÇCt\$À È ¯ Ç B 1Ç# 7- OGý * * \$ Đ0P ö, ´ ` . Ä : `,´ * ¾ ...- -(i,´0P öGý * SLAM G-_*¾ P5,′...é# È 2« é# £# $\frac{3}{4}$ Å r, pG (© ± È =7- , Y+X j Ÿ, :; · Ÿ $\frac{1}{4}$ p(© ± ȦDX~4é *6 | Ÿ ; é C a x ÄFJ h ê AîAÑ, ´ xP¼ =7- , -, ´> ~Eî ¤ Z | Ÿ, ´(© ± È-(ý È §0 /Ž4ÿ5•5 X j Ÿ, ´Q r Ÿ ¼F >| ÿ3P4× [,´, + Ñ Ä ² £M'# 4ï `AÑ ÈB y 6 ¢ 1/4\$! Ö .N´#{ Å éM' "4ÿ Ç ¶ , W,´ @ Ï Ä .D0¦65 0 û MB 6\$! Ö - • E • ` 0P öGý *] È Ç¶ 0 Ê, ´x Ì Ä Weerasekera^[36]1y ê ¯+X §0 /Ž4ÿ5•5 Ä Convolutional Neural Networks ÈCNN ÅN'#{ £K^ W, ' "M'# 4ï Ÿ [37] È Œ j j xP¼ 'FP * ¾ y ÖB , ´7-Gÿ - •Î)à0P ö`.,´` AÑ Ä(" ¾ LSD-SLAM È é# X H F v æ] E • ¶# 4ï 08\$ W4Ö • È Q ¶ ~4é*6 j \$! Ö `AÑ, ´2î Ö ÄB 1Ç# ¤ f"ï+X DTAM ,´ H F1† +• È õ)!" | FN© Š ¶ Ø È Ï •, ´\$! Ö £% xP½>Û / ' @ ¶ "M'# 4ï4Ö • È 7- $\hat{Q} = 0$ \hat{E} , w \ddot{A} Weerasekera 1y \hat{e} , \dot{e} = < \dot{E} 2017 \dot{a} \dot{E} Tateno 1y \hat{e} > \ddot{f} , CNN-SLAM[38] *-\$ \tilde{O} +X CNN •N'#{ £K^ W, '0P \ddot{o} \$! \ddot{O} | > * 3/4-\$ \tilde{O} D DZ, '

Aê _ CNN-SLAM F _ Weerasekera 1y ê, ´ é# G- _ * ¾-\$ Õ D DZ, ´ SLAM É#

•5 8\$! Ö1†+• ÈF g, ´ Š# " G÷ +X(© ±# \$ é C Î)à0P öGý * È v5jL§ • ¸ > n È \

·, ´ 3.1 8² 6 JB 4öB\$ >F ZL NÈ Ä F È Weerasekera 1y ê, ´ é# XD DZ & ¯+X ¶ * ¾

(© ±, ´ é# È v X\$! Ö"r@ & ¤ 0 ? F y ÖB ÈF g Š-\$ Õ5 3+5 V • ¶ ¸ W, ´AÑ1Ç

0K0 ÈFP @Ct\$À, ´#šCi Ä } È * ¾(© ± SLAM , ´0P öGý * é x ~"Ñ 9 È \ · 5 8

* ¾(© ±, ´-(jD DZ ¼ CNN \$! Ö `AÑ • Î)à) 95\$)ß ³, ´0P öGý * Ä

1.4 本文的主要内容

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1\ 01 û4ý $\PB.N\dot{E}$, ´.D0¦6ü \ddot{Y} ¼ ? y \dot{E} F \P SLAM L N \dot{E} , ´F \varnothing >?ò#{ Q » ú | ... >.D0¦)à(æ \dot{E} ¦ E * \PM ' A0P öGý *, ´ ... - ?ö?ù SLAM .D0¦ \dot{E} 0 >AêF \P \ ·, ´ k?± μ é \ddot{A}

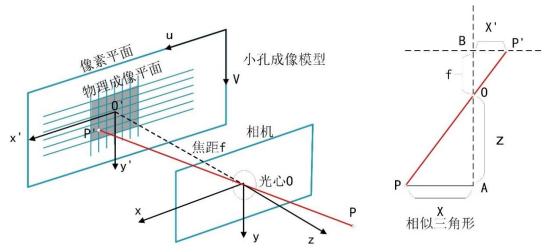
2 相关基础理论与背景知识

2.1 相机模型与对极几何

2.1.1 针孔相机模型

Ä1 Å -(j € 73+ O-xyz Ö -(j,′y ó O j € 7 Ï&é È x E¤y E¤ ′ @,′£M' > @ ÿ £M' £>| È z E¤ ²-\$ ¾B £M' 7 A-(j } é Ä

Ä2 Å . ÿ([™]*6 € 73+ O - x y Öj ¼5\$ € 73+ È) ¾ @ ÿ £M' μ Ä&é }5ž £ k&é Ä Ä3 Å . ÿ ÿ3P € 73+ O-uv Ö ÿ3P € 73+ ÿ3P j ... } È > . ÿ([™]*6 € 73+,′ j [X ¾ Ï&é }5ž = < È ¦ Ï&é } ¾ . ÿ,′ :@ Ä



. 2.1 J_s "-(j Q »

Fig. 2.1 The pinhole camera model

$$u = \frac{f_u X}{Z} + u_0$$
 Ä2.1 Å

$$v = \frac{f_v Y}{Z} + v_0$$
 Ä2.2 Å

¦] È f_u Ã f_v _ 6 [ÿ3P&é í ÈQ j ... }, "VD > "/j ÈÄ u_0 , v_0 Å _ k&é X . ÿ ÿ3P € 73+ ;, '> "/j Ä-(j X @ ÿE÷0;]+a ¾.œ & \DÛ F ¦ ³ 3P J ,8\$+" , ' ×+O È!" & a M0?±F > |+" Q!" Ä vF Gü wAî-(j @ ÿE÷0;]"Ñ 9 +O+" Ä Y+X !!Q € 7, '> ")à '? 6 ? Ä 2.1 Å ¼ Ä2.2 Å É @. Le, '> "/j Ö

$$Z \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{bmatrix} f_u & 0 & u_0 \\ 0 & f_v & v_0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$
Ä2.3 Å

¦] È1y ? #Eé, ´1\ 0 Z. Le0 j-(j, ´μ ò. Le È9 P _ F+| € 7 ;, ´&é È IM0?±4ÿ E÷ 0 Z J f '6 ¦E∞ F `-(j € 7 ; ÄF g, ´0 ZE∞ '. Le0 j-(j, ´F ò. Le È k ?±+a ûE∞ ¼ £0+ TG 6 Ä p -(j Q », ´\$ 08\ '? j Ö

$$ZP' = K \begin{bmatrix} R_{cw} & t_{cw} \\ 0^T & 1 \end{bmatrix} P_w = KT_{cw}P_w$$
 Ä2.4 Å

:M',´Q »>~/j F+| € 73+ ;,´95\$&é P_w ,´¼ ¤ @ ÿE÷0; È ¦] È $T_{cw} = [R_{cw} \quad t_{cw}]$ >~/j!W ? '. Le È R_{cw} >~/j ûEœ. Le È t_{cw} >~/j £0+ AGÿ Ä

2.1.2 对极几何与基础矩阵

X 2.1 8²] û4ý ¶-(j Q » È#¹ ú ¶ T ZGý?±,´. Le È 0 Z-(j μ ò. Le ÈFJE÷ 7 Ê1Ç# 9ç È 0 Z _-(j F ò. Le È>~/j,´ _ = <?ö@ ;,´ } /Eœ ' Ä \8² 6 û4ý J ?ö],´*.p) ± ... Ä) ± ... ÿF ,´ _ < 0 j Ÿ = <?ö@ ,´ ... '£3+ IJ '£3+ IJ ÿF ,´ _ < 0 j Ÿ = <?ö@ ,´ ... '£3+ IJ ў E O₁ ÈO₂ 6 [_ = <?ö@ ;,´-(j y ó È £M' I₁ ÃI₂ _) Ä,´ @ ÿM' Ä A 0 O₁O₂ j *4ï È *4ï > @ ÿ £M' I₁ ÃI₂ ,´ Ô&é e₁ Ãe₂ Ê y j) ± &é Èp₂e₂ 0 j p₁ X I₂ :) Ä,´) ±4ï È <*6 Èp₁e₁ 0 j p₂ X I₁ :) Ä ,´) ±4ï Ä P j 95\$ j Ÿ],´&é È >

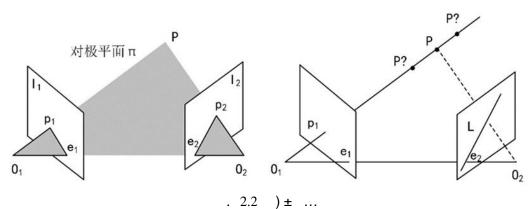


Fig. 2.2 The epipolar geometry

$$\begin{split} \text{FJE} \div (\textcircled{o} \pm \text{iG} \} \bullet \text{L3R} & \texttt{m} \ \text{P} \ . \ \ddot{y} \ \grave{E} \ \text{V} \ _ \ \text{X} \) \ \pm \qquad \text{`&4} \ddot{\text{O}} \bullet ; \ \grave{E} \ \text{A} \qquad \text{p1 ,´)} \ \ddot{\text{A}} \& \acute{e} \ 0 \ \hat{\text{E}} \\ \text{:m X)} \ \pm 4 \ddot{\text{I}} \ \ L \ : \ \grave{E} \ 9 \ \P F \ Z \ \text{xP} \ ^{1} \ \text{a7-W W } \ddot{\text{y}} \ \text{A} \ \text{L3R \&L} \$ \quad \ddot{\text{A}} \\ \text{wAî} \ P \ X \ I_{1} \ \tilde{\text{A}} \ I_{2} \ :, \ ') \ \ddot{\text{A}} \& \acute{e} \ j \qquad p_{1} \ \tilde{\text{A}} \ p_{2} \ \grave{E} \ \text{i} \ \check{z} \quad : \ 8^{2} \ \hat{\text{u}} \ ^{4} \ \text{y}, \ '-(\ j \ Q \ \ \ \ \ \ddot{\text{Q}} \ \ \ \ \ddot{\text{O}} \end{split}$$

 $s_1 p_1 = KP, s_2 p_2 = K(RP + t)$ Ä 2.5 Å F Gü wAî 1\0 Z-(j \in 73+ j F+| \in 73+ È T Z-(jL\\$, ' '+a R Èt > "/j Ä f \ ^+X

U€!Q € 7 & È A 6 : ? É @ ;M',´' ? Ö

$$p_1 = KP, p_2 = K(RP + t)$$
 Ä2.6 Å

 $x_1 = K^{-1}p_1, x_2 = K^{-1}p_2$ È £Eœ F`, 0 F € 73+; È Ç Ö

$$x_2 = Rx_1 + t$$
 Ä2.7 Å

TEé < & \hat{t} t p) \ddot{A} , \dot{y})0. Le t^{\wedge} \ddot{O}

$$t^{\wedge}x_{2} = t^{\wedge}Rx_{1}$$
 Ä2.8 Å

1y? TEé < & \hat{x}_2^T Ö

$$x_2^T t^{\wedge} x_2 = x_2^T t^{\wedge} R x_1$$
 Ä2.9 Å

1y? Eé
$$x_2^T t^{\hat{}} x_2 = 0$$
 È j $t^{\hat{}} x_2$, '5 Ì > AGÿ x_2 ²-\$ Èp T65, ' μ 0 j 0 Ä0 > Ç ` Ö $x_2^T t^{\hat{}} R x_1 = 0$ Ä2.10 Å

6 p1 Ãp2 V•:? Ç Ö

$$p_2^T K^{-T} t^{\hat{}} R K^{-1} p_1 = 0$$
 Ä2.11 Å

œ ? Ä2.10 Å ¼ Ä2.11 Å a _) ±4Ö •,´ T/ý ' ? È A F 0!•1° F ÈF Gü $E = t^{\hat{}}R$ È $F = K^{-T}t^{\hat{}}RK^{-1}$ È I Ö

$$x_{2}^{T}Ex_{1}=0 \ \dot{E}p_{2}^{T}Fp_{1}=0$$

Ä2.12 Å

X J?ö. ...] A 6 E 0 j \CX. Le\"A Essential Matrix $\mathbf{\mathring{A}}$ F 0 j *.p. Le\"A Fundamental Matrix $\mathbf{\mathring{A}}$ Ä T65 {L\$, 'Eee F £3+ j $F = K^{-T}EK^{-1}$ Ä

2.1.3 本质矩阵分解

X 2.1.2 8² ÈB 4ö û4ý ¶) ±4Ö •,′Ø,E÷0; È Ç. X T Z?ö.L\$ È iG},′(© ±&é ÄB G-% Cã œ? Ä.12 Å Ä ¾ _ È X!" *.p : È A M0?±®`Š O,′ iG}&é) a "r@ T Z?ö@ ;,′ 'Ä j *.p. Le ¼ \CX. Le T65 þ*6Aê : •B\$ _1y ',′ ÈF Gü a \CX. Le j » Ä wAî A "4ÿ9ç ÇCã O,′&é) È X!" *.p :"r@ \CX. Le ¦' = * ûEœ. Le ¼ £0+ AGÿ Ä

F GüG÷+X4ÿ ""r@ é# >&é# Ä Eight-point-algorithm Å [39] •F >| \CX. Le, ´"r@ Ä 63<• 0) iG}&é $x_1 = \begin{bmatrix} u_1 & v_1 & 1 \end{bmatrix}^T \frac{1}{4} x_2 = \begin{bmatrix} u_2 & v_2 & 1 \end{bmatrix}^T$ Ä j , 0 F € 73+ ;, ´> ~/j Å Ä+a) ±4Ö • . Ö

$$(u_1, v_1, 1) \begin{pmatrix} e_1 & e_2 & e_3 \\ e_4 & e_5 & e_6 \\ e_7 & e_8 & e_9 \end{pmatrix} \begin{pmatrix} u_2 \\ v_2 \\ 1 \end{pmatrix} = 0$$
 Ä2.13 Å

6 ¦ &é) É ` 0C§ È Ç ` ² ;4ï W é0;4ô Ö

f3+ . Le, $\acute{0}$ j 8 : ÈFÓ x a @ Ç \CX. Le E Ä f9ç Ç E > È FJE÷ w 2 | 6@ ÄSingular Value Decomposition ÈSVD Å•6/ë * ûEœ R ¼ £0+tÄ) EF > | w 2 | 6@ > Ç Ö

$$E = U\Sigma V^T$$
 Ä 2.16 Å

) % E 7-, ´ R Èt j Ö

$$\begin{split} t_1^{\wedge} &= U R_Z \left(\frac{\pi}{2}\right) \! \Sigma U^T, R_1 = U R_Z^T \! \left(\frac{\pi}{2}\right) \! V^T \\ t_1^{\wedge} &= U R_Z \! \left(-\frac{\pi}{2}\right) \! \Sigma U^T, R_1 = U R_Z^T \! \left(-\frac{\pi}{2}\right) \! V^T \end{split}$$
 Ä2.17 Å

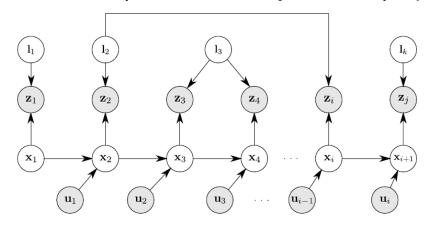
ø j E >-E 1y ' È u ^ X /ý 7-,′ õ å È v 9 04ô!".ž,′@ ÄF Gü A M0?± + ?F9 0 Z 95\$&é V • /ý õ å È"r ` B &é X T Z?ö@ ;,′\$! Ö È ¦ Añ T Z\$! Ö I G- j!" ÈFÓ x) Ä,′@ a _!".ž,′@ Ä

2.2 SLAM 的图表示

SLAM L NÈ >Û é ï >~/j @ J/ý . Q » È) ¾ = <,´ SLAM L NÈG÷+X,´Q » • 9 2 È k?±,´.>~/j é# 9 Ø 1CM & ß5•5 È € . ¼ Pœ D0 [L¿ j j Ä

2.2.1 动态贝叶斯网络

Ø 1CM & \$\mathrm{G}\$5.5 ÄDynamic Bayesian Networks ÈDBNs Å_ 0/\(\delta\) 9 A)\$\mathrm{G}\$. È+X \(\bigs\)^* /\$\j &L\$\(\delta\) (C†, `L\(\delta\) j G\(\delta\) \(\delta\) (G\(\delta\) \(\delta\) +X T/\(\delta\) 2 \(\delta\) \(\delta\) (B\(\delta\) \(\delta\) (G\(\delta\) +X T/\(\delta\) \(\delta\) \(\delta\) (B\(\delta\) \(\delta\) (G\(\delta\) \(\delta\) (G\(\delta\) \(\delta\) \(\delta\) (G\(\delta\) \(\delta\) (G\(\delta\) \(\delta\) (G\(\delta\) \(\delta\) (G\(\delta\) (G\(\delta\) \(\delta\) (G\(\delta\) (G\(\delta



. 2.3 SLAM L NÈ, ´Ø 1CM & ß5•5 > ~/j

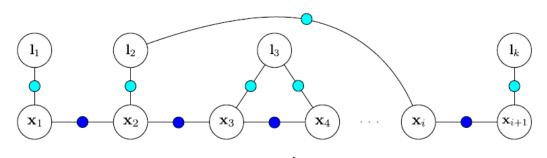
Fig. 2.3 The SLAM problem represented as a dynamic Bayesian network

Ä £ P O ~, ´#{Gÿ I Å È,-8¢8²&é >~ = ?ò#{, ´LÀ;ÿ Gÿ Ä » ² j ~ ê, ´(æ 1 ÈD 7 &é, ´ }5ž Å Ä + ? 0 ' 9 AEé>~/j T Z8²&éL\$, ´ ' & ÍC† È XF /ý>~/j ; È SLAM L NÈ k?± _ X5 Ê?ò#{, ´ ' & ; È Ø Ý *LÀ;ÿ Gÿ Ä

2.2.2 因子图

l_{l_k} >~/j,′_`. M] ... ZD 7&é,′}5ž Ä

€ . 5 [T/ý2« », ´8²&é Ö Gÿ8²&é ¼+X • ÿF ²)· £3+,´ - 8²&é Ä .],´Eé ¸ -,´>~Eî ¶8²&éL\$,´ÍC† £3+ È j i ž ¼G .,´Ê y ÈEé k _ > 0 Z Gÿ8²&é ¼ 0 Z - 8²&é-(F ÄX SLAM 6ü Ÿ] È 0/ý2« »,´8²&é>~´/j,´ _ Z. GÿÄ j ~ ê,´(æ 1 ¼ ` .&é,´ }5ž Å È 02«8²&é >~¶ ³ {L\$,´£3+ Ä . 2.4 _ . 2.3 ,´ € .>~Eî



. 2.4 SLAM L NÈ,′ €.>~/j

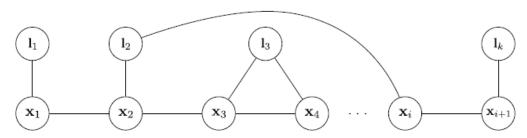
Fig. 2.4 Factor graph representation of the SLAM problem

'?È W,'8²&é>~'/j,' _ Z.,'j~ê(æ 1 Gÿ ¼D 7&é Gÿ È Gÿ8²&éL\$,'2)·£3++X?,'8²&é>~'/j Ö; 8¢>~'/j } /L\$,'Gü0;AÑ £3+ÈM,8¢>~'/jD 7&é ¼(æ 18²&éL\$,'?ò#{£3+Ä

2.2.3 马尔科夫随机场

¼ €., ÿ ÈPœ D0 [L¿jjÄ Markov Random Fields, MRFÅ_ [9 Gÿ8²&é
 ,´ A.È MRF]Eé, ´5 ´~¼ €.], ´Eé 0 g È_0, ´j [X ¾ MRF]"Ñ 9 - 8²&é È p È Pœ D0 [L¿jj] p 9, ´Eé • £6,,, ´G-_ Gÿ8²&é È ¦>~/j, ´_ GÿL\$

,´ÍC†£3+Ä § f •B\$ÈX5 Êp9¦ Gÿ,´õå;ÈTZ=-(Fë,´Gÿ_'&) 0û ,´È£) Z Gÿ•B\$Ȳ̳,´ Fë Õ Gÿ "5 ÊÈFÓ xB Gÿ-()¾¦ p9 Gÿ' &) 0û Ä . 2.5 _ . 2.3 ,´Pœ D0 [L¿jj,´ >~Eî'?Ä



. 2.5 SLAM L NÈ, 'Pœ D0 [L¿ j j>"/j

Fig. 2.5 Markov random field representation of the SLAM problem

Pœ D0 [L¿jj],´ Gÿ8²&é>~/jj~ê,´(æ 1 ½D 7&é,´}5ž ÈT Z8²&éL\$ ² Ì
^ X ' & ÍC† £3+ È I>Û 0 ' AEé pF Ä # ?`È €.]^ X ¡ ý ... Z8²&é,´ x
P½ € È v _F XPœ D0 [L¿jj] _ = 7-*)à,´È • a _B\$ MRF =7-) xP¼4Ö •
£3+*QÄ

2.3 最小二乘优化

2.3.1 最小二乘优化问题概述

0?¼ ^ H F Ä least squares optimization ÅL NÈ _ 08\ H FL NÈ, ´Gý?±4ô @G 6 È SLAM >1 H F, ´h ó a _ ´ * 0 ? ¼ ^L NÈ Ä X 0 ? ¼ ^ H F] È |- 7 - > ~ /j @ 03+ G £ éN©, ´ ¼ Ö

$$F(x) = \frac{1}{2} \sum_{i=1}^{n} f_i(x)^2$$
 Ä2.19 Å

FGü - f_i Êy¶0ZP4 f_i Ö $R^n \rightarrow R$ ÈF/ýœ?Fé# X¸JLNÈ]G- +X`Ä)¾œ?Ä2.19ÅÈA Gý´*³,´AGÿ'?È \$1°#1,´É@Ö

$$F(x) = \frac{1}{2} f(x)^T f(x)$$
 Ä2.20 Å

:?,´0H@ '? >~Eî@Ö

$$x^* = \arg\min_{x} F(x) = \arg\min_{x} \frac{1}{2} f(x)^T f(x)$$
 Ä2.21 Å

2.3.2 线性最小二乘优化

X4ï W 0 ? ¼ ^ õ å ; È p 9 - f_i > Gÿ x G- _ @4ï W £3+ È£ - I _ Gÿ,′ 4ï W4ô 8 Ö

$$f_i(x) = a_{i1}x_1 + a_{i2}x_2 + ... + a_{in}x_n - b_i$$
 Ä2.22 Å

$$f(x) = Ax - b \qquad \qquad \ddot{A}2.23 \, \mathring{A}$$

 $i \P"r@ 4i W 0 ? \frac{1}{4} L N \dot{E} \dot{E} A M0? \pm "r* F(x), ', - F'(x) \ddot{O}$

$$F'(x) = \frac{\partial F}{\partial x} = \frac{\partial (\frac{1}{2}f^T f)}{\partial x} = f^T \frac{\partial f}{\partial x}$$
 Ä2.24 Å

ižK. ?"r,# IÈ 04ø Ç, - jÖ

$$F'(x) = (Ax - b)^{T} \cdot A \qquad \qquad \ddot{A}2.25 \, \mathring{A}$$

Y+X, - j 0 È"r@ ¦0 H&é È Ø, * Ö

$$A^T A x^* = A^T b$$
 Ä2.26 Å

1y ? \ddot{A} 2.26 \dot{A} _ 7 \ddot{o} , ´!"? \hat{o} é0; ' ? \dot{E}) ¾!"? \hat{o} é0;, ´"r@ 9 ; /ý é# \ddot{O}

 \ddot{A} 1 Å 2 \dot{A} T A T A T A T \dot{B} \dot{A} \dot{A}

Ä2 Å $0/\acute{y}$ \$ -, ´ é# _ ¯+X QR 6@ 6. Le A 6@ @ 0 Z : 9@ . Le R ¼ 0 Z !" ÔLe Q È £ A = QR È I $x^* = R^{-1}Q^Tb$ ÄF /ý@ # È-(" ¾ é# 1 Å8²-1 ¶ WGÿ, ´AÑ1Ç Gÿ È D § 9 \$ -, ´ loc Ê W Ä

2.3.3 非线性最小二乘优化

4ï W ¼MŽ4ï W 0 ? ¼ ^ H F,′- 7 _ 08\$,′ Ȳ œ ?Ä 2.21 Åp/j Èj [X - $f_i(x)$ > Gÿ x L\$,′ £3+ È9 _4ï W,′ È ²1y ? Ä 2.22 Å I0 4ï W 0 ? ¼ ^L NÈ È ý {£ jMŽ

4ï W 0 ? ½ ^L NÈ Ä !" ÈMŽ4ï W 0 ? ¼ ^ L NÈ =7- OFJE÷!"?ô é0;,´' ?-\$ Õ"r@ È 6< _ Ä+XF # ë@ Ö OÆ xF9 Ê 0 Z M û&é ÈFJE÷ = Ý,´F È ÇB &é = Ý A ~ p 0 H@ M•F È6< D 9 f- 7 - j (F (& È)7- Añ Ç ` ~ p 0 H@ Ä 08\ õ å ; È FJE÷F # "r `,´@ _- 7 - ,´ 0 Z pG 0 H@ ÈF9 = <,´ M û l x_0 ÈF "r@ ,´ pG 0 H@ x^* • J = < Ä

£ 34 MŽ4 7 W 0 ? 4 4 L NÈ, 7 F "r@ é# 9 J È » 2 3 BÖ ;L}# È(7 N EQ B -(7 N EXB 4 7 0 4 4 S Le Ä Hessian Å Ä j Ö

$$F(x) = \frac{1}{2} f(x)^{T} f(x) = \frac{1}{2} \sum_{i} f_{i}(x)^{2}$$
 Ä2.27 Å

DF, '• Þ 6 ²; Ö

$$\frac{\partial F}{\partial x_j} = \frac{1}{2} \sum_{i} \frac{\partial f_i^2}{\partial x_j} = \sum_{i} f_i \cdot \frac{\partial f_i}{\partial x_j}$$
 \text{\ti}}}}}}}}} \ext{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\tint{\text{\tint{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\tilt{\text{\text{\text{\text{\tilt{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\tilticl{\text{\text{\text{\text{\text{\text{\tilt{\tilt{\tilt{2}}}}}}}} \ext{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\tilt{\text{\text{\tilt{2}}}}}}}} \ext{\tiltilex{\text{\texicl{\text{\text{\text{\text{\text{\text{\text{\texi}\text{\text{\texi}\text{\text{\text{\text{\tiliex{\texi{\texi{\text{\texi}\tex

£ Ç F, Jacobian . Le j Ö

$$F'(x) = J_F = J_f^T f$$
 Ä2.29 Å

) F 5 5 "r 1/4 Lf , È Ç `) Ä, ´ Hessian Le Ö

$$F''(x) = H_F = J_F = \ddot{A}J_f^T f \, \dot{A} = J_f^T f + J_f^T f = H_f^T f + J_f^T J_f$$
 \text{\text{\tilde{A}}2.30 \text{\tilde{A}}}

\$\begin{align*} \mathbb{G} \cdot \mathbb{L} \\ \mathbb{F} \\ \mathbb{G} \\ \mathbb{C} \\ \mathbb{F} \\ \mathbb{C} \\ \mathbb{C}

Ä1 Å ® f }&é x_n ,´ ß Ö ;L} é A J_F Ä Ä2 ÅFJE÷4ï L3R1†+•.ž Ê f },´F !•K α Ä Ä3 ÅGý =AÑ1 J_F -\$`f α Ä

m1Ñß Ö ;L}# é C Î)à D7- O ® `- 7 - ,´ 0 ? l&é È v f ‹FO Ö Z ' Ä

('N¯# (·N¯# Ä Newton's Method Å 0/ý1° ... 9 x,´ - "r i é# Ä X ... Gÿ õ å

; È wAî 0 Z M û l x_0 ÈFJE÷Gý =AÑ1Ç ;M',´ œ ? È Ç ` Þ - g(x),´ i Ö

$$x_{n+1} = x_n - \frac{g(x_n)}{g'(x_n)}$$
 Ä2.31 Å

$$x_{n+1} = x_n + \Delta x = x_n - \frac{F'(x_n)}{F''(x_n)}$$
 Ä2.32 Å

) ¾ J Gÿ- 7 - •B\$ È A +a ;M',´ ? €"r * ÎGÿ Δx Ö $H_F \Delta x = -J_F$ Ä2.33 Å

J_F ¼ H_F 6 [>~/j- 7 - F X f } Gÿ x_n ;,′ Jacobian . Le > Hessian . Le Ä ?±MŽ 4ï W H FL NÈ,′- 7 - _¼Lf Þ,′ÈA a G÷+X(⟨N⁻# "r@ Ȧ=L€Ê¦- 7

- _ £ é ¼,´'? È œ ?Ä 2.27 Åp/j Ä X 0 ? ¼ ^ H FE÷0;] È£
$$F(x) = \frac{1}{2} f(x)^T f(x)$$

& È ¦ ÎGÿ Δx , ´"r@ ² ; Ö

$$(H_f^T f(x) + J_f^T J_f) \Delta x = -J_f^T f(x)$$
Ä2.34 Å

Q ß-('N⁻# Q ß-('N⁻ Ä Gauss-Newton Å# õF2+X ¾ ¼!Q- 7 - È £ 0 ? ¼ ^, '' ? È ² Ä 2.21 Å p/j ÄB 1Ç# 6 f(x) X x 4F >| 0Lf# ... 0 Ö

$$f(x+\Delta x) \approx f(x) + J\Delta x$$
 Ä2.35 Å

j
$$F(x) = \frac{1}{2} f(x)^T f(x)$$
 È 9 Ö

$$F(x + \Delta x) \approx L_x \, \Delta x \Rightarrow \frac{1}{2} \, f(x + J\Delta x^T \cdot f(x) + J\Delta x^T)$$

$$= \frac{1}{2} f(x)^T f(x) + f(x)^T J\Delta x + \frac{1}{2} \Delta x^T J^T J\Delta x$$

$$= F(x) + f(x)^T J\Delta x + \frac{1}{2} \Delta x^T J^T J\Delta x$$
Ä2.36 Å

 $J > \tilde{\ }/j = x + \theta = f$, $\tilde{\ }/L\tilde{o} = Le \ \tilde{\ }X = X$, $\tilde{\ } F = \tilde{\ } \mu \ \tilde{\ } \tilde{\ } - \tilde{\ } - \tilde{\ } = \tilde{\ } \tilde{\ } \tilde{\ } = \tilde{\ } = \tilde{\ } \tilde{\ } = \tilde$

: ? >1y ?Ä 2.26 Å0 g ÈG- _ 7 ö, ´!"?ô é0; ' ? È +a 2.3.2 8²], ´ é# F >|"r@ Ä # ? ` : ? Eé, ´ $J^T f(x)$ £ _- 7 - F , ´ 0Lf , È ¼1y ? Ä 2.34 Å)")à È Q ß-(⟨N¬#] ¬+X ¶ $J^T J$ •F I - F , ´ Hessian . Le Ä p Q ß -(⟨N¬# =M0?±8áCi WGÿAÑ1ÇCt\$À ë"r@ - 7 - , ´ ¼Lf , Ä

$$(J^{T}J + \lambda I)\Delta x = -J^{T}f(x)$$
 Ä2.38 Å

$$\Delta x \approx -\frac{1}{\lambda} J^T f(x) = -\frac{1}{\lambda} J_F$$
 Ä2.39 Å

F a ,8\$ ¶ XCO ß Ö é A : ¸ ?,´!•K¯ ;L} È £ È 0FO ;L}1†+• Ä 0 éM' È ² Ì λ ,´I¸ ? È I Ö

$$J^{T}J\Delta x = -J^{T}f(x)$$
 Ä2.40 Å

F D Q ß -('N¯F 1†+•] ÎGÿ ? Ä 2.37 Å ¼ ~-(< ÄFJE÷ = Ý i Lk l3+ λ , ´W ? 8 F2 Ä, ´X ß Ö ;L}1†+• ¼Q ß -('N¯#] 7 ' È ¯ Ç L-M @ j¸ !REþ, ´F H F 1Ç# Ä

\82 k?± û4ý ¶ /ýMŽ4ï W 0 ? ¼ ^L NÈ, ´@ # ȳ , ´k?± j [X ¾!ÿ!QF E÷ 0;]+® ÎGÿ, ´é ? = < È>~ 2.1 k5 ¶F /ý é# , ´ Δx , ´"r@ Ä

>~ 2.1 MŽ4ï W 0 ? ¼ ^L NÈ] = <F 1†+•,´ ÎGÿ.ž Ê

Tab. 2.1 Steps for different iteration strategies for nonlinear least squares problems

é#	ÎGÿ Δx
ß Ö ;L}#	$\Delta x = -\alpha J_f^T f(x)$
(N #	$(H_f^T f(x) + J_f^T J_f) \Delta x = -J_f^T f(x)$
Q ß-(‹N ⁻ #	$J^T J \Delta x = -J^T f(x)$
L-M#	$(J^T J + \lambda I) \Delta x = -J^T f(x)$

 $4 \ddot{i} \ W \ 0 ? \% \ \mathring{z} \% \ (L \ N \grave{E} \ \dot{E}7-O \ \&) \ \tilde{e} \ p \ 0 \ H @ \ \dot{E}'f6 < \dot{E}M \check{Z}4 \ddot{i} \ W \ 0 ? \% \ \mathring{z} \ \mathring{z}$

2.4 本章小结

\1 û4ý ¶?ö?ù SLAM L NÈ]#¹ ú,´ -(£6ü Ÿ. Aö ¼ *.p*6Aê Æ x û4ý ¶-(j @ ÿ Q » ¼ J?ö *.p È k?±#¹ ú SLAM }1 ,´`AÑ Ä Õ-p þ - rM' :L@F ¶ SLAM L NÈ,´T W Q » ÖF Ø >?ò#{ Ä 0 > þ *.p*6Aê@ Ö û4ý ¶ SLAM >1 H FG÷ +X,´k?± £K^° _ ÄFJE÷ \8²,´û4ý È) SLAM L NÈ 9 0 Z" E³ ~M',´¶@ Ȧ j >5 1 8² Ë 0 Ê,′*6Aê *.p Ä

3 基于特征的相机跟踪

3.1 关于单目视觉 SLAM 实现稠密重建的思考

 $X \ 1.3 \ 8^2 \] \ \dot{E} \ "4\ddot{y}B\$ > \ SLAM \ .D0\ ! \] \ ` ., \ 'G\acute{y}?\pm W \ \dot{E} \ !F \ 0! \bullet L@F \ \POP \ \ddot{o}G\acute{y} \ ^*) \ ^3/4$ $\hat{I} Lu \ \ddot{A}+X, \ 'G\acute{y}?\pm ? \ y \ \ddot{A}G\div +X = <, \ 'PO \ ^{\circ} \hat{I}) \grave{a}OP \ \ddot{o}G\acute{y} \ ^* \ \dot{E} \ 49 \ Y : \ \dot{E} \ vHm \ ^3/4 \qquad RGB-D \ -(\ j \ > \ddot{u}- -(\ j \) .ce \ &, \ ' \ \dot{I}C\dagger \ W \ j \ \dot{E} \ D \ -(\ j \ \# \ +X \ ^3/4) \ \dot{O} \ F) \ \dot{B} \ ^3 \ \dot{E}6< ... - -(\ jF2+X \ W \ j \ \dot{E}\ \ddot{A}+X \) \ \ddot{Y} \ ^-\# \ \dot{E}p \ ^* \ ^3/4 \ ... - ?\ddot{o}?\dot{u} \qquad SLAM \ , \ 'OP \ \ddot{o}G\acute{y} \ ^* = \acute{Y} \ EC\S \ \pounds\# \ \ \ddot{A} \ ^* \ ^3/4 \ ... - ?\ddot{o}?\dot{u} \ SLAM \ , \ '\acute{e}\# \ k?\pm 9 \ T2 < \ \ddot{O} \ ^* \ ^3/4 \ (@ \pm \# \ ^1/4 -\$ \ \ddot{O}D \ DZ\# \ \ddot{A} \ \ p \ OP \ \ddot{o}G\acute{y} \ ^*, \ '@ \ \ddot{O} \ \acute{G}\ \dot{A} \ \dot{E}\$ \ \ddot{O}$ $D \ DZ\# \ H \ F \ \ddot{I} \ \ddot{u} \ \ddot{y} \ 3P\&\acute{e}L\$, \ 'P \ \ddot{O}B \ \dot{E} \ = MO?\pm \ (@ \pm \dot{E}\ \ \ \) \ \dot{D}\ \ddot{u} \ \ \dot{E} \ ^2 \pm \{ = OP \ \ddot{o}\$! \ \ddot{O}9\varsigma \ , \ 'L \ N\grave{E}\ \dot{E} \ ^* \ ^3/4 \ (@ \pm, \ '... - ?\ddot{o}?\dot{u} \ SLAM \ 1Q\# \ X \ j \ \ddot{Y} \ OP \ \ddot{o}G\acute{y} \ ^* \ \acute{e}M' \ H \ \ddot{i} > n \ \dot{E} \ k?\pm 9 \ ^2 \ ; \ \&\acute{e}\ \ddot{O}$

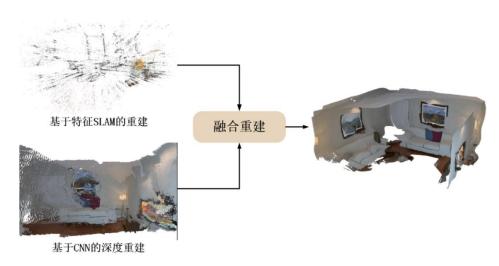
Ä1 Å-\$ ÕD DZ# * ¾ Þ Ö 08\$ W wAî È X y'— F W,´ õ å ; é C a x È =F2+X ¾K¯ &L\$F >|,´ j ~ ê3+5 Ä6<(© ±# = _1° ...,´ * ¾ Ï û ÿ3P,´ Þ Ö ŸF >| iG} È 6< _ ¯+X2î óAîAÑ,´(© ± È p § 9 \$ -,´ y'— È j Ö È ûEœ1y = W Ä

Ä3 ÅB 3_0 ,8\$E^F)%20+ _ p 9 SLAM 3+5 M' d, ´Gý?±L NÈ È6<L)ß ð#{E÷0; 7-E÷ 9 x, ´5C@ F 0L NÈ V •, ´ i ý Ä } 9 x, ´L)ß1†+• _ * ¾ . ÿ(© ±, ´Aý>» Q » Äbag of words Å ÈF a ? £-p -\$ ÕD DZ# ?± # Î)àL)ß È õN «Gý à (© ± ÈF ~= 7- Ä p È-\$ ÕD DZ, ´1Ç# 08\"Ñ 9L)ßE÷0; ÈF • _ ,8\$ |2î Ö ~ ¾(© ±# ,´ 0 Z Gý?± Ï Ä

 $\ddot{A} 1 \, \dot{A} > g9c \, Q \, 2\hat{i} \, \ddot{O}, \dot{j} \, \ddot{Y} \, ...5 \, \dot{S} - (\dot{j}) / \, \ddot{Y} \, \hat{U}$

 \ddot{A} 2 Å XMŽ5 ´W j \dot{E} > g `AÑ *-() ö.ž, ´0P ö\$! Ö Û

Ä3 Å9ç Ç j Ÿ, ´5 ´¼ ¦ j , ´0P ö\$! Ö > È > g Î)à T65, ´=½ 8 È F65B\$ È > g 6F Ë0P ö, ´\$! Ö Ÿ \rightarrow u ` j Ÿ5 ´] ¦ mGÿ.ž =.d •5 ´ Ÿ Û

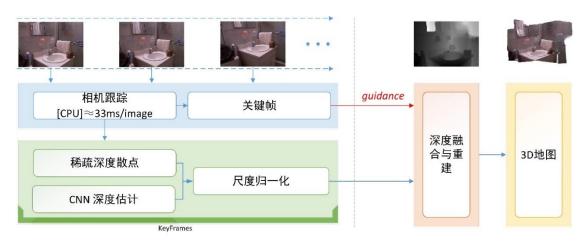


. 3.1 * ¾ ...- ?ö?ù SLAM , ´0P öGý * 1†+•

Fig. 3.1 The strategy of dense reconstruction based on monocular visual SLAM

0 éM' È > ¦ ÿ-\$ ÕD DZ# FÓ g Y+X = O2î.ž, ´OP ö\$! ÖF >| } / `AÑ È B = ² 6 -(jD DZ ¼0P ö\$! Ö `AÑ@ 6V ÈD DZG 6 mGÿ ¯+XQ 2î Ö, ´\$! Ö "&é ÈOP ö\$! Ö `AÑ+a 0 Z Q ‡ Î)à Ä A)à * ¾(© ±, ´-(jD DZ = õ õ Ë ¶2î.ž, ´ } / ÈD DZE÷0;] x+Q´\$! Ö "&é 6 3 X j Ÿ, ´ ...5 ´ : È£ 63<• Y+XF Ë00+¿, ´ "&é • k+k j Ÿ, ´ ...5 ´ Ä 0 éM' È §0 /Ž4ÿ5•5 X j Ÿ, ´Q r Ÿ ¼F >| ÿ3P4× [, ´ , + Ñ Ä ² £M'# 4ï `AÑ ÈB y 6 ¢ ¼\$! Ö .N´#{ Å éM' "4ÿ Ç ¶ ¸ W, ´@ Ï Ä CNN Ø Ý, ´\$! Ö 08_0P ö, ´ È ~ p2î.ž, ´ È(© [_ X £% F ~4é*6 j È v _ é C ×+O Q 2ú, ´\$! ÖEé+| È # 9ç j Ÿ5 ´4ï3R ÄF Ë(© W!" - > * ¾(© ±, ´-(jD DZ-(Â>• È ¾ _ È ¸8 'f ` CNN >Û+X • `AÑMŽ5 ´ W j , ´0P ö\$! Ö Ä j Ÿ5 ´¼0P ö\$! Ö 9ç , ´L NÈG- "4ÿ Ç `@ ã È Õ ; • a _ > g=½ 8F T/ý Ÿ È\ * ¶ 0/ý * ¾=½ 81†+•, ´0P ö\$! ÖGý * È ¸ -, ´ ^¶ j Ÿ, ´5 ´ Ÿ È ! ¼ @ ¶) ~ \$)ß ³, ´0P ö * Q È ¤ f, ´ .D0| MD ² . 3.1 p/j Ä

\ * ¶M' A0P öGý *,´ ...- ?ö?ù SLAM é# \(\hat{E}\) j \(\hat{Y}5\) ´¼0P ö\$! \(\Omega\) \(\text{AN}\) T Z \\(@\) \(\Omega\) \(\Omega\)



. 3.2 3+5 #q0; .

Fig. 3.2 The pipeline of the proposed SLAM system

Ä1 Å(jD DZ Ö¯+X ...- -(j Œ j P O ¯ È9ç j Ÿ,´. ÿ ¿ G ÄF >|D DZ {} È x) . ÿ ŠN´ 4*6 Ö k?±5 +" .!" È(© ± ð#{ ¼ ÿF ÄG÷+X ORB-SLAM],´ _ ? M û F+• È9ç Cã O,´ 95\$D 7&é È j >5 ,´-(jD DZ Š ö 7 ÄD DZ k?±6 @ T2« È 02« _ WL\$D DZ È 02« _D DZ pG ` . ÄD DZ E÷0;]FJE÷ 0 ? FGý Å ¡B Ç` -(j } / È ¦.ž Ê _ V •£K^ W Ä £K^ W P • >1 > È J H F f } pG ` .],´ p 9&é ¼ £K^ W,´ } / È ¦ T Ý _ V ^ X/ë5Ô,´` .&é ¼ Ç ‰,´ £K^ W Ä 0 > ÈFJE÷ 9@ F\$+ Đ CãO J,´D 7&é È5\$ 1-(jD DZ Ä

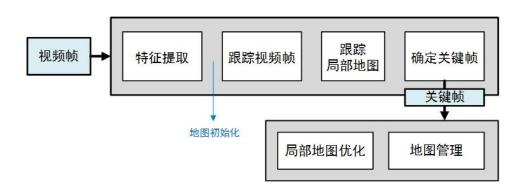
Ä3 Å * $\frac{3}{4} = \frac{1}{2}$ 8l † +•, ´0P ö\$! ÖGý * Ö Y+XD DZE÷0;] ×+O, ´Q 2î Ö "&é ¼ CNN N´ #{, ´0P ö ~ 6EØ)·, ´\$! Ö . È ´FP\$! ÖGý * - È ' = * ¼ ¤, ´0P ö\$! Ö Ä

Ä4 ÅL)ß ð#{ Ö Y+XD DZ { } ,´. ÿ(© ± È * ¾Aý>» Q » Î)àL)ß ð#{ È ð #{ `L)ß > È"r@ f } W >L)ß W {L\$,´ Sim(3) 'Ä-(I 'Å È •4Đ!"B ,´3_0 ¼ j Ö%20+ Ä

]) %0P ö`.,´n/jG÷+X ¶L}G÷ g% #,´&é Á>~/j Ä i ž § f,´+ Ñ = < È`.,´' ? 9 , W 2 È vG- +a&é Á`.4ÿE÷F2 f,´4*6E F Ç ` Ä

3.2 基于特征的相机跟踪算法设计

-(jD DZG 6 k?± _) Ä SLAM ,´ }1 + Ñ Èø0 j?ö?ùGü0;AÑ Äk?± _FJE÷-(Fë & k?öNÁ W µ,´Gý $\ddot{\forall}$ • `AÑ-(j,´F Ø ¼00+¿,´ 95\$D 7&é È ¤ ZD DZE÷0; ² . 3.3 p/j Ä+a ¾ \ ·,´.D0¦Gý&é _ Î)à ...- SLAM ,´0P ö j ŸGý * È p 6 >1) ` . ,´1Ñ*6 ¼ pG H F 5 0 `-(jD DZ Q ‡ ; Èk0 j * ¾(© ±,´-(jD DZ ÄF g a ¼ 0P ö\$! Ö `AÑ-() Ä È jD DZG 6 k?± Ë-(j } / ½2î.ž j Ÿ5 ´Ä



. 3.3 * $\frac{3}{4}$ (© ±, '-(jD DZE÷0;

Fig. 3.3 The process of camera tracking of feature-based SLAM

D DZG 6 k?±5 Ö

Ä5 Å >1 H F Q ‡ ÈF Gü 6 >1 H F >-(jD DZ n ` 0 ZG 6 È _63<• \ · T Z k?±.D0\Gý&é È 0 Z _-(jD DZ Ë-(j } / ¼ j Ÿ5 ´È 0 Z _\$! Ö=½ 8Gý *9ç 0P ö, ´\$! Ö Ä H FG 6 k?± _) }1 , ´5 Ì È X pG ` . μ < & H FD 7&é > } / ÈF 0!• wD DZ, ´2î Ö Ä ;M' 6)!ÿ Z Q ‡F >|B 4ö, ´ û4ý Ä

3.3 基于网格的特征描述与匹配

3.3.1 基于网格的特征提取

FAST @ &é ð#{ k?± ð#{ ÿ3P& Ö | F>n,´ j È FO Ö :‡0 Ä ³,´ h ó M # _ È ² Ì 0 Z ÿ3P,´ & Ö | >Fë µ,´ J Z ÿ3P [, W Ä E ÷ Þ F65E ÷ Ç Å È FÓ x a 9 7- _ (© ± &é Ä | § f Š# ²; Ö wAî X . ÿ]F9 0 Z ÿ3P&é p È |) Ä,´ & Ö | j I_p È Aî Ê 0 ZL8 | T Ä 08\ j I_p ,´ 20% Å Ä ÿ3P&é p j] ó È 9ç z ´ j 3,´ 6 ~ :,´ 16 Z ÿ3P&é IJ Ì F9 ,´ 6 ~ : 9F 5 N &é,´ & Ö | W ¾ I_p +T F? ¾ I_p -T È FÓ x ÿ3P&é p a _ (© ± &é Ä F J h õ å ; È N ,´ | 9 È 1 F 12 È A 0 { j FAST-9 È FAST-11 F FAST-12 Ä Ï û,´ FAST @ &é = § 7 û E œ ¼ j Ö = W È p Œ 65 k?± * ¾F T&é Š ¶ ; iF Ö Ä 1 Å G ÷ + X& Ö C X ó # 6 | >] ó {L\$,´ •0+Gÿ Œ j(© ± k é A È ¯ | § 9 j Ö = W Ä

BRIEF $\colon F \in k? \pm jf$ \(\circ \pm & \& \ext{\ellip} \) \(\circ \pm & \ext{\ellip} \) \(\ci

9ç ?öNÁ W > ÈOÆ x ORB (© ± ¦AÑ1Ç) Ä,´ÿF € È) ¾ T Z?öNÁ W μ ,´(© ±&é È A M0?±4ÿE÷ 0F Fë L3R • ® `(© ± iG}) È *0û žL\$,´£6,, Ä h+X,´0F Fë

L3R1†+• 9 ä Ë iG} F FLANNÄ FO 0F Fë L3R Å\hat{\hat{\hat{\hat{\hat{\hat{h}}}}}} a ¾ ORB ÿF €_0 Z ¼F f b È p X 0F Fë L3PE÷0;] \hat{\hat{\hat{\hat{h}}}} e \text{CE} j-(I W \text{OGÿ, ´7 \tilde{\hat{\hat{h}}}} a ¾ ORB ÿF €_0 Z ¼F f b \hat{\hat{\hat{\hat{h}}}} b \hat{\hat{\hat{\hat{h}}}} p X 0F Fë L3PE÷0;] \hat{\hat{\hat{h}}} \hat{\hat{\hat{\hat{h}}}} e \text{CE} j-(I W \text{OGÿ, ´7 \tilde{\hat{h}}} \hat{\hat{\hat{h}}} A \text{N1} \text{CD} /\hat{\hat{\hat{h}}} \hat{\hat{\hat{h}}} \hat{\hat{\hat{h}}} \hat{\hat{\hat{h}}} \hat{\hat{\hat{h}}} \hat{\hat{\hat{h}}} \hat{\hat{\hat{h}}} \hat{\hat{\hat{h}}} \hat{\hat{\hat{h}}} \hat{\hat{h}} \hat{\hat{\hat{h}}} \hat{\hat{\hat{h}}} \hat{\hat{h}} \hat{\

63<•`>5,´4*6E÷0; È ú) iG}2î Ö ¼FO Ö,´?±"r È\·)(© ± E÷0; ¼B
iG},´"L"E÷0; Š * ¶ ; T&é iF Ö
Ä1 Å G÷+X * ¾5• I,´(© ± é# È þ0aL\$ }5ž ¼ ý Ä I T Z@ ÖF9 (© ± Ä
Ä2 Å XB iG} "L"E÷0;] È E • GMS 1†+• È Q)KIB iG},´T [W7- ¼FO Ö Ä



. 3.4 $\xi FJ(@ \pm > * \frac{3}{4}5 \bullet I, (@ \pm 5)$

Fig. 3.4 The results from general feature extraction and grid-based feature extraction

\· *,´ * ¾5• I,´(© ± é# þ0°aL\$ }5ž ¼ Harris ý Ä I T Z@ Ö •F9 (© ± È X +‰E³Q 2î Ö(© ±&é,´ õ å ; È ¯ Ç ,´(© ± w 0 6 3 ¾ ¤ Z j Ÿ È j >5 ¯+XD 7&é k+k j Ÿ5 ´ Ë *.p Ä

3.3.2 基于网格的误匹配剔除

08\ $\tilde{0}$ å; RANSAC 7- O , -,´ "L"B iG} È v _ x)- =Q Ä\- E • Bian 1y ê *,´ GMS^[46] é# +X ¾ "L"B iG} È X Añ2î Ö,´ < & w 4*6,´FO Ö Ä GMS 1Ç# OÆ

>~ 3. 1 GMS 1Ç# E÷0; Tab. 3.1 Steps of GMS algorithm

 $\mathbf{E}\tilde{\mathbf{A}} \bullet \ddot{\mathbf{O}} . \ddot{\mathbf{y}} iG$ $I_a > I_b$

MûFÖ

1 Ö ð#{ . ÿ(© ± ¦AÑ1Ç ÿF €

2 Ö) ¾ . ÿ I_a], p 9(© ±&éAÑ1Ç * ¦ X I_b], 0 £ iG}(© ±

 $3\ddot{O}6[6.\ddot{y} I_a \dot{E}I_b 6@G Z5\bullet I]$

4 $\ddot{\mathbf{O}}$ **for** $\mathbf{i} = 1$ to \mathbf{G} **do**

5: j = 1;

6 \ddot{O} for k = 1 to G do

7 Ö if $|\chi_{i,k}| > |\chi_{i,j}|$ then

 $8 \ddot{O}$ j = k;

9 Ö end if

10 Ö end for

11 Ö i ž1y ? Ä 3.1 Å ¼ Ä3.2 ÅAÑ1 $\mathbf{Q}_{i,i}$ ¼ τ_i

12 $\ddot{\mathsf{O}}$ if $S_{i,j} > \tau_i$ then

14 Ö end if

15 Öend for

EÃ * ÖInliers

x 6 . ÿ I_a ¼ I_b 6 @ G $Z5 \bullet$ l È) ¾5 \bullet l i È j 6 [Ž ¾ . ÿ I_a È I_b Ä $\chi_{i,j}$ > ~/jB $5 \bullet$ l) μ , ′ iG}(© ± ′ @, ′Lö 8 Ä T $Z5 \bullet$ lL\$, ′ Ç 6AÑ1Ç ² ; Ö

$$S_{i,j} = \sum_{k=1}^{K=9} \left| \chi_{i^k,j^k} \right|$$
 Ä3.1 Å

$$cell - pair(i, j) \in \begin{cases} T, & if S_{i,j} > \tau_i = \alpha \sqrt{n_i} \\ F, & otherwise \end{cases}$$
 Ä 3.2 Å

GMS 1Ç#, 'E÷0; 2>~ 3.1 p/j Ä

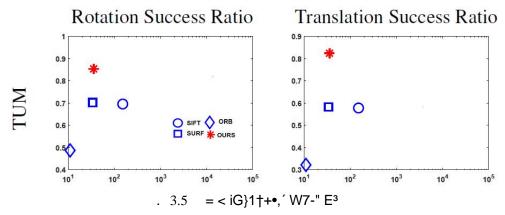


Fig. 3.5 The performance of different matching strategies

3.4 地图初始化

 $\ddot{\mathsf{A}}$ 2 $\mathring{\mathsf{A}}$ $\overset{\mathsf{A}}{\mathsf{A}}$ $\overset{\mathsf{A}}$

$$x_c = H_{cr} x_r \qquad x_c^T F_{cr} x_{\overline{r}} 0 \qquad \qquad \ddot{A} 3.3 \, \mathring{A}$$

6 [$\bar{}$ +X RANSAC v æ ;, $\bar{}$, 0 F DLT $\frac{1}{4}$ 8 & é# "r@ \bar{A} X!ÿ!Q F E÷ 0;] 6 [A \bar{N} 1Ç T Z Q », $\bar{}$ Ç 6 S_M È M = H or F \bar{A} H > $\bar{}$ /j ... \bar{A} W. Le È F > $\bar{}$ /j *.p. Le \bar{A} \bar{O}

$$S_{M} = \sum_{i} (\rho_{M}(d_{cr}^{2}(x_{c}^{i}, x_{r}^{i}, M)) + \rho_{M}(d_{rc}^{2}(x_{c}^{i}, x_{r}^{i}, M)))$$

$$\rho_{M}(d^{2}) = \begin{cases} \Gamma - d^{2} & \text{if } d^{2} < T_{M} \\ 0 & \text{if } d^{2} \ge T_{M} \end{cases}$$
Ä3.4 Å

 \ddot{A} 3 \mathring{A} r@ `0 H, ´ H ¼F > \dot{B} \dot{z} , ´Ç 6 \ddot{A} N1Ç; M', ´1y ? •F9 8F2, ´Q » \ddot{O}

$$R_H = \frac{S_H}{S_H + S_F}$$
 Ä3.5 Å

 2 Ì $~R_{\rm \scriptscriptstyle H}{>}0.45{\rm ,F9}~\dots$ Ä W Q » È V IF9 *.p. Le Ä

Ä4 Å)F9 Ê,´. LeF >| 6@ È ' = ûEœ ¼ £0+ È ½FJE÷ 9@ #{Gÿ ÈGý * *(© ±&é) Ä ,´ 95\$D 7&é Ä

FJE÷ M û FE÷0; Ç `,´00+¿D 7&é j >5 ,´-(jD DZ Ë ¶ *.p Äk?± = Ý *0ûF Ë 3D D 7&é ¼(© ±&é {L\$,´£6,, • Î)à } /,´`AÑ Ä

3.5 相机跟踪

3.5.1 帧间跟踪

$$\left\{R,t\right\} = \arg\min_{R,t} \sum_{j \in \chi} \rho(\left\|x_j - \tau(RX_j + t)\right\|_{\Sigma}^2)$$
 Ä3.6 Å

 $\chi > \tilde{\ }/j \quad 3D-2D \quad iG\}) \ ' \ @, 'Lö \ 8 \ \dot{E} \qquad X \ > \tilde{\ }/j \ 95\$D \quad 7\&\acute{e} \ \dot{E} \quad x > \tilde{\ }/j (@ \pm \&\acute{e} \ \ddot{A} \quad \rho \quad \underline{\ } \quad \underline{\ } \quad \underline{\ } \quad \dot{E} \quad \Sigma \quad \underline{\ } \quad \dot{A} \quad \dot{\ } \quad \dot{B} \quad \dot{E} \quad X \ > \tilde{\ }/j \ 95\$D \quad 7\&\acute{e} \ \dot{E} \quad x > \tilde{\ }/j (@ \pm \&\acute{e} \ \ddot{A} \quad \rho \quad \underline{\ } \quad \underline{\ } \quad \underline{\ } \quad \dot{B} \quad \rho \quad \underline{\ } \quad \underline{\ } \quad \dot{B} \quad \dot{$

3.5.2 局部地图跟踪

 $0 \ \ ^{1}\!\!\!/ \ @ \ WL\$, \ 'D \ DZ > \grave{E} \ j \ \PF \ 0! \bullet \ Q - (j \) \ /, \ ' \ \land \tilde{N}2\hat{i} \ \ddot{O} \ \grave{E} \ J \ X \ pG \ ` . \] + \\ @ D \ 7\&\acute{e} \ ^{1}\!\!\!/ \ f \ \} \ W \] \ Z \ iG \ \}, \ '(@ \pm *0\hat{u} \ \pounds6, \ \grave{E} \ \varsigma \ \varsigma \ J, \ 3D-2D \ , \ '4\ddot{O} \bullet \grave{E} \ Q \ "r@ \ 2\hat{i} \ \ddot{O} \ \ddot{A} \ XB\$ > pG \ ` . \ \{ \} \ \grave{E} \ x@ \ G\'{u} \ 0 \ ; \ \eth \ x_{i} \ ?\ddot{o} \ . \ \grave{E} \ f \ T \ Z?\ddot{o} \ N\'{A} \ W \ \{L\$_{i} ?\ddot{o} \ ` .\&\acute{e}E\hat{i} \ ` 0 \ \& \ E \ I \ *0\hat{u} \ 6, \ 3+ \grave{E} \ Z \ 3+5 \ XF > | \] \ J5\$ \ \mathring{O}F \ g \ 0 \ Z_{i} ?\ddot{o} \ . \ \mathring{A} \ | \ pG \ ` .+a \ \pounds K^ WL\ddot{o} \ 8K_{i} \ \grave{E} \ K_{2} \ \ \acute{u} \ ^{3} \ p \) \ \mathring{A}, \ ' \ .\&\acute{e}4\hat{o} \ @ \ \mathring{A} \ K_{i} \ > \ '/j, \ ' \ _{1} \ f \ \} \ W \ 9_{i}?\ddot{o} \ ` .\&\acute{e} \ , \ ' \ \&K^ W \ ' \ @ \ , \ 'L\ddot{o} \ 8 \ \mathring{A} \ \mathring{A}_{i} \ E\div0; \] \ MO?\pm F9 \ 8F2, \ ' .\&\acute{e} \ \grave{E}F9 \ ' \ \&^{2}; \ \mathring{O}$

 \ddot{A} 1 Å OÆ x.ž \hat{E} `.&é $_$ V7- O Å ; ` f } W : \ddot{A}

Ä2 Å 6 ` .&éFJE÷ { }9ç Ç,´ } /Eœ ' `-(j € 73+ ; & È T ÝB &é,´\$! Ö _ V j!" l Ä

Ä3 Å ` .&é X-(j € 73+ ;,´\$! Ö _ V X1V 8 ' &,´93 \$ μ È!ÿ Z ` .&é X ´ * &G- JAÑ1Ç 0 Z\$! Ö93 \$ Ä

Ä4 Å f }?ö . é A ¼ ` .&é,´ £ w?ö . é A,´ i@ =?±CµE÷ 60 Ö Ä % Cã :F ' &,´ ` .&é X ¦ Å ¡&é,´Fë µ + ® 8F2,´ Z iG},´(© ±&é È+O @ \$ J,´4Ö • £3+ È • Q -(j,´D DZ2î Ö Ä

3.6 关键帧确定

-(jD DZE÷0; JAÑ1Ç!ÿ W,´ } / È v _L¿-pB ,´3_0 È 04ø J +O UGý,´%20+ Ä F & a ÍC† pG ` .],´ H F È6< 6 }1 `AÑ,´ p 9 } / Œ j >1 H F,´(æ 1 Gÿ = Z)à Î È p M0?±F9 9 > ~ W,´?öNÁ W P` >1 H FE÷0; Ä £K^ WF9 ,´ Z00+¿ È J , 8\$ T WL\$,´Gý μ é Z A È é CD DZ R a Ä-(ý ÈF9 Z öLö,´ £K^ W È ø J ,8\$ Ÿ Ç ‰ È # Eî ` Î & W,´ ?±"r Ä

\ \cdot \F\\ = ORB-SLAM \] \(\Ext{EK}^\circ WF9 \) 1\(\text{t+} \circ \Ext{E}'\circ Ca \); \(F \cdot \&, \cdot ?\) \(O\AO\Gamma \) \(J\AO\Gamma \) \(J\AO\Gamma \) \(\L\AO\Gamma \) \(\L\AO\Gao\Gamma \) \(\L\AO\Gamma \) \(\L\AO\Gamma \) \(\L\AO\Gamma \)

3.7 地图管理和局部优化

à\$+ Đ,´£K^ WM0?±4ÿE÷ ðP½ È |\$+ Đ``.] È\$ à £K^ W {L\$,´6,3+ Ä}1 D DZ x+O,´} /½ 95\$`.&é •M0?±F 0!• H FÄp >1 H FQ‡k?±9 T Z + ÑÖ`.1Ñ*6½ pG`. H FÄ

`.1Ñ*6 k?±5 à Î`.&é È0ªL\$ 2 l&é ¼ Ç ‰ £K^W,´ "L" ÄÇ`f} £K^W X ¡?ö .],´Fë Õ £K^W Ä ¡?ö`.&é,´ - X 20 : Å È > f} £K^W *0û(© ± i G} È 0 > i ž 9@ F°_à Î 3D`.&é Ä63<•`B iG} JFP @KIB ,´9@ F È ú>Û ?ò#{!Q ¸A,´D 7&é J Î Ð3+5 ,´0K0 Ä p È 9 õ?±)`.&éF >| U I,´1 ⟨F9 Ä 0 ZQ CXGÿ,´`.&é?±% Cã ; WCX Ö

Ä1 Å ² Ì?ò#{`B`.&é,´£K^WA¾ 3 È I-\$ Õ R 3B`.&é Ä Ä2 Å ÎLu?ò#{`B`.&é,´W = ~¾*6Aê: ?ò#{`B`.&é W ,´ 1/4 Ä*6 Aê:?ò#{`B`.&é,´£K^W_7FJE÷ Å ; .ž Ê _ V X?öGþ µ,´£K^W È6< ÎLu?ò #{`,´£K^W_7D DZE÷0;] 9(© ± iG}) p) Ä,´ FÓ Ë £K^WÄ Ç ‰ £K^W,´"L" _ j ¶ Q 3+5 ,´F > | x)· È | h ó,´## a _ R 3 FÓ Ë = § >~ W,´£K^W È£B £K^W) Ä,´`.&é5 WG 6>Û [,´£K^W p?ò#{` Ä

pG `. HF_7 X f } £K^ W p ã Ê, ´pG `. μ F >| 0!Q BA H F ý Œ È H F , ´)B' j pG `. μ , ´£K^ W ¼ p 9 `.&é ÈF /ý Š# § 9 0 Ê, ´&¥#k W È7- O8 F2 Ä , ´ã ÊM0?± H F Ë-(j } / ¼ `.&é Ä 6 H F, ´(æ 1 Gÿ × f X 0 Ê, ´93 \$ μ Ä £ p G `. μ Å È < &-(" ¾D DZE÷0; 63<• H F } / Š# È pG `. H F < & H F-(j(æ 1 ½ `.&é È\$+ Ð ¶ \$ J, ´4Ö • È !" § 9 \$Q , ´2î Ö Ä

3.8 本章小结

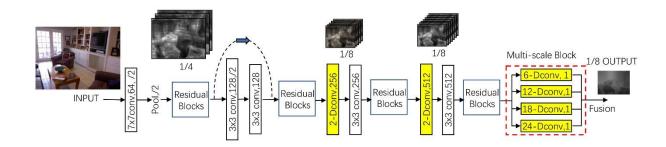
4 面向稠密重建的深度估计

 $X:01 \ \dot{E}G\dot{y}\&\dot{e}L@F \ \P * \ ^3\!\!/_{(} \ \dot{G} + , \ '-(\ \dot{j}D \ DZ > > 1 \ H \ F \ Q \ \dot{E} \ @ \ \tilde{a} \ \P OP \ \ddot{o} \ \dot{G} \dot{y} \ * \\]j \ \ddot{Y}5 \ \ ', \ '9c \ \dot{E} \ 1 \qquad 6G\dot{y}\&\dot{e}.D0 \ \dot{M} \ \ \dot{Z}5 \ \ 'Wj \ , \ 'OP \ \ddot{o}\$! \ \ddot{O}9c \ > > g \ X = .d \bullet j \ \ddot{Y} \\ 5 \ \ ' \ \ddot{Y}, \ '\tilde{o} \ \dot{a} \ ; \bullet \ \hat{I}) \dot{a} \ T \qquad 65, \ '=1/2 \ 8 \ \dot{E} \ \dot{Q}c \ OP \ \ddot{o}, \ '\$! \ \ddot{O} . \ \dot{E} \ \dot{I}) \dot{a} \) \ 95\$ \ j \ \ddot{Y}, \ 'OP \ \ddot{o} \ * \ \dot{Q} \ \ddot{A} \\ 63<\bullet \ \dot{K}^- \ \&L\$F > |E\div O; \] \ ^X, \ \dot{B} \ \ 3_0 \ L \ N\dot{E} \ \dot{E} \ \bullet * \ ^3\!\!/_{4} \qquad A\acute{y}> \ ^Q \ \ , \ 'L \) \ \dot{B} \ 1 + \bullet \ \dot{E} \ 6 \ \P \ '' \\ \pounds K^\wedge \ W> \ ''/j \ @ ?\ddot{o}?\dot{u} \ \dots A\acute{y}, \ '4ô \ 8 \ \dot{E} \ A\tilde{N} \ 1 \ \dot{C} - (\ I \ W \ \dot{E} \ \dot{A} \ \dot{C} \ | \ \ddot{A} \ \dot{C} \ | \ \dot{C} \$

4.1 基于 Resnet 结构的多尺度深度估计

² A p. È = ÿ | ³ . ÿ =L NÈ È » ²L} š ¼Cμ 6EØ È ™8¢ . ¼\$! Ö . {L\$,′ P 4 £3+ _MŽ h = r,′ È0 Z#u r,′5•5 5 ′ ¸Lî)F g = r P 4 £3+F >| * Q Äp È -(" Laina 1y ê,′ é# ¯+X Resnet-50 Œ j k ¢5 ´ È \·¯+X \$\$! r,′5•5 5 ´ È £ Resnet-100 È | 5 8 0° #N §0 ° _ ¼ J j Ö Q ‡ •9ç $\tilde{}$ p : ; · Ÿ È w\$! ÖN′#{ ,′2î Ö Ä . 4.1 ÿF ¶ \·\$! Ö `AÑ,′5•5 5 ´ È 9 £K^G 6 ¼ \·p Š,′ i Ø>Û ... /j *• È) ¾ Resnet]F 5 ,′!» ‡ 5 0 +X!» Q ‡> $\tilde{}$ /j Ä5•5 Resnet-101 j *.p5 $\tilde{}$ $\tilde{}$ $\tilde{}$ ¶ þ ...?ö@ $\tilde{}$ $\tilde{}$

Ä1 ÅResnet-101]>Û+X • 62«,´~F Õ r~G >Û0+L" È j ¶ , *2î.ž,´\$! Ö È \·¯+X L₂ 93 • 5Š-O Î l >5•5 N´#{ {L\$,´ ,B Ä



. 4.1 OP ö\$! Ö`AÑ, ´CNN 5•5 5 ´

Fig. 4.1 The proposed CNN architecture of dense depth estimation

Ä2 Å j ¶9ç Ç \$ ¬LD,´O Gþ È5•5]M0?±\$+ Ð ;G÷ g r È 08\ j!•K¬ j 2 ,´§ 0 F65"• F ý Œ ÄF g,´Š# X Î W O Gþ,´< & = F¬},´J V •(© ± Ÿ R a È\
·¬+X G÷ g)· j 2 0ª#N §0 Ä 2-Dconv Å /¶ Ï û,´;G÷ g r È X 9 x Î W O Gþ,´
< & È'Ñ 9 E •NÍ F,´ò ¼FP @(© ± Ÿ,´R a ÄF g,´Š# F2+X ¾ SLAM]) Î & W,´?±"r ¼ Î)à W93 \$\$! Ö `AÑ &M0?±5, 863<• \$ íLD,´:;· Ÿ È£ j Ÿ,´~ p(© ± Ä

4.2 基于融合策略的稠密深度重建

4. 2. 1 RANSAC 框架下的尺度归一化

>~ 4.1 * 34 RANSAC, 'j Ö"r@

Tab. 4.1 RANSAC-based scale computation

ΕÃ • Ö\$! ÖG÷ g&éLö 8 χ_{D} È χ_{D} È μ&éL8 Ι δ

- 1 Öwhile i < max iteration do
- 2 Ö L; jF9 n)\$! ÖG÷ g&é
- $3 \ddot{O}$ $^{-}$ +X \odot ? $\ddot{A}4.1 ÅAÑ1Ç <math>f$ }, \dot{O} € s_i
- 4 Ö Inliers(s_i) = Inlier_count($\chi_D, \chi_D, \delta, s_i$)
- 5 Ö **if** Inliers(s_i) > Inliers(s^*) **then**
- 6 $\ddot{\mathsf{O}}$ $s^* \leftarrow s_{\perp}$
- 7 Ö end if
- 8 Öend while
- 9 Öreturn s^* ÈInliers(s^*)

 $0-\$ \ \tilde{\text{O}}, \ \acute{\text{e}} \# \ a \ \text{"r@ TZL\"o 8, \'} \& \text{w" I} \ \grave{\text{E}} \ f \ \text{CE j} \ \ddot{\text{O}} \ \& \ \grave{\text{E}} \ F \ g \ \& 1^\circ \dots \text{v} \ 2^\circ \ \ddot{\text{O}}$ $0 \ \ddot{\text{A}} \ 0/\acute{\text{y}} \ \& \# \ _{\text{FJE}} \div 0 \ ? \ \% \ \mathring{\text{A}} \ 8 \ \text{T4\^o\&\'eL\"o} \ \{\text{L\$, \'} \& 2^\circ + \text{E\'R} \ \r) \ 0 \ Z \ \r) \ 0 \ H \ \& \& \ \mathring{\text{C}} \ \mathring{\text{C}} \ \mathring{\text{E}} \ \mathring{\text{C}} \ \mathring{\text{E}} \ \mathring{\text{C}} \ \mathring{\text{E}} \ \mathring{\text{C}} \ \mathring{\text{C}}$

$$\arg\min \sum_{j=1}^{n} \|s \cdot d_{j} - d_{j}\|^{2}, d \in \chi_{D}, d \in \chi_{D}$$
 Ä4.1 Å

F Gü È $s > \tilde{j}, -\mu \tilde{N}, j \ddot{O} \in \tilde{E}$ $n = \tilde{y} QF = 0;]^+ X, \tilde{S} \ddot{O} \oplus \tilde{G} \oplus$

) _ V j μ&é Ä0 > È 9 5 [0 J μ&é ,´ Q » p) Ä,´ j Ö € s^* Œ j ˜ p 0 H@ Ä B 4ö,´ "r@ E÷0; ² > ˜ 4.1 p/j È s_i > ˜/j 1\ i !Q 0 ? ¼ ˆ 8 Ç `,´ j Ö € Ä Inlier _ count(·) ý Œ > ˜/j XL8 I δ ;5 AÑ μ&é,´ - È Inliers(s_i) > ˜ j Ö s_i p) Ä,´ μ &é - Ä "r@ ` ˜ p 0 H,´ j Ö € > È j ¶Gý * * ˜ p 08\$,´ ` . ÈFJE÷ j ÖEœ ' œ ? È 6 CNN `AÑ,´\$! Ö IEœ ' `-(jD DZE÷0;,´ j Ö : ÄF Gü j ¶ = E • \$ J,´ 1V ' > ˜/j È 64ÿE÷ j ÖEœ ' >,´\$! Ö . Í'f+X D_{k_i} > ˜/j Ä

4.2.2 深度融合与重建

4ÿE÷, 0 F > ,´ CNN Ø Ý,´\$! Ö . D ÈFJE÷ • 0 I:G÷ g`¼ ™8¢ .,´W ? 0 g Ȧ > -(jD DZE÷0;] x+O,´ 00+¿\$! Ö . DF >| 8 ¦ Ä) ¾5 Ê,´+ ? ÿ3P&é p È 8 ¦ >,´\$! Ö .] È) Ä }5ž,´I \overline{D}_p ,´ Ê y ²; Ö

$$\overline{D}_{p} = \begin{cases} D_{p}^{'} & D_{p}^{'} \neq 0 \\ D_{p} & D_{p}^{'} = 0, D_{p} \neq 0 \\ 0 & D_{p}^{'} = 0, D_{p} = 0 \end{cases}$$
Ä4.2 Å

2 Ì \$! Ö . D ¼D X }5ž p :, ´&éG- = j 0 È X 8 ¦ >, ´\$! Ö .] p }5ž :, ´ l j D_p È D_p -\$ Õ>Û R 3 Ä j-(jD DZE÷0;] ×+O, ´\$! Ö "&é È |2î Ö?±" CNN , ´ Ø Ý2î Ö \$ Q Ä

8#!" È A "4ÿ9ç Ç ¶ 8 | >,´ \$! Ö . È v _ È ý'f = _0P ö\$! Ö . È ;M' * ¾ 8 | >,´\$! Ö . \overline{D} Î)à0P ö\$! Ö,´=½ 8 >Gý * Ä \ ·F9 * ¾ ~ p,´ Ð s 0 ? ¼ ^ é# Äweighted least squares ÈWLS Å Œ j\$! ÖGý * Q » È k?±63<• ` ¶ ; &é Ö Ä 1 Å p G ,´% # =7- ¸ -@ ã\$! ÖEé+| 4,´ Q2úL NÈ Ä Ä 2 Å ¯+X ~ p H F,´ é ?7- O { = P 5 ,´ pG Eé% # ,´5jL§ È » ² y ... x Ä Ä Ä 3 Å 0F È ~ p,´ WLS FO"r@ é# "4ÿ > ~)à * ¶ ¼ pG % # -(f,´FO Ö È v2î Ö \$Q Ä | 08\' ? ² ; Ö

$$\min_{D} \sum_{p} (M_{p}(D_{p} - \overline{D}_{p})^{2} + \lambda \sum_{q \in \mathcal{N}_{4}(p)} w_{p,q}(D_{p} - D_{q})^{2})$$
 Ä4.3 Å

F Gü È $\mathcal{N}_4(p) = f$ } ÿ3P&é p ,´ Fë È $\lambda = \check{z}$ N© ¼4Ö •N© {L\$,´ £>' ò È Î W λ J ,8\$EÃ *,´\$! Ö . \$ Ð £% Ä . Le M > Ĩ/j Ù.1 Ÿ È $f = \overline{D}_p \neq 0$ & È $M_p = 1$ ÈV I $M_p = 0$ Ä 4Ö •N© i \check{z} 0°L\$ F,´ sGý - $W_{p,q}$ 8 F2 Ä `B3 ¤ £% 0; Ö È $W_{p,q}$ > Ĩ/j ÿ3P p ¼ q {L\$ £% W4Ö •3+ È+a 7 , 'Ê y Ä 63<• `\$! Ö - TM 8¢ . ÿ) {L\$, j,´ ...5 ´-(£

W È D TM8¢ . ÿ, ´Eé+| Ÿ 0 Ê0; Ö : ý P ¶\$! ÖEé+| È p Y+X TM8¢ . ÿ Œ j E , Ÿ ÈLb!'\$! ÖGý *E÷0;] ×+O WGÿ Q2ú, ´Eé+| Ä $w_{p,q}$ X TM8¢ . :, ´ Ê y ² ; Ö

$$w_{p,q} = \exp(-(C_p - C_q)^2 / \sigma^2)$$
 Ä4.4 Å

 $FG\ddot{u}, \sigma > \tilde{j}, \dot{\sigma}$ é ò Ä

ÎP¼E÷0;])à :F Gý * Ç `,´\$! Ö .-(" ¾-\$ Õ+a CNN `AÑ,´\$! Ö . È2î Ö 9 ¶ 0 Ê, ´ w È vF _ ^ X Q2úEé+| È,8\$Gý *, ´ j Ÿ5 ´ +O ¶F0 F Ä 6 ÀGý *E÷0; $Q = 7.5 \text{ } \text{$^{\circ}$} \text$ "Ñ X WGÿ2î Ö-()E³ ~, ´\$! Ö ž] Ä=½ 8Gý *E÷0;]"Ñ 9 u 6 U *F Ë £K^\$! Ö "&é, ´2î Ö ¼0°L\$ }5ž H ϊ Ä Ä2 Å Ù.1. Le M ,´ Aî5ž <1y0; Ö,´-; μ ¶ Q 2î Ö, ´\$! Ö "&é ¼ CNN Ø Ý, ´\$! Ö \ddot{V} ÈF > n = 8*6, ´Äp * ¾ : T&é \ddot{C} * È=½ 8 Gý *E÷0; È? \pm j = <\$À, ´\$! ÖAî5ž = <, ´ sGý È £5ž Ö × £ ï _ CNN ØÝ,´\$!ÖÈ $E\acute{e}+|*,`2\^{i} \ddot{O} \ddot{A} p \dot{E})^3/48|>,`\$! \ddot{O}.$ \overline{D} $j!`*) \ddot{A},`5\check{Z} \ddot{O}.$ $H \bullet /\dot{U}.1.$ Le M Ä)5ž Ö . H] , \ddot{y} 3P&ép È 2 % Cã $p \in \{p \mid \overline{D}_p = 0\}$ È I $H_p = 0 \times H_p = 1$ È f $p \in \{p \mid D_p \neq 0\}$ £FÓ Ë •8 -(jD DZE÷0;],´ \$! Ö "&é È § 9 0Q ,´5ž Ö Ä v _ ÈF /ý * $\frac{3}{4}$ 4ß ..., ´ é# 4ÿ h JF• ~? ö L NÈ , ´ È\$ § f ` •B\$ È f-(jF ØE÷0;] "Ñ 9Cã O,´£0+ & È×+O,´\$! Ö "&é] J 5 0 Ë I , W,´ F&é Ä m1ÑF Ë 2 l&é,´ - MŽ h ? Èv63<• `F Ë&é,´ £K^ W ú X \ é#],´Gý?±0; Ö ÈM0?±>Û 8*6 ` "L" ¹ Ä p +X ;1†+• •.ž ÊD DZE÷0;] ×+O, ´\$! Ö "&é, ´5ž Ö Ö

$$H_p = \min((\frac{d_{\text{max}}}{\overline{D}_p})^2, 1), p \in \{D_p \neq 0\}$$
 Ä4.5 Å

 $d_{\text{max}} > \tilde{\ }/j, ' = \text{CNN } \emptyset \acute{\ }, '\$! \ddot{\ } \ddot{\ }$

$$H_p = \min((\frac{d_p}{d_{\text{max}}})^2, 1), p \in \{D_p = 0, D_p \neq 0\}$$
 Ä4.6 Å

 $d_p \sim /j = f$ } ÿ3P&é p D /ë\$! ÖEé+|,´D /ë È £\$! ÖEé+| ÿ3P&éLö 8 $\mathcal Q$]/ë p 0F

,´ ÿ3P&éAÑ1Ç Ç `,´D /ë Ä ¯+X ™8¢ .] ,´Eé+| Ÿ •F |>~/j\$! ÖEé+| \mathcal{Q} È d_{max} jL8 | Ä CNN `AÑ\$! Ö & # ¸ - 4*6Eé+| È é C ×+OE³ W,´ • È p -() ¾ £% j È !5ž Ö \$ ~ Ä 04ø,´5ž Ö Q » ² ; Ö

$$H_{p} \begin{cases} 0 & \overline{D}_{p} = 0 \\ \min((\frac{d_{\text{m a}}}{\overline{D}_{p}})^{2}, 1) & D_{p}^{'} \neq 0 \\ \min((\frac{d_{p}}{d_{\text{max}}})^{2}, 1) & D_{p}^{'} = 0, D_{p} \neq 0 \end{cases}$$

$$\ddot{A}4.7 \, \mathring{A}$$

+X5ž Ö. / Ù.1. Le M È Ç ` Ö

$$\min_{D} \sum_{p} (H_{p}(D_{p} - \overline{D}_{p})^{2} + \lambda \sum_{q \in \mathcal{N}_{4}(p)} w_{p,q}(D_{p} - D_{q})^{2})$$
 Ä4.8 Å

FJE÷ ? Ä 4.8 Å, ´ß Ö j 0 È 04ø, ´0 H@ FJE÷ ;M', ´4ï W3+5 "r@ Ö

$$(H + \lambda W)d = H\overline{d}$$
 Ä4.9 Å

 d, \overline{d} _\$! Ö . D ½ \overline{D} ,´ AGÿ '? ÈH _)@ . Le È)@ 4ï s3P 95ž Ö . H 5 * Ä W >~/j0aL\$ F,´ ù ž ù ß. Le È 9 - $w_{p,q}$ ã Ê Ö

$$W\left(r,c\right) = \begin{cases} \sum_{l \in \mathcal{N}_4(r)} w_{r,l} & c = r \\ -w_{r,c} & c \in \mathcal{N}_4(r) \\ 0 & otherwise \end{cases}$$
 Ä4.10 Å

(r,c) > $\tilde{'}/j$ \tilde{y} 3P, $\tilde{'}$ 3R E \tilde{A} \tilde{j} \tilde{q} "r@ F Z4 \tilde{i} W Q » $\tilde{E}\tilde{A}$ •, $\tilde{'}$ 00+ \tilde{i} \$! \tilde{O} \tilde{z} \tilde{d} X5 \tilde{z} \tilde{O} . \tilde{i} \tilde{d} X5 \tilde{z} \tilde{O} . \tilde{i} 1, \tilde{i} \tilde{G} \tilde{G} . \tilde{G} \tilde{G} . Le \tilde{G} \tilde{G} \tilde{G} . \tilde{G} \tilde{G} \tilde{G} \tilde{G} \tilde{G} . \tilde{G} \tilde{G} \tilde{G} \tilde{G} \tilde{G} \tilde{G} \tilde{G} \tilde{G} . \tilde{G} \tilde

$$d(r) = (S_{\overline{d}} \cdot / S_h)(r) = \frac{((I + \lambda W)^{-1} \overline{d})(r)}{((I + \lambda W)^{-1} h)(r)}$$
Ä4.11 Å

¦] È I > ~/j ... }. Le È r j ÿ 3P3R E Ä: ? 6 Ï •, ´ H FL NÈ@ 6V @ T Z1° ..., ´ €L NÈ È 6 [_ 6 €], ´ $S_{\bar{d}}$ ¼ 6 €], ´ S_h È $(I + \lambda W)^{-1}$ > Û-; @ _ 6 [Œ+X ¾ \bar{d} ¼h :, ´% # . Le Ä 0 > 0P ö\$! ÖEÃ * d + a $S_{\bar{d}}$ ¼ S_h) Ä }5ž, ´&éL"AÑ1Ç Ç ` Ä ? Ä 4.11 Å], ´ 6@ é ?@ ã ¶-\$ Õ"rF6 V •, ´ I = 0c ÊL NÈ È v ý 'fM' d"r@ Ci &, ´L NÈ Ä 0 Ë é# [51] 7- O ¸ -, ´"r@ ? Ä 4.11 Å], ´4ï W3+5 È _ -(" ¾ pG % # é# [53] FO Ö ¸ ' ÈL€ f

j¶L@F 1D FO"r@ 1Ç € È A Ê y "ï x é A,´ 05\$ ') Ä,´4ï W3+5 ²; Ö $(I + \lambda^x W^x) S_d^x = \overline{d}^x$ Ä4.12 Å

<code>|] È $d^x > \tilde{d}^x > \tilde{d}^x</code>$

 $X * \frac{3}{4} = \frac{1}{2} 81 + \bullet$, ´OP ö\$! ÖGý *E÷0;] Èj ¶ ¼ @5 ´ Ÿ ¼0P ö\$! Ö, ´ -(Â = ½ 8Ä 0 éM' È þ T Z@ Ö • 0 WL€ Ö ` +‰ j Ÿ, ´ ...5 ´ Ö Ä 1 ÅGý * -] ¯+X ™8¢.

Œ j 7 , Ÿ ÈLb!'\$! ÖEé+| 4 ×+OE÷ £% x Ä È ¯Ç j Ÿ5 ´ +OF0 F Ä Ä 2 Å) ¾ = <\$À, ´\$! Ö Ÿ ÈAî5ž = <, ´5ž Ö Ä Î ĐD DZE÷0;] ×+O, ´\$! Ö "&é, ´ Ö È X =½ 8E÷0;] u 6 U ¦2î Ö ¼0°L\$ }5ž(© W Ä 0 éM' È+a ¾ SLAM L NÈ) Î & W, ´?± "r ¸Q È"r@ E÷0;] 6 * ¾ ¼5\$. ÿ , ´7-Gÿ - 9>| 9 G 6@ @ 05\$L NÈ È ¦ E • ˜ p , ´05\$ FO"r@ 1Ç € F "r@ ÈX Añ2î Ö = •a, ´õ å ;Eî `¶ SLAM E÷0;) Î & W, ´?±"r Ä

4.3 闭环检测与校正

-(j XK¯&L\$F Ø > È = F¯},´J +OE¯F)%20+ Èj¶@ ãF ØE÷0;]B 3_0 ,´L NÈ ÈGý**¯p 08\$ W ` . È X-(jF Ø ZE÷0;]?±Î & `F >| L)ß ð#{ ¼ Q!"Ä 4. 3. 1 闭环检测

L)ß ð#{,´ h ó M # a _ T Ý f }-(j _ V ` "4ÿAïL E÷,´ `é È£ j Ÿ ½Aö [L NÈ ÄFJE÷") f } £K^ W ¼ ž à],´¶ " £K^ W • ð#{L)ß È§ f,´") é ? 08\ _G÷+X * ¾Aý>» Q » 6 . ÿ5F.1 @?ö?ù ...Aý ÈFJE÷AÑ1Ç?ö?ù ...AýL\$,´-(I W • T Ý T P . ÿ _ V>~/j,´ _ < 0 Z j Ÿ Ä

) ¾ f } £K^ W ÈFJE÷ j?ö ..ž Ê ¦ pG £K^ W ÈAÑ1Ç {L\$,´-(I W Ç 6 È | ® ` 0 ?-(I Ö s_{min} ÄFJE÷?ö?ùAý " ĐFO,´ é ? L3R ¶ " £K^ WAÑ1Ç > > f } W,´-(I W Ç 6 ² W ¾ s_{min} È I.ž Ê jL)ß IF9 W Ä) IF9 WF >|F 5 W ŏ#{ È £ IF9 W,´-(Fë W > f } W,´-(Fë W • ^ XL)ß £3+ Ä 0 >) IF9 W],´ p 9 £K^ WF >| 64ô È) ¾ 0 Z IF9 W È > !-\$ Õ-(F,´ IF9 W 6 j 04ô ÈAÑ1Ç p 9,´4ô Ç 6 È | Ç `!ÿ4ô] Ç 6 0Q,´ £K^ W Ä 0 > f } £K^ W ¼ E³Q 4ô Ç 6],´-(I Ö 0Q,´ £K^ W ' @ L)ß £3+ Ä X ...- SLAM] ÈB 3_0,´8 +a Ö j 7 È 98 +a Ö,´ ûEæ ¼ £0+ Ð : 0 Z j Ö € Ä p j ¶ Î)àL)ß È)¾ f } £K^ W K_i ¼L)ß WK_i ÈF M0?±AÑ1Ç 0 Z-(I 'Ä Sim3 Å S_{il} È6< = õ õ _!W? 'Ä SE3 Å ÄOÆ XAÑ1Ç T W {L\$ p?ò#{ ` .&é,´) Ä £3+ È FJE÷ > {£6,,,´ ORB (@ ±F >| iG} Ä Ç ` 04ô 3D-3D ,´ iG}&é) È Y+X Hom[56] 1y ê *,´ é# X RANSAC v æ ;"r@ -(I '. Le S_{il} Ä ² ÌB -(I '. Le § 9Cã O J,´ µ&é È IF 0!• H F S_{il} È L3R T W {L\$,´(© ±&é È ® `\$ J,´ 2D-3D ,´) Ä £3+ $k \Rightarrow j$ Ä £(© ±&é ` .&é,´) Ä Å

$$\begin{split} e_i &= x_{i,k} - \pi_i(S_{il}, X_{l,j}) \\ e_l &= x_{l,j} - \pi_l(S_{il}^{-1}, X_{i,k}) \end{split} \label{eq:elliptic_elliptic} \begin{aligned} \ddot{\mathsf{A}} 4.13 \ \mathring{\mathsf{A}} \end{aligned}$$

ΙμΗ F,´- 7- jÖ

$$C = \sum_{h} (\rho_{h}(e_{i}^{T} \Omega_{i,k}^{-1} e_{i}) + \rho_{h}(e_{l}^{T} \Omega_{l,j}^{-1} e_{l}))$$
 Ä4.14 Å

$$\begin{split} &\Omega_{i,k},\Omega_{l,j}>\tilde{\ }'j \text{ £K^ W } K_i \text{ $^{1}\!\!/\text{L}$)} \text{ \mathbb{K} W K_l } \text{ \mathbb{K} $\mathbb{K$$

4.3.2 闭环校正

FJE÷:M', '!•PÔ Ç `¶L)ß W ¼) Ä, ´-(I 'ÈI F>|L)ß Q!" Ä OÆ x È f } £K^ W, ´ } / T_{iw} FJE÷-(I '. Le S_{ii} F >| Q!" È Z Q!" E÷0; < g PFB ` K_i , ´ Fë u È ¬ÇL)ß, ´ T1)U€ ÄL)ß W K_i ú ¦Fë u p?ò#{ `, ´ `.&é>Û Å ; ` f } W K_i ú ¦Fë u : È | X Å ;&éLtF L3R iG}, ´(© ±&é Ä) ¾ iG} :, ´ `.&é D _ µ&é, ´ õ å ; ÈF >| `.&é=½ 8 Ä #¹ ú `=½ 8, ´ £K^ W X ;?ö .] \$ àF Õ £3+ Ä wAî } / . H F], ´ £ K^ W, ´ } / j S_{iw} ½ S_{jw} ÈFÓ x T W {L\$, ´-() ' j $S_{iy} = S_{iw} \cdot S_{jw}^{-1}$ È v _ j `AÑE÷0;] ^ X • È ,8\$1y ? | = @ ¶ Ä p Ê y ;M', ´B N© Ö

$$e_{i,j} = \log_{Sim(3)}(S_{ij}S_{jw}S_{iw}^{-1})$$
 Ä4.15 Å

 $S_{ij} > \tilde{\ }/j, \tilde{\ }_- T \text{ WL$, '-()-(I ' \ddot{\ } 9) }$ SE $\ddot{\ } 3 \text{ Å } \} / \varnothing , 6 < \bullet \ \dot{\ } \dot{\ } j \ \ddot{\ } j \ \ddot{\ } 1 \text{ Å } \dot{\ } \dot{\ } S_{iw}$ 1 Å $\dot{\ } S_{iw}$ 1 Å $\dot{\ } \dot{\ } \dot{$

4.4 本章小结

\1 k?± û4ý \(\text{M}'\) A0P öGý \(\frac{\chi}{\chi}, \cdots \text{...-} ?ö?ù \) SLAM \(\epsilon \) | J) \(\text{O} \) \(\text{SLAM} \) \(\epsilon \) | SLAM \(\epsilon \) | SLAM \(\epsilon \) | Slam \(\text{SO} \) | Sla

5 实验与分析

5.1 实验设置与数据集介绍

5.1.1 实验设置与说明

M' A0P öGý *,´ ... - ?ö?ù SLAM 3+5 È ORB-SLAM ,´ $= f \lor = j \cdot p$ È 6 j } !å È >1 È 0P ö\$! ÖGý * >L)ß ð#{1y ZQ‡Ä}!å k?± _ * ¾(© ±,´-(jD DZ È > 1 j` .1Ñ*6 ¼ pG H F È 0P ö\$! ÖGý *4ï0; 5 * ¾ CNN ,´\$! Ö `AÑ ¼\$! Ö=½ 8Gý *ÄP¼] ¯+X,´.œ &)ß ³ j 0 CPU j Intel 2.4GHz Xeon ,´ ? j È 7 32GB RAM ¼ Nvidia TitanX 12GB GPU ÄCNN AÝ4óE÷0;] ¯+X,´ žLö j NYU Depth V2^[57] ÈB žLö 5 [¶•8 464 Z j Ÿ,´ 1449 4ô ™8¢. >) Ä\$! Ö . Ä 5•5 Q »AÝ4ó & È 6 1449 4ô .ÿ) B 6 @AÝ4óLö ¼P¼AñLö TG 6 È 6" [5 [795 4ô ¼654 4ô Ä5•5 Q »,´] * G÷+XFJ+X,´\$! Ö $- \bullet \lor = 0$ Tensorflow È $- \bullet = 0$;],´ ØGÿ ò j 0.9 È M û,´ $- \bullet \to 0$. Aî 5½ j 10^{-4} ÈF 0Ež > È $- \bullet \to 0$. ÿ j $\bullet = 0$ % $\bullet = 0$ %

- Root mean squared error Ärmse Å Ö $\sqrt{\frac{1}{N}\sum_{p}(d_{p}^{gr}-d_{p})^{2}}$
- Average log error Älog Å Ö $\sqrt{\frac{1}{N}\sum_{p}(\log(d_{p}^{gt})-\log(d_{p}))^{2}}$
- Absolute relative error Äabs.rel Å Ö $\frac{1}{N}\sum_{p}\frac{\left|d_{p}^{gt}-d_{p}\right|}{d_{p}^{gt}}$
- Accuracy with threshold thr Öpercentage Ä% Åof d_p s.t. $\max(\frac{d_p^{gt}}{d_p}, \frac{d_p}{d_p^{gt}}) = \delta < thr$

 \vdots] È d_p^{gt}, d_p >~/j X ÿ3P&é p 4-O Î, ´\$! Ö ¼ `AÑ, ´\$! Ö Ä

5.1.2 实验数据集介绍

ÎP¼]#¹ ú,´ žLö 9 NYU Depth V2 dataset^[57] ÈTUM RGB-D SLAM dataset^[58] ¼ ICL-NUIM dataset^[59] Ä

NYU Depth V2 dataset žLö+a Kinect X 464 Z = < j Ÿ ;G÷Lö È] 5 1449 4ô0P ö\$! Ö ¼2« [71® 7# ,´.-() È 407024 Z Z4ÿ 7# ,´?öNÁ W ÄX = EC§!— y,´ õ å ; È \·]1°0 j NYU žLö Ä žLö 5 9 ZG 6 Ö 7# G 6 È Ï û ž ¼ § 1á Ä " 7# ž _ Ï û ž,´ €Lö È+a ™8¢ . È4ÿE÷ <!•,´\$! Ö . ¼0P ö,´2« [71®4ô @È ; . 9/ý2« » ž,´ .../j Ö

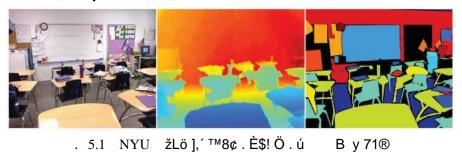


Fig. 5.1 The color image of NYU with its corresponding depth and semantic labels

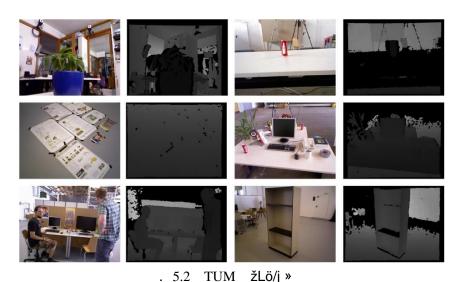


Fig. 5.2 Some samples of TUM dataset

ICL-NUIM dataset -("¾:M',′TZ žLö G-_X-OÎjŸ;G÷Lö9ç,′È ICL-NUIM I_ê 8 @ žLö È+a M-*6 W − Ë ÈjéïE+XÈ·]1°0 j ICL žLö Äk?±9Òμ¼ÎœÔTZjŸ;,′8 Z.ÿ¿GLö4ô @ Ȧ]ÈÒμjŸ;,′ž Ë™8¢.È)Ä\$!Ö.È-(j,′}/¼jŸ,′95\$Q»È6<ΜÔjŸ;=5 95\$Q»Ä žLö FË> TUM RGB-D ž-(¬é,′(x\È£F2+X¾p9 TUM RGB-D žLö],′4*6¼Aô'§Ä ICL-NUIM žLö,′G 6.../j²;Ö









. 5.3 ICL-NUIM žLö/j »

Fig. 5.3 Some samples of ICL-NUIM dataset

5.2 相机跟踪与闭环检测实验

-(jD DZE÷0; ˶ ö.ž,´ } / ¼00+¿,´` .&é È Ë Ÿ,´2î Ö) ¾ Î)à * ¾ ...
- ?ö?ù SLAM ,´0P öGý * •B\$8# £Gý?± Ä 0 éM' È 3_0 ,´ 4*6) ´ * ~ p 08\$,´
` . 9Gý?± ? y Ä\8² k?±Aô '\·],´-(jD DZE÷0;,´ } / `AÑ ¼L)ß x Ì Ä B G 6
ÎP¼] ¬+X,´ § f ?öNÁ ¿ G²; Ö ICL žLö],´ of kt0 Ãof kt1 Ãlr kt0 Ãlr kt1 × TUM
žLö],´ fr3/long_office_household (fr3_long) Èfr3/structure_texture_far (fr3_str) Ä

>~ 5. 1 -(j,´5)E~F)B ,´" E³ Ä ... } Ö2£ Å

Tab. 5.1 Comparison in term of Absolute Trajectory Error [m]

ž Lö	ICL/of	ICL/of	ICL/l	ICL/lr	TUM/fr3_	TUM/fr3	Ava
	kt0	kt1	r kt0	kt1	long	_str	Avg.
LSD-SLAM [21]	0.538	0.768	0.516	0.480	1.826	0.987	0.853
The Proposed SLAM	0.420	0.780	0.473	0.133	1.206	0.740	0.625

>~ 5.1 ÿF $\P \ 3+5 \ \tilde{A}\tilde{N}, -(j) / \frac{1}{4}$ LSD-SLAM , "E³ È "E³ 7 7 j ATE È " $\frac{3}{4} \ 4.1 \ 8^2 \] û 4 \acute{y} \ddot{A} \ þ>~] -; * È \ · * <math>\frac{3}{4}$ ORB-SLAM D DZE÷0; \hat{I})à, \hat{A} , -(\hat{I}) \ A \tilde{N} > n, 'H $\frac{3}{4}$ LSD-SLAM , '5 \hat{I} \ddot{A} X T2« \hat{I} \dot{I} \dot{I} C \dot{I} \dot{I} \dot{I} LSD-SLAM , '5 \hat{I} • H \dot{I} \dot{I}

 $\P * \frac{3}{4} (\textcircled{o} \pm, \stackrel{?}{?} \ddot{\circ} ? \dot{u} \quad SLAM , \stackrel{?}{E} }{2\hat{i}} \ddot{\circ} ? \pm H \quad \frac{3}{4} + \mathring{\circ} \ddot{\circ} DDZ\# , \stackrel{?}{E} }{2\hat{i}} \ddot{\circ} \ddot{A} . \qquad 5.4 \] \ 6D \ DZE \div 0; \] - (jE^{-}F) \ \mathring{A}_{i} \ \frac{1}{4} + \mathring{A}_{i} + \mathring{A}$

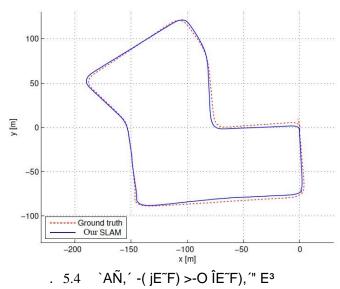


Fig. 5.4 Comparison between estimated trajectory and groundtruth trajectory

5.3 稠密深度重建实验

OP ö\$! ÖGý *G 6,´ÎP¼ k?±#¹ ú `\$! ÖGý *7-Gÿ - ,´ò .ž Ê ÎP¼ ÈJ/ý j Ö €"r@ é# ,´)" ÎP¼ ÈGý *\$! Ö,´ OP ö W ½2î ÖAô ` ú Q ‡ 9 x W,´ P¼Añ ÎP¼ Ä B G 6+X `,´ žLö 9 NYU Depth V2 žLö],´ bathroom 0003 ¼ kitchen 0046 xTUM RGB-D SLAM žLö],´ fr1 rpy Èfr2 dishes Èfr2 desk Èfr3 long office household Èfr3 nostructure texture near with loop ú fr3 structure texture far xCL-NUIM žLö],´ lr kt0 È lr kt1 Èlr kt2 Èof kt0 Èof kt1 ¼ of kt2 Ä

5.3.1 参数适配与尺度因子求解实验

\$! $\ddot{O}=\frac{1}{2}$ 8 > \ddot{G} * - \ddot{A} 4.8 Å] ^ X T ZC μ ò λ $\frac{1}{4}\sigma$ È j ¶.ž Ê ò)"r@ 0c Ê W $\frac{1}{4}$ '=\$! \ddot{O} CXGÿ, ' į ý ÈB G 6 ÎP¼ XF9 Ê, ' NYU žL \ddot{O} : \ddot{O} +X ¶ = <, ' ò Aî5ž • Aô ` 04ø, 'Gý *\$! \ddot{O} . \ddot{A})F9 Ê, '!ÿ 0 Z? \ddot{O} NÁ ¿ GF G \ddot{U} AÑ1Ç ¶Gý *\$! \ddot{O} , ' rmse ½

accuracy ÄAÑ1Ç é ??ñ4.1 8² Å È . 5.5 .../j ¶B 5 Ì Ä .] Eé4Ò8¢,´4åE¤>~/j rmse È #Eé; 8¢,´4åE¤>~/j2î Ö ÈL8 I j 1.25 Ä1\ 0>| _ þ £K^ W] Z ,´ RGB ‡ ¼) Ä,´ -O Î\$! Ö ‡ Ä X σ ,´ "4ï .] È j ¶ \$ -,´6 ÀL NÈ È "4ï] 0 Ë £K^&é }5ž,´) Ä\$!

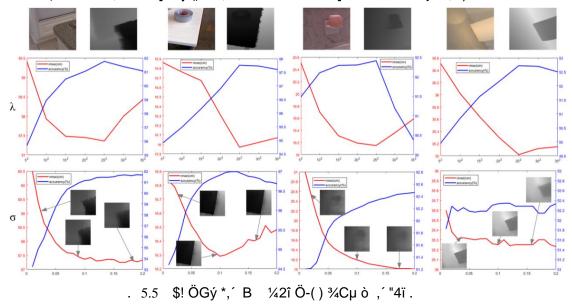


Fig. 5.5 Average keyframe reconstruction rmse and accuracy with respect to λ , σ

Ö ‡>Û .../j * • Èé ï >1\ 0>|],´-O Î\$! Ö" E³ Ä £ ¾ ò • O W,´ 6 À ¼.ž Ê ² ; Ö 1 Å λ ÖB ò _+X •B3 ¤Gý * -] žN© ¼!" I FN©,´ £>' È 8F2,´ I7- O)5 ÌF >|F2 f,´ £% È £ 5 ¹ m 7- J,´ F&é Ä'f6< È IE÷ W & J > n ,8\$5 Ì ,´2î Ö ;L} Ä . 5.5] n/j ¶ $\lambda \in [5^2,35^2]$ & ×+O,´5 Ì,´ rmse ¼2î Ö,´ "4ï . Ä # ? È X $\lambda = 25^2$ 4 _ 0 Z &é È D!" 4Gý * *,´\$! Ö . § 9 0Q 2î Ö ¼ 0 ~B È p 6 λ ,´ IAî Ê j 25^2 Ä

 $2 \ \text{Å} \ \sigma \ \text{Ö7-G\"{y}} - \ \text{Ä} \ 4.8 \ \text{Å}, \ \hat{\text{E}} \ \text{W} \ \text{W} \ \text{Ö} \bullet , \ \text{\'} j \ \text{Ö} + a \ \hat{\text{E}} \ \text{y} \ \text{X} \ \text{TM} \ \text{8¢} . :, \ \text{\'} \text{O}^a \text{L} \ \text{F} \ \text{s} \ \text{G\'{y}} - \\ 8 \ \text{F} 2 \ \text{Ä}, \ \tilde{\text{A}} \ \hat{\text{E}} \ \hat{\text{E}} \ \text{`o} \ \sigma \times \text{f} \ \text{¶} \ \text{s} \ \text{G\'{y}} \ \text{)} \ \text{$^!} \ \text{\"{O}} \ \text{E\'{e}} + |, \ \text{\'} \text{\'} \ \text{\'} \ \text{\o} \ \text{\o} \ \text{\'} \ \text{$^!} \ \text{\'} \ \text{O} \ \text{`i} \ \text{`} \ \text{$^!} \ \text{`o} \ \text{`i} \ \text{`o} \ \text{`e} \ \text{`i} \ \text{`o} \ \text{`i} \ \text{`o} \ \text{`e} \ \text{`o} \ \text{`i} \ \text{`o} \ \text{`e} \ \text{`e} \ \text{`o} \ \text{`o}$

X 3.4 8² û4ý ¶ j Ö €,´ = <"r@ é# È £ mean ratio È least square ¼ RANSAC based least square fittin Ä \ 8² 6 þ ÎP¼@ Ö)" F /ý é# ,´ H Ó W È 6 [Y+X :M',´

é# "r@ j Ö € È ¦ X&é)Lö 8 χ_D ¼ $\chi_{\overline{D}}$:AÑ1ÇB ÈF9 L8 I $T \in [0.18, 0.46]$ È5 AÑ µ&é p •,′" » È5 f-\$ é . 5.6 È žLö j NYU Ä \$5 Š,′-; `È \ · p ¯+X ,′ * ¾ RANSAC 0 ? ¼ ˆ 8,′ é# X = <L8 I ;G- > n H ¾ | ³ T/ý é# Ä

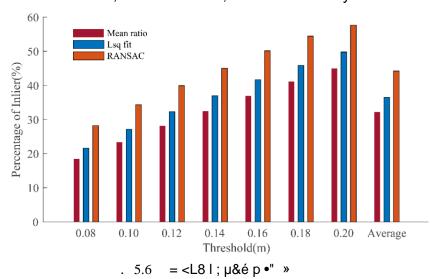


Fig. 5.6 Percentage of inliers under different threshholds

5.3.2 深度重建的稠密性评估

>~ 5. 2 7 ö žLö :Gý *0P ö W, ´" E³

Tab 5.2	Comparison in	term of reconstruction	n density on	benchmark datasets
1 aD. D.Z	Companison in	Term of reconstruction	m aensiiv on	Deficilitatik datasets

žLö —	Percentage of Correct Depth							
2L0 —	ORB	LSD	LSD-BS	Remode	CNN-SLAM	Ours		
ICL/ 'lr kt0'	0.03	0.36	1.43	4.48	12.84	24.28		
ICL/ 'lr kt1'	0.02	0.06	3.03	2.43	13.04	37.10		
ICL/ 'lr kt2'	0.01	0.17	1.81	8.68	26.56	28.69		
ICL/ 'of kt0'	0.02	0.33	0.60	4.48	19.41	13.65		
ICL/ 'of kt1'	0.02	0.04	4.76	3.13	29.15	43.46		
ICL/ 'of kt2'	0.04	0.08	1.44	16.71	37.23	39.59		
TUM/fr3_long	0.03	0.09	3.80	9.55	12.48	19.26		

TUM/fr3_str	0.03	0.04	6.45	6.74	27.40	40.07
Average	0.03	0.23	3.03	7.65	22.46	28.98

>~ 5.2 G * ¶ p 9 :F é# {L\$, ´" E³5 Ì Ä ¸ > n È \ · ' = , ´\$! Ö ." ORB-SLAM ÈLSD-SLAM ?±0P ö, ´J È+a ¾ ³ "Ñ 90P öGý *, ´ é x Ä £ ï _ ¼ Remode È CNN-SLAM -(" È \ · *, ´ é# 9ç Ç, ´\$! Ö0P ö W • \$Q Ä F È þ ž] -; `LSD-SLAM Gý *5 Ì, ´ £ w0P ö W?±F F Q ¾ ORB-SALM , ´5 Ì È _ j-\$ ÕD DZ# H F, ´ _ Ï û ÿ3P, ´ Þ Ö Ÿ È Gý * * z0P ö ` . Ä*6Aê : •B\$ È * ¾F 2 « é # , ´0P öGý * é x È Remode ÈCNN-SLAM È \$ Ð é C ´ *0P ö ` . È-(" ¾ \ · * ¾ (© ±, ´ SLAM é# Ä v _ È X ÎLu õ å ; È Ç, ú ¾=½ 8 v æ ; J Z Q ‡ • <+X È \ · *, ´ é# X5 W J žLö :9ç Ç ¶ \$ -, ´5 Ì Ä

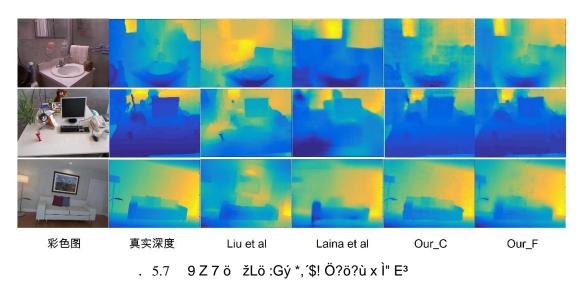


Fig. 5.7 Visual comparison of reconstructed depth on three benchmark datasets

5.3.3 深度重建的精度评估

Gý *\$! Ö, ´2î ÖAô ` k?± þ T Z@ Ö ÈÊ W ¼ ÊGÿ@ Ö" E³ ¶ \ , ´\$! Ö=½ 8 é# ÄOur_F Å È Weerasekera 1y ê *, ´£% W4Ö • Ä PE_S Å ¼ "M'# 4ï4Ö • Ä PE_N Å, ´ ...- 0P öGý * é# ÈB é# _ * ¾-\$ Õ# , ´ 0 ? F Þ ÖB Î)à0P öGý *, ´ > ˜ È ¯+X ¶ T/ý4Ö • ' & È 6 [¼ @ ¶ ... - 0P ö * . Ä F È \ ÎP¼] • G * ¶ CNN \$! Ö `AÑ , ´5 Ì È ï ¾\ \ *, ´ CNN 5 ´9ç Ç, ´5 Ì Ä Our_C ÅF > | " E³ Ä X 7 ö ž Lö :, ´ ÊGÿ" E³5 Ì ²> ~ 5.3 p/j È § f, ´ 6 À ² ; Ä

>~ 5. 3 7 ö žLö :Gý * 2î Ö,´ ÊGÿ' E³

Tab. 5.3 Quantitative comparison of reconstruction results on three benchmark datasets

žLö / é# -		Error	(lower is	better)	Acc	Accuracy (higher is better)			
		rmse	log	abs.rel	δ <1.25	δ <1.25 ²	δ <1.25 ³		
	Eigen et al.	0.64	0.23	0.16	0.74	0.94	0.98		
	Liu et al.	0.73	0.33	0.33	0.59	0.81	0.91		
	Laina et al.	0.51	0.22	0.18	0.84	0.94	0.97		
NYU Depth V2	Our_C	0.50	0.19	0.16	0.81	0.95	0.98		
	PE_S	0.52	0.21	0.12	0.83	0.95	0.98		
	PE_N	0.45	0.17	0.09	0.89	0.96	0.99		
	Our_F	0.44	0.16	0.09	0.90	0.97	0.99		
	Eigen et al.	1.41	0.37	0.23	0.54	0.82	0.92		
	Liu et al.	0.86	0.29	0.25	0.54	0.87	0.90		
	Laina et al.	1.07	0.39	0.25	0.49	0.75	0.88		
TUM RGB-D	Our_C	0.70	0.28	0.20	0.63	0.88	0.93		
	PE_S	0.69	0.25	0.13	0.79	0.89	0.96		
	PE_N	0.65	0.24	0.12	0.83	0.90	0.96		
	Our_F	0.62	0.23	0.10	0.83	0.95	0.97		
	Eigen et al.	0.83	0.43	0.30	0.47	0.78	0.90		
	Liu et al.	0.81	0.41	0.45	0.47	0.71	0.87		
	Laina et al.	0.54	0.28	0.23	0.59	0.83	0.95		
ICL-NUIM	Our_C	0.36	0.18	0.16	0.74	0.96	0.98		
	PE_S	0.32	0.18	0.12	0.83	0.97	0.99		
	PE_N	0.22	0.12	0.07	0.93	0.99	0.99		
	Our_F	0.30	0.13	0.14	0.89	0.99	0.99		

||Q È \ \ ,´\$! Ö=½ 8 é# Ä Our_F Å X } T Z žLö]>~)à?± H ¾ Weerasekera 1y

ê,´ é#ÄPEÅÈv_X ICL žLö:?± 0&éÄ j PE_N é# 0?F,´_ÞÖB È

 $[\]downarrow$ } _ \triangleright Ö 08\$ W wAî ÈF X-O Î j \ddot{Y}] \downarrow = k _ @0û È 6 J ,8\$ `AÑ, ´5 \dot{I} x+O W

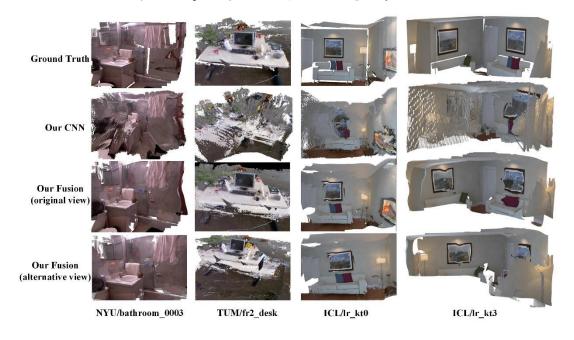
[,] B Ä-(ýÈ\-,´=½8é# = F 0 wAî,´4Ö•Èp X-OÎ žLö:,´5 Ì>˜)à

[,] '\$ - $\ddot{A} \times 0 > 0 \times 28 \otimes \ddot{z} = \ddot{b} - ; * \dot{y} = \ddot{y} = \ddot{N} + ... F \stackrel{.}{E} = \ddot{N}$

PE é#]ižÞÖ 08\$ WF>|B HF ˶¸ W,´Î,ú Ä 0 éM' È-(jF ØE÷0;] =

O u 6,´£0+ ú4ß ûEœF Ø,´^X ,8\$ * ¾(© ±,´D DZ2î ÖL} ~ È04ø ¯ Ç \ ·,´=½ 85 Ì,´W7- 9 p ;L} Äk,´•B\$ È\ · *,´*¾ ...- ?ö?ù SLAM ,´0P öGý * é# Eî ` ¶N´O,´?±"r È T \ X63<• ` õ ¯+X ¶ ± \00+¿,´D 7&é,´ õ å ; È \ 5 [,´\$! Ö4ï 3RF F ~ ¾ PE 1y é# Ä

 $0 > \dot{E} m1\tilde{N} ORB (© \pm \&\acute{e}_300 + \grave{e} \dot{E} v_{+})5\check{z} 1/4 | 5 [¶G\acute{y}?\pm, ´\$! \ddot{O}4\ddot{i}3R \dot{E}F \ddot{E}4\ddot{i}]$



. 5.8 9 Z 7 ö žLö:Gý *,´`.,´.../j

Fig. 5.8 reconstructed dense map on three benchmark datasets

3R -> CNN Ø Ý, ´\$! Ö-(Â>• Ä ORB (© ±&éLö] X4é*6 ` ü, ´ j È 5 Eé+| 4 È 6<CNN Ø Ý\$! ÖEé+| ^ X Ó ï Ä-(ý È CNN X ~4é*6 j ý'f `AÑ * ~ p2î.ž, ´ \$! Ö ÈF _ * ¾(© ±, ´ SLAM p # Š `, ´ Ä > ~ 5.3 ¼ . 5.7] •Añ > ¶F 0&é Ä+a ¾ =½ 8 ¶ T/ý Â>•, ´ = <\$À\$! Ö È \ .9ç , ´\$! Ö5 Ì-(" ¾ õ ¯+X CNN 5 ´, ´N´#{5 Ì È?± \$2î.ž Ä 0 éM' È . 5.8], ´95\$ Gý *5 Ì n/j \ · *, ´0P öGý * é# 2î.ž , ´ ' = ¶ j Ÿ, ´5 ´ ¼>~M' ÈF • þ?ö?ù x Ì @ Ö ½!QAñ > ¶ \ ·=½ 8 v æ, ´ 9 x W Ä



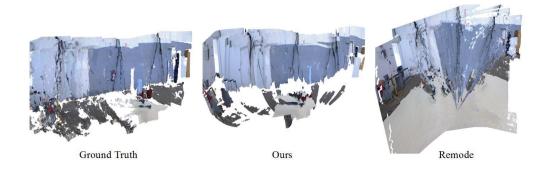
. 5.9 4ß ûEœF Ø;, Gý*xÌ

Fig. 5.9 Reconstruction results under pure rotational camera motion

5.3.4 模块有效性的验证

Q ± 9 x W, P¼Añ k ?± 6 @ 9 ZG 6 È 0 Z CNN \$! ÖØÝQ ‡ X A H W)ß 3; , ´R± WP¼Añ È FTZG 6 =½ 81†+•, ´9 x W ¼5ž Ö ,, ´8*6 WP¼Añ Ä } M' \ \ *,' SLAM 3+5 -(" 3/4 P5,'...-?ö?ù SLAM,'H ï { 0 a X 4ß ûEœF Ø F65 ~4é*6)ß ³; ý'f7- OR± ,´F >| È<-'f2î Ö J `; ý ÄF _ j v æ] ^-+X ,′ CNN ØÝ\$! ÖQ $^+$ XF Ë 0 3; §90 0 £,′9xWÄj 0 Añ>F 0& 0 £È\ G 6 ÎP¼ X TUM \check{z} Lö;, fr1 rpy ¼ fr2 dishes ? \check{o} NÁ; G] F > $| \P G \circ \mathring{v} \rangle$ fr1 rpy" ¿ G]-(j, ´ ō Š4ß ûEœF Ø È fr2 dishes ?öNÁG÷Lö ¾ ~4é*6 j Ÿ ; È » ² y%, ´É ñ È |M' % Ä . 5.9 $] n/j \P \setminus$, ´é# % LSD-SLAM) fr1 rpy", ´Gý *5 Ì È -; *\·é#, ´Gý *5 Ì] Í'f7- O-;\$5 j Ÿ, ´ ...5 ´ È6< LSD-SLAM ,´5 Ì] +O¶ UGý•" 'ÄF _ j4ß ûEœF Ø.d•¶0û f?ö?ù],´ ...) Ä £3+ È v CNN \$! \ddot{O} `A \tilde{N} \ddot{m} " \tilde{N} 9 ` + ... ; \dot{y} \ddot{A} 0 \dot{e} M' \dot{E} . 5.10] ... ¶ \ \, ' \dot{e} # ½ Remode é#) fr2 dishes , Gý *5 Ä+a %Remode é# _ * ¾ 4ß ..., é# È ÍC† ¾ Þ Ö 0 8\$ W wAî ¼0û f iG} ° _ •"r@ 0P ö\$! Ö Èv _5j •Cã O4é*6,´ j Ÿ ¯ Ç iG}2î Ö UFO ;L} È 04ø È ,8\$ Remode Gý *,´ j ŸF0 F UGý Ä-() •B\$ È \ · é# ×+O,´ j Ÿ Q » \$ ÕF -O Î õ å Ä

j ¶P¼Añ=½ 81†+•,´9 x W ¼5ž Ö .,´8*6 WP¼Añ È ÎP¼]6 Liu 1y ê ¼ Laina 1y ê,´ CNN \$! Ö Ø Ý é# >\-,´=½ 81†+•-(5 8 È £ Liu et al+Fusion ¼ Laina et al +Fusion Ä" E³,´5 Ì X>~ 5.4] G * È -; * È X p 9 žLö : ÈG-9 , > n,´x Ì



. 5.10 ~4é*6)ß ³ ;, 'Gý * x Ì
Fig. 5.10 Reconstruction results under low-texture scene

>~ 5. 4 7 ö žLö : Q ‡ 9 x WP¼Añ

Tab. 5.4 Quantitative verification of models on three benchmark datasets

žLö / é# -		Error	Error (lower is better)			Accuracy (higher is better)		
		rmse	log	abs.rel	δ <1.25	δ < 1.25 ²	δ < 1.25 ³	
	Liu et al	0.73	0.33	0.33	0.59	0.81	0.91	
	Liu et al + Fusion	0.65	0.30	0.29	0.62	0.83	0.94	
	Laina et al	0.51	0.22	0.18	0.84	0.94	0.97	
NYU Depth V2	Laina et al + Fusion	0.44	0.19	0.16	0.85	0.95	0.98	
	Our_C	0.50	0.19	0.16	0.81	0.95	0.98	
	Our_F w/o Confidence	0.48	0.20	0.16	0.83	0.95	0.98	
	Our_F	0.44	0.16	0.09	0.90	0.97	0.99	
	Liu et al	0.86	0.29	0.25	0.54	0.87	0.90	
	Liu et al + Fusion	0.81	0.28	0.24	0.56	0.89	0.95	
	Laina et al	1.07	0.39	0.25	0.49	0.75	0.88	
TUM RGB-D	Laina et al + Fusion	0.91	0.32	0.22	0.57	0.82	0.92	
	Our_C	0.70	0.28	0.20	0.63	0.88	0.93	
	Our_F w/o Confidence	0.67	0.26	0.18	0.67	0.90	0.94	
	Our_F	0.62	0.23	0.10	0.83	0.95	0.97	
ICL-NUIM	Liu et al	0.81	0.41	0.45	0.47	0.71	0.87	

Liu et al + Fusion	0.64	0.32	0.34	0.55	0.82	0.92
Laina et al	0.54	0.28	0.23	0.59	0.83	0.95
Laina et al + Fusion	0.41	0.23	0.19	0.65	0.89	0.98
Our_C	0.36	0.18	0.16	0.74	0.96	0.98
Our_F w/o Confidence	0.35	0.17	0.16	0.76	0.97	0.98
Our_F	0.30	0.13	0.14	0.89	0.99	0.99

5.4 运行时间分析

\ \ * ,´ SLAM \(\epsilon \) k? \(\epsilon \) ZG \(6M0? \) \(\ext{8A} \) \(\ext{SL} \) \(\

> 5. 5 = < 9. ÖGý * é#, 2î Ö > &L\$" E³

Tab. 5.5 Comparison of different depth reconstruction method in accuracy and time

é#	GF[54]	DT[53]	WLS[60]	Ours
Runtime(s)	0.16	0.09	3.9	0.15
RMSE(m)	0.45	0.47	0.36	0.39
Accuracy (δ < 1.25)	75.2%	77.3%	88.4%	85.6%

5.5 本章小结

\1 û4ý ¶M' A0P öGý *,´ ...- ?ö?ù SLAM é# -(£ ÎP¼,´AîAÑ ¼ 6 À ÆÆ x û4ý ¶ ÎP¼E÷0;],´Aî5ž >+X`,´ žLö Ä k?±,´ÎP¼Aô `E÷0; 6 @ 9 ZG 6 È1\ 0 ZG 6 _-(jD DZ ¼L)ß x Ì,´P¼Añ Ä -(jD DZE÷0; ˶ ö.ž,´} / ¼00+¿,´` .&é j >5 ,´Gý * ËGý?± Í ž È2î.ž,´L)ß1Ç# ¸ -,´ÿE« ¶B ,´3_0 È) ´ *~ p 08\$,´`

结 论

*¶...-?ö?ù SLAM]^X,´L NÈ ÖGý *,´`. Z00+¿ Èõ õ F2+X ¾-(j,´Ê} È6< # Ä+X`ÎLuL NÈ] È !" * ¶ * 3/4\$! Ö=1/2 81†+•, ´ M' A0P öGý *, ´ ...- ?ö?ù SLAM é# Ä Õ-p È\· û4ý ¶ SLAM L NÈ, ´-(£ *.p*6Aê >6ü Ÿ. Aö È-pGýAØAê ¶ = <2« » P O ~ Î)à0P öGý *, ´ H5j&é È)" ¶(© ±# >-\$ ÕD DZ# SLAM X0P öGý * éM', Lî &é >5jL§ È ¦ 5 80P öGý *]?±@ ã,´9 ZL NÈ Ö Ä1 Å > g9ç Q 2î Ö,´j Ÿ ...5 $' > -(i) / \ddot{A} \ddot{A} \ddot{A} \ddot{A} \times \dot{A} \times \dot{$ $C_i \ddot{Y}, 5 ' \% \dot{I}_i, 0P \ddot{o} \dot{S}! \ddot{O} > \dot{E} > g \hat{I}) \dot{a} T65, '= \% 8 \dot{E} F65B \dot{E} > g 6F \ddot{E}0P$ \ddot{o} , \dot{s} ! \ddot{O} \ddot{Y} \rightarrow \dot{u} \ddot{y} 5 \dot{f} 5 \dot{g} 6 \ddot{g} 7 \ddot{g} 7 \ddot{g} 8 \ddot{g} 9 $\ddot{g$ Ä 04ø.ž Ê ¶ \ ·,´.D 0¦ MD Öþ j Ÿ5 ´¼0P ö\$! Ö`AÑ T Z@ Ö * È u 6 Y+X ¶ * ¾(© ±,´-(jD DZ 1/4 CNN \$! ÖØÝÈ9x,′5 8¶4ß ...,′1†+•¼\$! Ö-•M#È **¾=½81†+•,′ 0P \ddot{o} \$! \ddot{O} Gý * \dot{E} \hat{I}) \dot{a} ¶) 95\$ \dot{I} \ddot{Y} , \dot{O} P \ddot{o} * \dot{Q} \ddot{A} 0 > \dot{E} \-FJE \div ¼ \dot{y} , \dot{I} P¼P¼Añ ¶-(\dot{I} D DZ >0P ö\$! ÖGý *,´ 9 x W È ¦ 6 À ¶ ¤ Z3+5 ,´ Î & W È5, 8>~ > ¶M' A0P öGý * ...- ? \ddot{o} ? \dot{u} SLAM 3+5 § 9 $E^{3}Q$, '2 \dot{l} \ddot{O} >R± W \ddot{A}

\-,´k?±CQ)^5 Ö

Ä1 Å ´*¶0‡¼ ¤,´ M' A0P öGý *,´...- ?ö?ù SLAM 3+5 ÈOÆ!@ 8¶ *¾(© ±,´-(jD DZ > CNN \$! Ö Ø Ý Q‡È *6 -(jD DZ ¼0P ö\$! Ö `AÑ@ 6V,´## È u 6 Y+X00+¿D 7&é,´0°L\$ }5ž ¼2î Ö H ï È > CNN \$! Ö `AÑ ' @ Â>• È ¼ @ j Ÿ,´0P öGý * Ä

Ä3 Å ´* ¶ * ¾ Resnet 5 ´,´ J j Ö\$! Ö `AÑ5•5 È ¯+X0ª#N §0 / Ï û ;G÷ g r È X Î W O Gþ,´ < & È"Ñ 9 E •NÍ F ò ¼FP @(© ± Ÿ,´ • a Ä J j Ö Q ‡,´ E • È Ç5•5 7- OAö [J j Ö,´)B' $|2\hat{i}.z \varnothing Y *|)$ Ä,´\$! Ö Ÿ Ä 0 éM' È CNN \$! Ö `AÑ j ...- ?ö?ù SLAM Ë ¶ 0 Z j Ö ò63 $|\hat{i}|$ ¶1Ç# X ~4é*6 j Ÿ ¼4ß ûEœ F Ø ;,´R± W Ä

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参考文献

- [1] Ventires M. Stanley: The robot that won the darpa grand challenge [J]. Journal of Field Robotics, 2006. 23(9):661-692.
- [2] Urmson C, Anhalt J, Bagnell D, et al. Autonomous driving in urban environments: Boss and the urban challenge[J]. Journal Field of Robotics, 2008. 25(8):425-466.
- [3] Durant-Whyte H, Bailey T. Simultaneous localization and mapping: part I[J]. Robotics & Automation Magazine, 2016. 13(2):99-110.
- [4] Scaramuzza D, Fraundorfer F. Visual odometry [Tutorial][J]. IEEE Robotics & Automation Magazine, 2011. 18(4):80-92.
- [5] H#TM È1 ÈR½<~ Ë. *¾ ...- ?ö?ù,′ < & Ê }¾`.´* é# 5,F [J]. AÑ1Ç jEμ ÙAîAÑ >.'--Õ È 2016. 28(6):855-868.
- [6] Longguet-Higgins H. A computer algorithm for reconstructing a scene from two projections[J]. Nature, 1981. 293(10):133-135.
- [7] Harris C, Pike J. 3D positional integration from image sequences[C]. Alvey Vision Conference. 1988:87-90.
- [8] Frahm J, Georgel P, Gallup D, et al. Building rome on a cloudless day[C]. European Conference Computer Vision. 2010:368-381.
- [9] Cadena C, Carlone L, Carrillo H, et al. Simultaneous localization and mapping: present, future, and the robust-perception age[J]. IEEE Transactions on Robotics, 2016. 32(6): 1309 1332.
- [10] Q 6 È P#Ë È H!õ È1y. ?ö?ùSLAM q Aâ Ö þ*6Aê`ÎD% [M]. G Ü Ö+e € J *(x/n È 2017.
- [11] Strasdat H. Local accuracy and global consistency for efficient visual SLAM[D]. London: Imperial College London, 2012.
- [12] Davison A J, Reid I D, Molton N D, et al. MonoSLAM: Real-time single camera SLAM[J]. IEEE Transactions on Pattern Analysis & Machine Intelligence, 2007. 29(6):1052-1067.
- [13] Shi J, Tomasi C. Good features to track[C]. IEEE Conference on Computer Vision and Pattern Recognition. 1994:593-600.
- [14] Pupilli M, Calway A. Real-time camera tracking using a particle filter[C]. British Machine Vision Conference. 2005.
- [15] Civera J, Davison A J, Montiel J M M. Inverse depth parameterization for monocular SLAM[J]. IEEE Transactions on Robotics, 2008. 24(5):932-945.
- [16] Mourikis A I, Roumeliotis S I. A multi-state constraint Kalman filter for vision aided inertial navigation[J]. 2007, 22:3565-3572.

- [17] Strasdat H, Montiel M.M, Davison A. Visual SLAM: Why filter[J]. Image and Vision Computing, 2012. 30(2):65-77.
- [18] Klein G, Murray D. Parallel tracking and mapping for small AR workspaces[C]. IEEE International Symposium on Mixed and Augmented Reality. 2007:225-234.
- [19] Klein G, Murray D. Improving the agility of key-frame based SLAM[C]. European Conference on Computer Vision. 2008:802-815.
- [20] Mur-Artal R, Montiel M. M, Tardos J. D. ORB-SLAM: A versatile and accurate monocular SLAM system[J]. IEEE Transactions on Robotics, 2015. 31(5):1147-1163.
- [21] Engel J, Schops T, Cremers D. LSD-SLAM: Large-scale direct monocular SLAM[C]. European Conference on Computer Vision. 2014:834-849.
- [22] Tan W, Liu H, Dong Z, et al. Robust monocular SLAM in dynamic environments[C]. IEEE International Symposium on Mixed and Augmented Reality. 2013:209-218.
- [23] Qin T, Li P, Shen S. VINS-Mono: A robust and versatile monocular visual-inertial state estimation[J]. IEEE Transaction on Robotics, 2018. 34(4):1004-1020.
- [24] Keller M, Lefloch D, Lambers M, et al. Real-time 3D reconstruction in dynamic scenes using point-based fusion[C]. IEEE International Conference on 3D Vision. 2013:1-8.
- [25] Whelan T, Kaess M, Johannsson H, et al. Real-time large scale dense RGB-D SLAM with volumetric fusion[J]. The International Journal of Robotics Research, 2015. 34(4):598-626.
- [26] Whelan T, Salas-Moreno R. F, Glocker B, et al. Elasticfusion: Real-time dense SLAM and light source estimation[J]. The International Journal of Robotics Research, 2016. 35(14):1697-1716.
- [27] Newcombe R, Izadi S, Hilliges O, et al. Kinectfusion: Real-time dense surface mapping and tracking[C]. IEEE International Symposium on Mixed and Augmented Reality. 2011:127-136.
- [28] Concha A, Hussain M. W, Montano L, et al. Manhattan and piece wise-planar constraints for dense monocular mapping[C]. Robotics: Science and System. 2014
- [29] Flint A, Murray D, Reid I. Manhattan scene understanding using monocular, stereo, and 3D features[C]. IEEE International Conference on Computer Vision. 2011:2228-2235.
- [30] Concha A, Civera J. DPPTAM: Dense piecewise planar tracking and mapping form monocular sequences[C]. IEEE/RSJ International Conference on Intelligent Robots and Systems. 2015:5686-5693.
- [31] Newcombe R. A, Lovegrove S. J, Davison A. J. DTAM: Dense tracking and mapping in real-time[C]. IEEE International Conference on Computer Vision. 2011:2320-23327.
- [32] Pizzoli M, Forster C, Scaramuzza D. Remode: Probabilistic, monocular dense reconstruction in real time[C]. IEEE International Conference on Robotics and Automation. 2014:2609-2616.

- [33] Herrera D, Kannala J, Heikkila J, et al. Depth map inpainting under a second order smoothness prior[C]. Scandinavian Conference on Image Analysis. 2013:555-566.
- [34] Engel J, Cremers D. Semi-dense visual odometry for a monocular camera[C]. IEEE International Conference on Computer Vision. 2013:1449-1456.
- [35] Greene W. N, Ok K, Lommel P, et al. Multi-level mapping: real-time dense monocular SLAM[C]. IEEE International Conference on Robotics and Automation. 2016.
- [36] Weerasekera C. S, Latif Y, Grag R, et al. Dense monocular reconstruction using surface normals[C]. IEEE International Conference on Robotics and Automation. 2017:2524-2631.
- [37] Eigen D, Fergus R. Predicting depth, surface normals and sematic labels with a common multi-scale convolutional architecture[C]. IEEE International Conference on Computer Vision. 2015:2650-2658.
- [38] Tateno K, Tombari F, Laina I, et al. CNN-SLAM: Real-time dense monocular SLAM with learned depth prediction[C]. IEEE Conference on Computer Vision and Pattern Recognition. 2017:6565-6574.
- [39] Hartley R. I. In defense of the eight-point algorithm[J]. IEEE Transactions on Pattern Recognition and Machine Intelligence, 1997. 19(6):580-593.
- [40] Rosen D, Kaess M, Leonard J. An incremental trust-region method for robust online sparse least-squares estimation[C]. IEEE International Conference on Robotics and Automation. 2012: 1262-1269.
- [41] David G L. Distinctive image features from scale-invariant keypoints[J]. International Journal of Computer Vision, 2004. 60(2): 91-110.
- [42] Herbert B, Tinne T, Luc Van G. SURF: speed up robust features[C]. European Conference on Computer vision. 2006:404-417.
- [43] Rublee E, Rabaud V, Konolige K, et al. ORB: an efficient alternative to SIFT or SURF[C]. International Conference on Computer Vision. 2011:2564-2571.
- [44] Edward R, Gerhard R, Tom D. Real-time video annotations for augmented reality[C]. International Symposium on Vision computing. 2005:294-302.
- [45] Michael C, Vincent L, Christoph S, et al. BRIEF: binary robust independent elementary features[C]. European Conference on Computer Vision. 2010:778-792.
- [46] Bian J, Lin W Y, Matsushita Y, et al. GMS: Grid-based motion statistics for fast, ultra-robust feature correspondence[C]. IEEE Conference on Compute Vision and Pattern Recognition. 2017:2828-2837.
- [47] Liu F, Shen C, Lin G Èet al. Learning depth from single monocular images using deep convolutional neural fields[J]. IEEE Transactions on Pattern Analysis and Machine Intelligence, 2016. 38(10):2024-2039.

- [48] Eigen D, Fergus R. Predicting depth, surface normals and semantic labels with a common multi-scale convolutional architecture[C]. IEEE International Conference on Computer Vision. 2015:2650-2658.
- [49] Laina I, Rupprecht C, Belagiannis, Tombari F, et al. Deeper depth prediction with fully convolutional residual networks[C]. IEEE International Conference on 3D vision. 2016:239-248.
- [50] Lang M, Wang O, Aydin T, et al. Practical temporal consistency for image-based graphics application[J]. ACM Transactions on Graphics, 2012. 31(4):1-8.
- [51] Koutis I, Miller G L, Tolliver D. Combinatorial preconditioners and multilevel solvers for problems in computer vision and image processing[C]. International Symposium on Advances in Visual Computing. 2009:1067-1078.
- [52] Krishnan D, Fattal R, Szeliski R. Efficient preconditioning of Laplacian matrices for computer graphics[J]. ACM Transactions on Graphics, 2013. 32(4):1-15.
- [53] Gastal E, Oliveira M. Domain transform for edge-aware image and video processing[J]. ACM Transactions on Graphics, 2011. 30(4):1-12.
- [54] He K, Sun J, Tang X. Guided image filtering[C]. European Conference on Computer Vision. 2010:1-14.
- [55] Min D, Choi S, Lu J, et al. Fast global image smoothing based on weighted least squares[J]. IEEE Transactions on Image Processing, 2014. 23(12):5638-5653.
- [56] Horn B. K. P. Closed-form solution of absolute orientation using unit quaternions[J]. Journal of the Optical Society of America, 1987. 4(4):629-642.
- [57] Silberman N, Hoiem D, Kohli P, et al. Indoor segmentation and support inference form rgbd image[C]. European Conference on Computer Vision. 2012:740-760.
- [58] Sturm J, Engelhard N, Endres F, et al. A benchmark for the evaluation of RGB-D SLAM system[C]. IEEE/RSJ International Conference on Intelligent Robots and System. 2012:573-580.
- [59] Handa A, Whelan T, Mcdonald J, et al. A benchmark for rgb-d visual odometry, 3d reconstruction and SLAM[C]. IEEE International Conference on Robotics and Automation. 2014:1524-1531.
- [60] Farbman Z, Fattal R, Lischinski D. Edge-preserving decompositions for multi-scale tone and detail manipulation[C]. ACM SIGGRAPH. 2008:1-10.

攻读硕士学位期间发表学术论文情况

- Dense Reconstruction from Monocular Slam with Fusion of Sparse Map-Points and Cnn-Inferred Depth. **Xiang Ji**, Xinchen Ye, Hongcan Xu, Haojie Li. IEEE International Conference on Multimedia and Expo, 2018, 1-6. EI index, EI number: 20190706509031. Ä\.... \Aê ·1\ 9 Å 1 Å
- 2 Underwater image enhancement using stacked generative adversarial networks. Xinchen Ye, Hongcan Xu, Xiang Ji, Rui Xu. Pacific-Rim Conference on Multimedia, 2018, 514-524. EI index, EI number: 20184105924432.

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